

Learning Intervention

# Evaluation of the Raise Mentoring Program: Final Outcome Evaluation Report

Faculty of Education, The University of Melbourne

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#### ACKNOWLEDGEMENT OF TRADITIONAL OWNERS

The University of Melbourne acknowledges the Aboriginal and Torres Strait Islander traditional owners of the unceded land on which we work and learn. We pay respect to the Elders, past and present, and the place of Indigenous knowledge in the academy.

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# **List of Abbreviations**

Abbreviation	Full text
ACARA	Australian Curriculum, Assessment and Reporting Authority
BBBS	Big Brother Big Sister
HKM	Healthy Kids Mentoring
SBM	School-Based Mentoring

### **Executive Summary**

#### Key message

- Compared to similar students, those who participated in Raise had better helpseeking behaviour, connectedness, and peer support immediately at the end of the program.
- There were promising findings in relation to social-emotional wellbeing and school engagement.
- Student gender and difficult life event influenced the outcomes for Raise students.

#### **Confirmed Student Outcomes**

Small benefit	Medium benefit	Large benefit	
	Connectedness Peer Learning Support	Helping-Seeking Behaviour	

### **Promising Student Outcomes**

Optimism	Engagement	
Resilience	School belonging	

#### Introduction

The independent outcome evaluation of the Raise Mentoring Program was conducted by the University of Melbourne, Australia's highest ranked university, as commissioned by the Raise Foundation, using funding provided by the Australian Federal Department of Health.

This final report details the process and outcomes of the evaluation of the Raise Mentoring Program. The main aim of the Raise Mentoring Program is to improve outcomes in resilience, school belonging, hope for the future and help seeking for young people at risk of disengagement from education or poor wellbeing.

# Aims and Objectives of the Evaluation

This evaluation aims to determine the differences in post-program outcomes between students who receive the Raise Mentoring program compared to a matched comparison group.

The primary outcome for the comparison was student help-seeking behaviour. This was prioritised after discussions with the Evaluation Advisory Group and Raise's Theory of Change, which determined it to be the key outcome which would be expected to change by the end of the program.

Raise's Theory of Change indicates that through changes in help-seeking behaviour, this would lead to improvements in our secondary outcomes of student well-being, school engagement and resilience.

#### **Key Evaluation Questions**

The key evaluation questions are:

- What are the outcomes of the program related to students:
  - help-seeking behaviour;
  - o social-emotional wellbeing;
  - school engagement; and
  - o resilience

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- What are the student characteristics that are related to the outcomes we have observed?
- What are the student perceived enablers and barriers related to the outcomes of the program?

#### Methodology

Raise's Program Logic Model was utilised throughout the current independent evaluation.

A quasi-experimental design was employed to inform the benefits of the intervention compared to current practice and identify for whom and under what conditions the intervention is beneficial.

The evaluation comprised multiple components involving pre- and post-program surveys of the wider cohort's help-seeking behaviour, social-emotional wellbeing, resilience, and school engagement using validated measures. Qualitative survey data from Raise mentees was also collected.

Data collected by the evaluation team was supplemented with data collected internally and provided by Raise.

Overall, 75 schools were approached to participate, with 13 schools providing consent across 4 Australian states.

The final sample consisted of:

- 148 Raise mentees and 278 matched comparison students from the participating schools.
- 9 Raise mentees providing qualitative data about their experience.

#### **Key Learnings**

#### **Program effectiveness**

 Overall, there was statistically significant evidence that Raise mentees had better helpseeking behaviour, connectedness (feeling valued and supported) and peer learning support, when compared to the matched comparison group.

- There was a trend of findings benefiting Raise students above an effect size of 0.2, although not all were statistically significant.
- This magnitude of difference for wellbeing outcomes indicates a medium level of benefit, which has implications for long-term benefits.
- Those with low functioning in each domain at baseline were more likely to report improvements at follow up.
- Student gender and difficult life event at baseline showed evidence of influencing the degree of benefit for these students.
- Students who had difficult life events in the past month (such as financial difficulty, house insecurity) were less likely to report positive benefits.

#### **Program experience:**

- Raise mentees identified the opportunity to talk to someone about their issues, problems, and shared interests as the most significant benefit of participating in the program.
- Raise mentees suggested they may benefit if the program went for longer or had more contact points.

#### Recommendations

# Recommendation 1: Use of existing and new mental health and wellbeing data to prioritise students:

Increase support for schools to use existing health and wellbeing administrative data (e.g. National Child Mental Health Check), to identify and prioritise students who are more likely to benefit from the program.

# Recommendation 2: Further examining of why demographic variables are predictive of the changes observed Raise mentees:

Student gender and difficult life event influenced the outcomes achieved for these students. Consideration should be given as to why these differences occur, and whether this warrants adaptation to Raise.

# Recommendation 3: Continue to build evidence base of the Raise Mentoring Program:

This is the first quantitative evaluation conducted by Raise by an independent team using a comparison group. The evidence for Raise could be further strengthened through either replication studies using a quasi-experimental approach (as used here) or via a larger, cluster randomised

## **Executive Summary**

controlled trial of the program. Pragmatic and financial implications of each approach should be used to inform how Raise continues to build its evidence. Given Raise has conducted strong internal evaluations for the past 15 years, the use of independent evaluations would be important to continue to build the evidence for the Raise Mentoring Program.

# Recommendation 4: Consider ways to increase the reach and length of the program:

Given students reported a preference for increasing the length and/or quantity of the program, Raise should consider ways to enable this to occur. This would need identification of the barriers for expansion of all programs. For instance, training of existing school staff to deliver the program may increase the number of schools who can offer Raise instead of relying solely on external mentors. Another approach would be to consider a second year of enrolment, for students who may benefit from a second year in the program with a different mentor to help consolidate or extend benefits.

# **Learning Intervention**

The Learning Intervention Academic Group brings together academics in Learning Interventions from a range of education focused contexts, including early years, primary, secondary, and tertiary settings and into adulthood. It is situated in the Faculty of Education (FoE, previously the Melbourne Graduate School of Education) at the University of Melbourne, which is the highest ranked university in Australia and 14th in the world, as per the Quacquarelli Symonds (QS) World University Rankings. The FoE is also the number 1 ranked faculty in Australia for Education, according to the Times Higher Education World University Rankings by subject 2024.

# **Report Overview**

This independent evaluation was commissioned by the Raise Foundation (Raise), using funding from the Federal Department of Health, to examine the evidence of outcomes of the Raise Mentoring Program. The independent evaluation was intended to strengthen Raise's internal evidence base and identify opportunities for innovation to inform their monitoring and evaluation framework into the future.

This updated interim outcome evaluation report has been prepared for Raise and details the evaluation methodology, data and findings of evaluation questions and actionable insights and recommendations.

The overall report is structured in five sections, followed by the references and appendices.

- **Section 1 Evaluation Overview** provides a background to the Raise Mentoring Program, the purpose of the evaluation and the key evaluation questions/areas of investigation.
- Section 2 International Review of the Evidence for School-Based Mentoring Programs for Adolescents provides a succinct overview of current literature in this area.
- Section 3 Methodology presents the rationale and design of the evaluation, the sampling, data collection, and analytical methods.
- Section 4 Findings presents the evidence relating to the key evaluation questions.
- **Section 5 Discussion and Recommendations** concludes with an explanation of the findings and the resulting recommendations.

#### 1. The Evaluation Overview

#### 1.1 The Raise In-School Mentoring Program (The Evaluand)

The Evaluand for this project is the Raise Youth Mentoring Program (Raise Program). The Raise Program was developed by the Raise Foundation and aims to improve outcomes in resilience, school belonging, hope for the future and help-seeking for young people at risk of disengagement from education or poor wellbeing.

Raise work with each school's wellbeing teams to identify up to 15 students between the ages of 13 and 15 who will most benefit from the mentoring program. These students are matched with a volunteer mentor who has been recruited from the community and trained by Raise. Mentor training consists of eight online modules covering a range of topics including mentoring skills, mentor self-care and youth safety followed by six hours of group face-to-face training (either in-person or online) which provide practical skills and the opportunity to work through scenario-based activities. Each program is facilitated by a degree qualified Raise Program Counsellor, who is on-site for each mentoring session and provides additional supervision support to mentors at the conclusion of the mentoring session. The Program Counsellor matches mentors and students (mentees) based on their observations during an initial session called Jitters alongside mentee preference, amongst other factors. During the Raise Mentoring Program, the mentee and mentor work together through an evidence-informed curriculum over 23 weekly sessions which build practical skills, understanding and knowledge.

The Raise Mentoring Program has operated in Australia since 2008 and now operates in over 150 schools in all Australian jurisdictions (except for Tasmania and the Northern Territory). Raise conducts regular internal evaluation and has used these results, plus feedback and experiences of participants, to measure and improve the impact of their program. Raise have previously used a combination of internally developed measures and validated measures, such as the Children's Hope Scale (Snyder et al., 1997), Brief Resilience Scale (Smith et al., 2008), School Connectedness Scale (Resnick et al., 1997), and Growth Mindset Scale (Dweck, 1999). Raise's internal evaluations have shown positive benefits for mentees, with their 2023 evaluation finding statistically significant improvements in hope for the future, resilience, asking for help and school belonging for the whole mentee cohort.

#### 1.2 Aims and Objectives of the Evaluation

The Raise Foundation sought an independent quantitative evaluation of the impact of the Raise Program to strengthen their evidence base and identify opportunities for innovation to inform their monitoring and evaluation framework into the future, using funding provided by the Australian Federal Department of Health.

The outcome evaluation was commissioned from the Faculty of Education at the University of Melbourne. The contract commenced in March 2022 with a project design and set-up phase, with data collection occurring during 2023 and concluding in December. This outcome evaluation aims to determine the differences in post-program outcomes between students who receive the Raise Mentoring Program compared to both matched cohort within schools, and all students in the same year level.

The scope of the evaluation sought by the Raise Foundation was to focus on the impact of the Raise Program on mentees, considering the following questions:

- Have the outcomes changed for the mentees?
- How have outcomes differed for different cohorts of mentees?

### 1.3 Key Evaluation Questions

In response to the scope of evaluation sought by the Raise Foundation and based on the program logic model the key evaluation questions are:

- What are the outcomes of the program related to student's help-seeking behaviour, social-emotional wellbeing, school engagement, belonging and resilience?
- What are the student characteristics that are related to the outcomes we have observed?
- What are the student perceived enablers and barriers related to the outcomes of the program?

# 2. International Review of the Evidence for School-Based Mentoring Programs for Adolescents

The following section provides a rapid review of existing research which has described the characteristics and outcomes of other school-based mentoring programs provided to adolescents in Australia and internationally. The purpose of the review was to inform the development of the evaluation of the Raise Mentoring Program through identifying potential theories of change for school-based mentoring programs, as well as key populations, characteristics and barriers that are likely to influence the theory of change. In addition, the review highlights the potential differences between the Raise Program and other evaluated school-based mentoring programs.

#### 2.1 Overview of School-Based Programs

Student success is not only dependent on academic skills, but also social skills and emotional wellbeing (Claro & Perelmiter, 2022). Mentoring programs propose that adult role models can support youth to build strengths such as competence, confidence, and connection which can be preventive factors against school disengagement and misbehaviour (Claro & Perelmiter, 2022; Lerner et al., 2005).

School-based mentoring (SBM) programs have gained increasing popularity since the 1980s, largely due to their inexpensive implementation, ease of mentee recruitment, lower time commitment for mentors and promising social and academic benefits for students (Herrera & Karcher, 2014; Wood & Mayo-Wilson, 2012; Herrera, 2004). Positive student outcomes include improved academic performance, increases in confidence, resilience, connectedness, and perception of support (non-familial adult), along with reductions in school absenteeism and poor behaviour such as discipline referrals and school-related misconduct (Gordon, Downey & Bangert, 2013; Schwartz, Rhodes & Herrera, 2012).

Students are often referred for SBM by teachers due to behavioural or academic difficulties, allowing these programs to reach youth who are most in need of support (Randolph & Johnson, 2008; Wood & Mayo-Wilson, 2012). Additionally, SBM programs provide students with one-on-one or small group support, allowing for individualised student specific goals and outcomes.

Research highlights the importance of school-community partnerships, like mentoring programs, for their ability to provide and contribute to improvements in family and child wellbeing through the fostering of supportive relationships outside of the immediate family. Whilst many schools implement peer-to- peer mentoring programs internally, there is significant literature outlining the positive outcomes for students, especially males, where intergenerational, one-on-one youth mentoring programs are implemented (Raposa et al., 2019; Gwyther et al., 2019). Community led SBM programs implementing intergenerational, one-on-one mentoring often report stronger relationships between mentor and mentee, contributing to the longevity of experienced positive outcomes (Bayer et al, 2015; Raposa et al, 2019).

#### 2.2 Characteristics of Individual School-Based Mentoring Programs

Mentoring programs can vary greatly, including length of programs and weekly contact time, content of mentoring session, mentor training and ongoing support, and outcome focuses of the program (Herrera, 2004; Randolph & Johnson, 2008; Raposa et al, 2019). Occurring typically at least once a week in a supervised school setting, mentoring activities can be structured to the specific goals of the program or up to the mentor-mentee pair including academic help, social activities, and school events support (Herrera, 2004; Randolph & Johnson, 2008).

An evaluation of the Big Brothers Big Sisters (BBBS) program implementations for students from Grade 3 to high school in the USA reported that 70% of mentors were spending half an hour to 1 hour a week with their mentees, whilst 20% reported spending between one to two hours a week. Session content also varied considerably with approximately 50% of mentors reported spending time assisting mentees with

homework, 85% reported spending time on social activities, and 33% reported that they attended school activities such as sport or extra-curricular events with their mentees (Herrera, 2004).

In contrast to the BBBS program, the Healthy Kids Mentoring (HKM) Program, designed for Grade 4 students, outlined that mentors and mentees would meet for one hour each week to cover four program components: relationship building, self-esteem enhancement, goal setting, and academic assistance building. HKM mentors received an initial training session and a program guidebook to refer for examples of activities (King et al., 2002).

The Building Positive Relationships mentoring program was designed to develop relationship skills in girls from Grade 3 to high school. This program was delivered in a group format and split by age groups, Grades 3-4, Grades 5-8, and high school aged girls. Session lengths also varied by age group with Grades 3-4 meeting for one hour sessions, Grades 5-8 for 1.5 hours, and high school aged group for 1.5 hours. The content of the sessions also differed by age group to align with their developmental periods (Westhues et al., 2001).

# 2.3 Immediate and Long-Term Outcomes Achieved via School-Based Mentoring Programs

Due to the variability in SBM programs, significant findings have been inconsistent and can depend on selected outcomes; however, overall positive student outcomes are consistently reported (Raposa et al., 2019; Gwyther et al., 2019; Claro & Perelmiter, 2022).

An evaluation of the BBBS program (USA) found that surveyed mentees reported significant positive changes in peer network (i.e., ability to make friends;  $r^2$  = .48); social skills (i.e., mentees' confidence in communicating with other and ability to express their feelings;  $r^2$  = .29); positive classroom behaviour (i.e., following class rules and working without disturbing others;  $r^2$  = .55); and attitudes towards school (i.e., how much mentees' liked school;  $r^2$  = .39) for mentees' who continued sessions for over 9 months, compared to mentees who engaged with their mentor for 6 months or less (Herrera, 2004). Alternatively, a systematic review and meta-analysis found that evidence from eight SBM studies did not improve academic achievement, school attendance, mentee behaviour, and/or psychological outcomes (Wood & Mayo-Wilson, 2012).

A national study in the USA of 255 mentoring programs suggested that SBM programs are most likely to affect school connectedness (measured by school efficacy and bonding); however, effect sizes varied between studies from d = .38 to d = .35 (Herrera & Karcher, 2014). The HKM program also reported significantly higher school connectedness (measured by relationship with teachers and attitude towards school) and family connectedness (measured by feeling close with their family) in mentored students compared to non-mentored students at post-test. Furthermore, mentored students were significantly less likely to have physically fought with a peer and more likely to talk with their parent/guardian to seek advice when they have problems, however effect sizes for later observed outcomes were not reported (King et al., 2002).

A meta-analysis that examined effects of SBM programs on emotional wellbeing found that mentees experienced greater decreases in internalising behaviour, as measured by negative affect (d = 0.20), than externalising behaviour. Additionally, mentoring programs were found to have a positive effect on emotional wellbeing (d = 0.34) and self-esteem (d = 0.45), suggesting that mentoring relationships may influence emotional wellbeing though self-esteem (Claro & Perelmiter, 2022).

Long-term student benefits from SBM programs included increases in likelihood of high school completion and further vocational education, gaining employment and progressing career paths, and increased likelihood of good social-emotional and mental health such as lower levels of depression, higher self-

esteem and confidence, greater resilience and feeling more connected and supported (Raposa et al, 2019; DuBois et al, 2002).

#### 2.4 Differences in Outcomes Based on Student Characteristics

When examining which groups may best benefit from SBM programs, past literature has largely examined the age and gender of mentees (Herrera, 1999; Herrera et al., 2007; Karcher, 2008). Herrera et al. (2007) found that females showed improvements in academic performance and decreased school misbehaviour, which was not observed in males. Karcher (2008) further examined age and gender together and demonstrated positive outcomes in young males and teenage females, but negative outcomes in high-school aged males. It was suggested that younger youth may view focused attention from an adult positively and feel proud to have a mentor, however, high school adolescents may feel embarrassed by the additional attention due to the stigma of mental health interventions (Herrera, 1999; Karcher, 2008).

Raposa et al. (2019) reported that male and female students were often referred to SBM programs for different reasons, which may also impact on student outcomes. Female students were more commonly referred because of "relational challenges with their primary caregivers", whilst males are most often referred from "the need for a male role model" (Raposa et al. 2019). Raposa et al.'s (2019) meta-analysis of SBM program outcomes suggested that female students were more "prone to co-rumination in their dyadic relationships" than male students. This excessive focus on close relationships may reduce the positive effects of SBM programs for female students and further highlight why programs with a higher male student population often report statically significant stronger effects sizes for male students (B = .38, t = 2.19, p < .05) (Raposa et al. 2019; Splendelow et al. 2017; DuBois et al. 2011).

High risk environmental factors such as poverty, socioeconomic status and neighbourhood violence may also affect SBM program outcomes, with afflicted students benefiting by SBM programs more than their peers (Raposa et al. 2019). However, no significant differences in effect sizes were reported based on indicators of youth risk at baseline (i.e. percentage of single-parent households, students receiving free or reduced-price lunch, below grade level academic performance and exhibiting problematic behaviours) (Raposa et al. 2019).

Literature also indicates that program length, mentor support, and mentor-mentee match characteristics may influence the outcomes on SBM programs (Herrera, 2004; Wood & Mayo-Wilson, 2012). Studies have consistently found that most benefits of SBM are not observed until at least one year of regular mentoring sessions (Herrera, 2004; Lee & Cramond, 1999). Furthermore, higher levels of mentor training and support have been associated with relationship quality and match continuation (Herrera et al., 2007). Teacher support to mentors has also been linked with higher engagement and increased benefits in mentees (Herrera et al., 2008).

Two aspects of mentor-mentee match characteristics – relationship quality or "closeness" and match suitability – are commonly reported to influence SBM program outcomes (Grossman & Johnson, 1999). Mentees who have a closer and more supportive relationship with their mentors were more likely to have positive outcomes (Herrera et al., 2007). Additionally, matching by gender and/or race was not found to impact the closeness of the mentoring relationship; however, mentors and mentees who shared interests had greater engagement and were more likely to demonstrate better outcomes (Herrera, 2004; Sipe 1996).

#### 2.5 Enablers and Barriers Related to Student Outcomes

In a report of the BBBS program, mentors reported enablers and barriers related to student outcomes. These included the agency providing additional support/structure for match interactions, ensuring a consistent meeting place, facilitating structured communication with teachers throughout the duration, and providing feedback and advice from parents to create longer and stronger mentor-mentee relationships (Herrera, 2004).

Mentor characteristics have also been reported to impact the effectiveness of SBM programs outcomes, including mentor age and profession/experiences (Raposa et al., 2019). Programs that had more male mentors (B = .36, t = 2.14, p < .05) and mentors who worked "within helping professions" (F(1, 289) = 5.49, p < .05; B = .25, t = 2.34, p < .05) reported higher program impact effect sizes (Raposa et al. 2019).

#### 2.6 Known Evidence Gaps in School-Based Mentoring Programs

While evaluations of SBM programs are plentiful, there is a lack of recent rigorous, systematic program evaluations on the effects of mentoring, especially outside of the USA (or USA populations), where most evidence for SBM largely originates. As context (including school culture, support from teachers/school staff, relationship models of mentoring style) can influence the impact of mentoring program outcomes and can vary considerably between schools, the generalisability of already mixed results cross-culturally remains ambiguous. While there are meta-analyses of SBM programs, effects sizes of outcomes are often inconsistent between studies and vary depending on the program's outcome domains (Raposa et al., 2019; DuBois et al., 2011).

#### 2.7 How Does Raise Compare to Previously Evaluated Interventions?

Raise is an Australian school-based mentoring program that has been in operation since 2008 and is currently endorsed by two Australian state departments of education. Unlike previous evaluated SBM programs, Raise is a curriculum-based mentoring program that focuses on delivering early intervention whilst engaging and fostering relationships with schools, forming "part of a wider referral system". Guided by wellbeing support principles such as the Australian Wellbeing Framework, Raise is a research-driven, evidence-informed 6-month program that regularly collects data and undergoes internal evaluations to continue to improve the program and its impacts. Unlike many SBM evaluations outcomes, Raise focuses on outcomes on three levels: school, mentor, and mentee (Year 7-9 high school student).

Additionally, Raise provides comprehensive mentor training in mentoring skills, youth safety and the Raise curriculum. Qualified professional counsellors provide ongoing support and weekly training, guidelines for mentoring sessions and enables ongoing one-to-one support with the same mentor throughout the entire program. This ensures that the program can provide individual solutions for the needs of each student.

To continue to improve evaluative evidence, Raise has initiated an external evaluation of the impact of their SBM program on student outcomes, addressing the lack of rigorous SBM program evaluation within Australia.

### 3. Methodology

#### 3.1 Evaluation Design

The current evaluation aimed to independently examine the evidence of impact on student outcomes of the Raise Youth Mentoring Program in the schools that participated in this evaluation.

A quasi-experimental comparative pre-post design was employed to achieve the evaluation objectives. This design involved the collection of data to establish a comparison group, enabling an understanding of the counter-factual to be established. Therefore, the project compared the student-reported outcomes between those who received the Program against matched students who did not receive the Program.

Matching occurred based on baseline student and school demographics of students who received the program, but also considered the response rates across the evaluation data collection periods. Therefore, matching of the cohort occurred after all data collection had occurred.

See Figure 1 for an overview of the design approach.

The evaluation was comprised of a number of components involving baseline and follow-up cohort survey data collection, Raise mentee pulse surveys, Raise mentor surveys, and interview and focus group data from Raise mentees, as well as the analysis of additional mentee and mentor survey data provided by Raise.

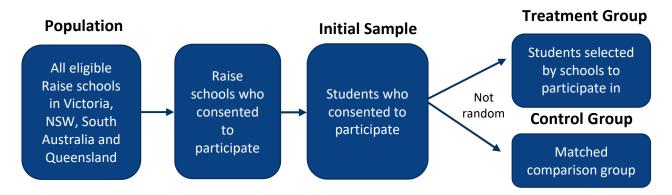


Figure 1. Overview of the Raise evaluation design.

#### 3.2 Measures

The measures chosen for this evaluation were based on their alignment to Raise's Theory of Change and Program Logic (see Appendices A and B), discussions with Raise stakeholders, as well as a review of published research papers which have evaluated mentoring programs delivered in secondary school settings for students with increased school disengagement risk factors. Where possible, Raise's own internal data collection measures were aligned with evaluation data collection measures. This enabled further analyses to occur that included both sets of data.

#### 3.2.1 Demographic Information

Student demographic questions were designed to better understand for whom and under what conditions the intervention is beneficial, given Raise's desire to understand whether differing characteristics affect program outcomes for mentees. These questions were in alignment with the Australian Curriculum, Assessment and Reporting Authority's (ACARA) agreed Student Background Characteristics, specifically with regards to gender and language other than English spoken at home. A question about Aboriginal and/or Torres Strait Islander status was initially included but was removed to avoid further delays in ethics and

departmental approval processes (see section 3.3). The following student demographic information was collected:

- Year level (Year 7/Year 8/Year 9/Year 10)
- Gender (Male/Female/Other/Rather not say)
- Whether the student was born in Australia (Yes/No)
- Whether the student speaks a language other than English at home (Yes, mostly/Yes, sometimes/No)
- Who the student lives with (One parent/Both parents/Carer or guardian/Other)
- Whether the student is a person with disability (Yes/No/Rather not say)
- Whether the student is a carer for family or friends with disability, a health condition or mental illness (Yes/No/Rather not say)
- How many school days the student was absent in the past two weeks (Free text)
- Whether the student has experienced any difficult home or personal events (Yes/No) and if so, the impact of these on their ability to engage in school activities (None of the time/A little of the time/Some of the time/Most of the time/All of the time)

#### 3.2.2 Help-Seeking Intentions

#### General Help-Seeking Questionnaire (Deane et al., 2001)

The General Help Seeking Questionnaire (GHSQ) assesses formal help-seeking intentions for nonsuicidal and suicidal problems using a matrix design. In this evaluation, only the questions for nonsuicidal problems were used and the measure was adapted for relevancy to an adolescent population by removing husband, wife, and de-facto as examples of intimate partners (in Question A). Ten items rated on a 7-point Likert scale from 1 (extremely unlikely) to 7 (extremely likely) were presented to students, with higher scores representing higher intentions of seeking help.

Wilson et al. (2005) used an 18-item version of the GHSQ with a sample of Australian public high school students (aged 12-19 years) and found that the GHSQ is a flexible measure of help-seeking intentions applicable across a range of contexts. The survey has also been used as part of the Longitudinal Study of Australian Children, a federally funded project since 2014, as well as by the Federal Department of Health in various data collections with adolescents. The scale is considered reliable, with a Cronbach's alpha of .70 and test-retest reliability assessed over a three-week period being .86 (Wilson et al., 2005). Convergent and divergent validity of the GHSQ were also supported, with intentions to seek counselling correlating positively with prior mental health experiences, and negatively with self-reported barriers to seeking professional help (Wilson et al., 2005).

#### **Raise-Developed Help-Seeking Questions**

This measure was adopted from Raise's own internal surveys, to further understand student's general help-seeking confidence. Three questions were presented on a 4-point Likert scale from 1 (never) to 4 (always), with higher scores representing higher help-seeking confidence.

- Do you feel able to trust adults who can help when you need it?
- Do you feel able to ask for help from others when you need it?
- Do you feel like you know where to get help when you need it?

#### 3.2.3 Social-Emotional Wellbeing

**EPOCH Measure of Adolescent Wellbeing (Kern et al., 2016)** 

The EPOCH Measure of Adolescent Well-being (EPOCH) measures interpersonal and intrapersonal characteristics in young people (aged 10-18 years) that might foster wellbeing and physical health in adulthood. It utilises a 5-point Likert-type scale with 20 items across the five subscales of engagement, perseverance, optimism, connectedness, and happiness, yielding a range of 5-20 on each, with higher scores representing higher wellbeing. Happiness did not align with Raise's Theory of Change and was therefore not utilised in this evaluation. However, it remained in the survey as advice from the measure developer advised that removing it from the way the tool is presented may negatively influence the validity of the other domains captured in the survey.

Evidence for the EPOCH measure is drawn from ten samples of adolescents from the US and Australia (N = 4,480). Internal consistency estimates (alpha) range from 0.74 to 0.92 (Kern et al., 2016). Across 3-week, 4-month, and 2-year, 9-month intervals, test-retest reliabilities range from 0.23 to 0.71 (Kern et al., 2016). Scale scores have shown expected significant relationships with existing measures and unidimensionality was established using confirmatory factor analysis (Kern et al., 2016).

#### Short Warwick-Edinburgh Mental Wellbeing Scale (Stewart-Brown et al., 2009)

The SWEMWBS is a short version of the Warwick-Edinburgh Mental Wellbeing Scale (WEMWBS), which was developed to enable the monitoring of mental wellbeing in the general population and the evaluation of projects, programmes and policies which aim to improve mental wellbeing. Seven of the WEMWBS's original 14 statements that focus on functioning rather than feelings are used in the SWEMWBS. These positively worded statements are rated based on the respondent's experiences over the past two weeks on a 5-point scale from 'None of the time' to 'All of the time'. Scores range from 7 to 35, with higher scores indicating higher positive mental wellbeing.

The SWEMWBS has been validated for populations of young people aged 15-21 (McKay & Andretta, 2017; Ringdal et al., 2018). While the test-retest reliability of the SWEMWBS has not been reported for most populations, the WEMWBS test-retest reliability within 7-8 days after first completion was moderate in a UK population of 13- to 16-year-olds (Clarke et al., 2010). Likewise, one-week test-retest reliability of the WEMWBS was high in a UK population of university students (Tennant et al., 2007).

#### 3.2.4 School Affective and Cognitive Engagement

#### Student School Engagement Measure (Hazel et al., 2013)

The SSEM was developed from a model of student school engagement, comprising aspirations, belonging, and productivity. The SSEM has 22 items across three factors: aspirations (4 items), productivity (12 items), and belonging (6 items). The survey uses a Likert-style scale ranging from 1 (Strongly disagree) to 10 (Strongly agree). Scores from all items are combined to get an overall engagement score, with higher scores indicating greater engagement.

The SSEM was validated with data from 396 eighth graders in an urban school district in the US by Hazel et al. (2013). Structural equation modelling showed that the SSEM model fit the data well, had good reliability for the three factors, and was predictive of district-identified risk factors and state standardized academic assessment results. Cronbach's alpha for the three empirical factors ranged from .83 for belonging, to .92 for productivity (Hazel et al., 2013). Another study (Hazel et al., 2014) found strong and significant positive correlations (.80) between the SSEM and the 2 measures of engagement (the School Engagement Measure and the Student Engagement Instrument) and a weak but significant positive correlation (.35) between the SSEM and a measure of life satisfaction (the Student Life Satisfaction Survey). These findings support the use of the SSEM as a valid measure of adolescents' engagement with school.

#### Student Engagement Scale – Peer Support for Learning (Appleton et al., 2006)

The Student Engagement Scale (SEI) is a self-report survey measuring student's engagement at school, with the Peer Support for Learning (PLS) subscale specifically measuring psychological engagement in school.

The SEI – Peer Support for Learning subscale has 7 items that are rated on a 5-point Likert scale, from 'Disagree a lot' to 'Agree a lot'. It is scored by adding all items together, resulting in a possible score range from 7 to 35, with higher scores indicating higher levels of engagement.

Appleton et al. (2006) reported an internal consistency coefficient alpha of .82 for the Peer Support for Learning subscale and demonstrated the construct validity of the six subscales using a confirmatory factor analysis. Betts et al. (2010) confirmed evidence of the validity of all subscales except extrinsic motivation. Engagement subscales correlate with measures of academic performance and behaviour, demonstrating criterion-related validity through positive relationships with grade point average and reading and mathematics achievement, and negative relationships with frequency of suspensions. The instrument has been used in several research studies on engagement in school (Reschly et al. 2008; Lewis et al. 2009).

#### 3.2.5 Resilience

#### Brief Resilience Scale (Smith et al., 2008)

The Brief Resilience Scale (BRS) assesses the perceived ability to bounce back or recover from stress. This scale was developed to assess a unitary construct of resilience, including both positively and negatively worded items. The BRS has six items rated on a 5-point scale from 'Strongly disagree' to 'Agree'. It is scored by reverse coding the three negatively worded items and then finding the mean score of all six items, resulting in a possible score range from 1 to 5, with higher scores indicating higher resilience.

The BRS has demonstrated good internal consistency, with Cronbach's alpha ranging from .80-.91, and test-retest reliability, with an intra-class correlation of .69 for one month and .62 for three months (Smith et al., 2008). A review found it to be among the resilience measures with the best psychometrics (Windle et al., 2011). Although it was developed for an adult population, it has been used with children to measure their resilience (e.g. Sharma & Nagle, 2018).

Table 1. Alignment between evaluation questions/outcomes and data collection activities.

Key Evaluation Outcomes			Key Evaluation Questions			
		Domain	What are the outcomes related to student's help-seeking behaviour?	What are the outcomes related to student's social-emotional wellbeing, school engagement and resilience?	What are the student or mentor characteristics that are related to the outcomes we have observed?	What are the student perceived enablers and barriers related to the outcomes of the program?
Student der	mographics	Student/mentor characteristics			х	
	General Help-Seeking Questionnaire	Help-seeking	Х			
	Raise-developed Help-Seeking Questions	Help- seeking	Х			
Collected	Short Warwick-Edinburgh Mental Wellbeing Scale	Wellbeing, school engagement and resilience		х		
at pre and post Raise	EPOCH Measure of Adolescent Wellbeing	Wellbeing, school engagement and resilience		Х		
delivery	Student School Engagement Scale	Wellbeing, school engagement and resilience		х		
	Student Engagement Scale – Peer Support for Learning	Wellbeing, school engagement and resilience		х		
	Brief Resilience Scale	Wellbeing, school engagement and resilience		х		
Mentee 'Pu	lse' surveys	Student/mentor characteristics			X	
Mentor sur		Student/mentor characteristics			х	
Raise admir	nistrative data	Help-seeking Wellbeing, school engagement and resilience Student/mentor characteristics	х	х	х	
Mentee qua	alitative perspective data	Enablers and barriers				X

#### 3.3 Ethics and State Approvals

Ethics approval was sought from the University of Melbourne's Human Research Ethics Committee (HREC) in October 2022 and was granted in December 2022 (HREC #2022-24962-35335-3). Individual state-based education department approval was then required before school recruitment could commence. The National Application Form, a single form for applying to conduct school-based research in more than one Australian jurisdiction, was completed and individually submitted in January 2023 to each governmental education department in jurisdictions the Raise program operates in. This included the Australian Capital Territory, New South Wales, Queensland, South Australia, Victoria, and Western Australia. Each jurisdiction assesses applications to conduct research in its school sites according to their own research approval guidelines, although they apply broadly similar criteria. A process of feedback, revision and resubmission was therefore followed with each jurisdiction as required. Feedback received and implemented generally related to clarification of evaluation methodology, mechanisms to identify and support students experiencing distress, and clarifications to study consent documents.

The earliest approval was approximately five weeks after application submission, and the latest was approximately 13 weeks after application submission. Feedback was received from the Australian Capital Territory at a stage deemed too late to be worthwhile implementing, given the small number of eligible Raise schools in the jurisdiction. Western Australia was the first jurisdiction to provide feedback and request revisions to the research application. However, they only provided an update in November 2023 due to their own internal staff shortages, which they had raised during the initial application process.

Final approval was granted by the following states:

- New South Wales
- Queensland
- South Australia
- Victoria

It is important to note that there were a number of research restrictions that limited the scope of the approved data collection. Most states did not allow the collection of student's personal information from the school prior to the student consenting, and restricted what student information was able to be collected once they had consented. For instance, asking students if they identify as Aboriginal and/or Torres Strait Islander was flagged by multiple states as requiring further advice or approval from their First Nations departmental areas. To avoid further delays to state-based approval and commencement of data collection, the question was therefore removed from the cohort survey. Many states also do not allow researchers to incentivise student participation. However, a cohort-level report of student wellbeing was able to be provided to schools at follow-up to incentivise their engagement and encouragement of student participation. Table 2 contains a summary of the states enforcing each research restriction.

Table 2. Research restrictions by state

State	Identifying First Nations students	Incentivising students
New South Wales		X
Queensland	X	X
South Australia		
Victoria	X	X

#### 3.4 School Recruitment

Schools were eligible to participate in the study if they had been implementing the Raise Program for at least one year, a decision made through consultation with Raise and the Evaluation Advisory Group. This was to avoid an additional level of burden for schools with a new relationship with Raise, and because

Raise's experience indicated that schools needed a year to properly understand the Raise Program and to readily identify which students may benefit from participation.

Three online school information sessions about the independent evaluation were hosted by the Raise Foundation on 8<sup>th</sup> December 2022, 28<sup>th</sup> February 2023, and 9<sup>th</sup> March 2023. The University of Melbourne evaluation team attended these sessions to present information about recruitment and data collection and to answer questions from schools. As these sessions occurred prior to receiving state-based education department approval for the evaluation, an online Qualtrics form was set up where schools could submit an Expression of Interest (EOI) in participating in the evaluation. Twenty-five schools completed the EOI form. When the relevant state approval was received, these schools were emailed a letter including the Plain Language Statement and consent form for the evaluation, which was then followed up with a phone call and subsequent reminders.

In addition to the EOI schools, the Raise Foundation identified 50 further schools appropriate to contact about the evaluation. This resulted in a total of 75 schools across the four states being contacted by the University of Melbourne to determine their interest in participating in the independent evaluation. Schools were contacted via phone, which was followed up with an email including information about the evaluation if the school's contact was unavailable, followed by multiple reminders. Raise State and Program Managers were also engaged by the Raise Foundation to support the evaluation and liaised with and sent information to eligible schools.

Many schools were not responsive to contact, particularly those who did not submit EOIs. Reasons cited by responding schools who declined to participate were predominantly around a lack of capacity within the school to facilitate the cohort-wide survey, but also included concerns about the potential for additional surveys for Raise mentees to be burdensome and impact their experience in the program. One school consented to participate but later dropped out due to not having originally realised the cohort-wide scope of the survey. Discussions with Raise around recruitment were ongoing at meetings and regular updates were provided to Raise via email to facilitate support from Raise State and Program Managers. Ultimately, 13 schools consented to participate in the independent evaluation prior to the baseline survey (See Table 3).

At follow-up data collection, schools who had initially expressed interest but had either indicated a lack of capacity at that time or dropped off communications were reapproached either via email or phone call to invite them again to participate. Only one additional school responded and consented to participate, but later dropped out due to being unable to run the survey past their region's Legal Branch in time.

Table 3. Consented independent evaluation schools.

State	Number of schools
South Australia	1
Victoria	4
New South Wales	7
Queensland	1

#### 3.4.1 Delays in Approval Processes

Delays in the approval process impacted the lead-in time for the evaluation team to liaise with schools and increase engagement, even with support from the Raise Foundation in disseminating information and Expression of Interest processes. While the process of seeking approval from a Human Research Ethics Committee and from the relevant state education department can naturally be time-consuming (Thomas,

2009), these timelines were significantly affected by ongoing COVID-19 impacts. The Victorian Department of Education's Research in Schools and Early Childhood (RISEC) was suspended during the COVID-19 pandemic and was still suspended in January 2023, requiring an exemption process for approval. The Western Australia Department of Education flagged delays in their research approval process due to reduced staff capacity also resulting from COVID-19 lockdowns. After providing feedback required to progress the application, the Western Australian Department of Education only contacted the evaluation team in November 2023 – 10 months after the initial application.

The evaluation team submitted applications addressing all feedback and revisions as promptly as possible and, in collaboration with the Raise Foundation, leveraged the Department of Health funding and existing relationships with state education departments to ensure that approvals were received as soon as was possible.

#### 3.5 Challenges in Evaluation Recruitment

The challenges of engaging adolescents in school-based research, as experienced in this evaluation, have been well documented in published literature through both research and experience (Bonnell et al., 2018; Trimmer et al., 2020; Hatch et al., 2023; Thomas, 2009). Such challenges include difficulties accessing schools, low staff buy-in and low student survey response rates, which result from a consistent lack of capacity in schools, often complex ethics approval processes or procedures, and approvals and engagement from gatekeepers. These challenges with school-based research engagement have only been amplified since the COVID-19 pandemic (Waechter et al., 2023) likely due to increased school burdens, often reduced school capacity and delays in gatekeeper approvals. The Raise Foundation also reported difficulties engaging with their school key contacts and surveying mentees in 2023, both more broadly and with some of the specific evaluation schools.

Eligible Raise schools in this evaluation were all public schools which can be particularly difficult to access, as researchers are required to go through multiple gatekeepers before approaching students. These gatekeepers can include the Human Research Ethics Committee, relevant state education departments, school principals and other staff (Bonnell et al., 2018). A high level of necessary yet difficult gatekeeping was noticeable in the evaluation, such as reports from some schools that they were passing the online survey to their Legal or Information Technology branches for approval before allowing students to complete it. This was despite the survey having received state education department approval and the school having already consented to participate. This resulted in one school dropping out of the evaluation, as they could not seek the relevant advice within the provided timeframe.

The response rate of schools consenting to participate in the evaluation once contacted was also low, despite frequent and varying communications from the evaluation team. This may be attributable to low capacity in schools, where teachers have continued to feel stressed and burnt out since COVID-19 (Kotowski et al., 2022). Even when schools consented to participate, some changed their mind when they received the survey or completely disengaged from contact, which often happens in research — an experience detailed by Australian researcher Nicholas Flegg (Trimmer et al., 2020). Some schools that did remain consented and engaged in the evaluation found fulfilling their participation commitment difficult, another regular occurrence in research due to time and resource constraints and competing priorities (Hatch et al., 2023). To address these limitations, it has been recommended that survey-based research methods should align with school priorities and be led by schools and tailored to their timetable and resources (Hatch et al., 2023). The objectives of this research were naturally aligned to the priorities of eligible schools, given they had already been implementing the Raise Program. The evaluation's methodology also followed these recommendations by allowing schools to flexibly distribute surveys in a way and at a time that best suited them.

Finally, there were some challenges with student response rates, even once schools had consented and were engaged. Required opt-in consent procedures used in the evaluation are known to yield much lower

response rates than those using passive or implied consent (Thomas, 2009). Furthermore, it was reported by some school staff that adolescents in the control condition did not see the value of committing their time to surveys due to not being involved in the Raise Program. This is where school engagement was particularly crucial, as the most engaged and keen evaluation schools had the highest consent rates. While the evaluation team attempted to minimise these difficulties and maximise participation by advising on multiple occasions that schools provide time and encouragement to students in a wellbeing or homegroup period to complete the survey, final consent numbers from some schools indicated that this was likely not provided. Similarly, there were difficulties engaging Raise mentees to consent in focus groups or interviews where access to the students was only possible via student-supplied school email, as per ethics and education department approvals (see 3.3). As a contingency plan, qualitative questions about mentee experiences of the Raise Program were incorporated into the follow-up survey. Intended interview questions were provided in an additional optional online REDCap survey.

#### 3.6 Data Collection

#### 3.5.1 Cohort Surveys

Surveys were provided to consenting students pre (baseline) and post (follow-up) the Raise Mentoring Program, using the measures outlined in 3.2. See Appendix C for the cohort surveys.

#### **Baseline Survey**

The baseline survey was conducted from May to June 2023, to measure student outcomes prior to commencement of the Raise program.

Schools were provided with a written piece to place in their newsletter to publicise the study to parents and a poster to publicise the study to students. Surveys were built into the University of Melbourne's REDCap, a secure web platform for building and managing online databases and surveys. Schools were provided with a general online survey link and QR code to the online survey, to distribute to all students in the cohort. This led students through the study's Plain Language Statement before obtaining their consent and providing them with the opportunity to complete the baseline survey. It was recommended that schools provide a wellbeing or homegroup period time for students to complete the survey. Schools that provided this time and had leadership who saw the value of having comparison cohort data had the highest survey response rates. Some key contacts at schools raised difficulties with providing this time and engaging students in a non-compulsory survey, particularly those who were not selected for the Raise Program and had no vested interest. Ongoing contact was made with schools who communicated such difficulties to assist with problem-solving and to increase their understanding of the evaluation and its objectives to promote engagement, including reminders of the cohort-level wellbeing report incentive.

Students were required to self-identify as mentees who had been selected for the Raise Mentoring Program in 2023. This was due to the restriction of student information collection prior to the initial survey. Students who indicated that they were participating as mentees were not required to re-answer questions they had already been asked in Raise's own internal surveys.

As required by certain states, students were identified as distressed if they answered 'Yes' to having experienced any difficult home or personal events and answered 'Most of the time' or 'All of the time' to the follow up question as to the degree that these issues have impacted their ability to engage in school activities. These students were either contacted and provided with contact information for relevant support services or re-identified to the school principal for follow-up.

#### **Follow-Up Survey**

The follow-up survey was conducted from October to November 2023, to measure outcomes after conclusion of the Raise program.

Students who had consented to and completed the baseline survey and provided their email were emailed a unique link to their follow-up survey. Students received two reminder emails in the subsequent two weeks if they had not completed their survey. Students who had not provided their email were mailed a letter via the school with a unique QR code and access code to their follow-up survey. Schools were again provided with a general survey link and QR code to facilitate participation of any additional students who did not participate in the initial survey, or those who had difficulty accessing their individual links. Schools also received reminder emails and it was again recommended that they encourage students to complete the survey during a dedicated wellbeing or homegroup period time. The same process was followed as in the baseline survey for supporting students identified as distressed. At the conclusion of this data collection to thank schools for their participation, schools were mailed cohort-level reports of student wellbeing based on the EPOCH measure.

#### 3.5.2 Pulse Surveys

Pulse surveys (see Appendix D) were conducted to track mentee outcomes and engagement with the program to determine impact on outcomes. These pulse surveys were mailed to schools in hard-copy format individually addressed to mentees. This was to simplify distribution by Program Counsellors, who were engaged by Raise to facilitate this data collection during program time. Data was manually entered into REDCap when surveys were returned via pre-paid mail satchels.

Pulse surveys were initially intended to be conducted monthly. However, concerns were raised by school key contacts and Raise Program Counsellors during the first two pulse surveys about feedback received from both mentees and mentors about this frequency and the challenges of motivating mentees to complete multiple surveys. It was therefore decided that only one further pulse survey would be distributed, to focus on where mentee outcomes were tracking at the midpoint of the program. This resulted in a maximum of three pulse surveys.

The pulse surveys collected demographic information from mentees including their date of birth, year level and gender. Mentees were asked how many Raise sessions they had attended in the past four weeks, and whether they had to match with a different mentor during individual mentoring time. At the third midpoint survey, only mentees were asked to answer whether they had experienced any difficult home or personal events (Yes/No) and if so, the impact of these on their ability to engage in school activities (None of the time/A little of the time/Some of the time/Most of the time/All of the time), as per the cohort surveys.

Mentees were then asked to indicate on a 5-point Likert scale (from 'not at all' to 'always') how much they agreed with the following statements about their relationship with their mentor:

- I feel safe with my mentor
- My mentor helps me to share my problems
- My mentor helps me come up with ideas to my problems

Additionally, mentees were asked to indicate on the same 5-point scale how much they agreed with the following statements related to the outcomes of the evaluation in the previous month:

- I am finding the Raise program useful for me
- I feel that I really belong in my school
- I can bounce back quickly after hard times
- I feel hopeful about the future
- I am able to seek help from others when I need

Schools were mailed physical copies of pulse surveys individually addressed to each Raise mentee. Nine schools facilitated completion of and returned at least one pulse survey via post.

#### 3.5.3 Mentee Perspectives

Mentees were asked open-ended questions about their experiences in the Raise program at the end of the follow-up survey (see Appendices B and E). These questions asked about benefits and negatives of participating in Raise, as well as the mentee's engagement with their mentor.

Mentees were also asked on the follow-up survey whether they were interested in hearing more about opportunities to further provide their experience in participating in the Raise program. Seven mentees indicated their interest and were offered the opportunity to participate in either a focus group or interview. These mentees were not responsive to contact or follow-up emails and were instead offered the opportunity to answer the questions via a second online survey (see Appendix G). These questions focused on the enablers and barriers for mentees to program participation and any changes they had seen in themselves. Two mentees consented to participating and completed the additional survey.

#### 3.5.4 Raise Administrative Data

The evaluation measures informed the measures used in Raise's survey design for 2023, with mutual measures being the Raise-developed help-seeking questions, the perseverance, optimism, and connectedness scales from the EPOCH, and the BRS. This provided the opportunity to utilize Raise datasets. Raise therefore provided demographic information and all survey data for mentees and mentors at consenting evaluation schools to supplement the evaluation dataset.

It is important to note that these responses were not collected for research purposes as they are part of Raise's standard evaluation data collection for their own internal use, which may involve sharing with key stakeholders. Raise's data collection process was different to the evaluation's, with mentees provided with Raise Program time to complete internal surveys in the presence of the Raise mentors, as opposed to the opt-in process during class or a student's own time for the evaluation surveys. Therefore, it is likely that the two different approaches to data collection may be influenced by different types of bias (e.g. selection bias for the evaluation cohort, response bias for the Raise collected data).

It is also important to note there was a difference in how the three help-seeking questions were asked. Specifically, the response option for Raise involved a 5-point scale (Almost Never, Sometimes, Often, Very Often, Almost Always), while the evaluation collected data on a 4-point scale, as outlined (see 3.2.2). The evaluation was decided in consultation with Raise where their internal data would be transposed as follows:

- Raise's Almost Never (1) to the Evaluation's Rarely (2)
- Raise's Sometimes (2) and Often (3) to the Evaluation's Sometimes (3)
- Raise's Very Often (4) and Almost Always (5) to the Evaluation's Always (4)

Given this discrepancy, responses to these three questions were collected from mentees at follow-up on the 4-point scale through the evaluation's data collection process, in addition to Raise's internal data collection.

#### 3.6 Data Management

#### 3.6.1 Missing Data

Missing data is common in research and evaluation projects, particularly when reliant on self-report and when time to complete data collection activities is not explicitly provided or supported. Data may be absent due to three overarching factors:

- 1) Missing completely at random (MCAR) is the most stringent assumption, suggesting that the absence of a data point is entirely unrelated to both observed and unobserved data.
- 2) Missing at random (MAR) is a more realistic assumption than MCAR, indicating that the reason for a data point being missing can be explained by the observed data.
- 3) Missing not at random (MNAR) implies that the missingness is influenced by the unobserved values.

When encountering missing data, it is crucial to refrain from excluding instances that lack information (analyses conducted after such exclusion are referred to as complete case analyses). Multiple imputation is a method that replaces missing values with plausible numbers. These numbers are generated based on the distributions and correlations of the observable variables in the dataset.

The several imputation method is distinct from single imputation methods as it involves filling in missing data several times, with numerous feasible values calculated for each missing value. Employing several plausible values enables a precise assessment of the uncertainty associated with predicting the potential missing values, hence preventing the generation of misleading accuracy (which can occur with single imputation). Multiple imputation is a statistical technique that yields precise estimates of quantities or relationships of interest. These include treatment effects in randomised trials, sample averages of specific variables, correlations between two variables, and their respective variances. By doing this, it decreases the likelihood of reaching incorrect positive or negative conclusions.

For this dataset, multiple imputation involved the generation of 50 imputed datasets. These datasets each uniquely predicted the value of the missing variables, with analyses taking into account the mean of these datasets. Specifically, multiple imputation was focused on imputed outcome scores, based on variations in baseline scores for each individual variable.

#### 3.6.2 Establishing the Matched Comparison Group

A comparison analysis was conducted between students who receive the Raise intervention and a matched cohort. Propensity score matching (PSM) is a statistical technique used to evaluate the causal effect of a certain programme by comparing outcomes between individuals who received the program and those who did not, while accounting for potential confounding variables.

PSM mitigates the selection bias that could exist in non-experimental data. Selection bias occurs when units (such as individuals, communities, or schools) are not randomly assigned to a certain programme, and the units that choose or are able to participate differ systematically from those that do not. A propensity score is a calculated likelihood that a unit would be subjected to the programme, based on the unit's observed characteristics. The propensity scores of all units in the sample, including both beneficiaries and non-beneficiaries, are utilised to form a comparison group for measuring the program's impact. Raise mentees were matched against students in the same cohort who did not/have not participated in the Raise program and were matched on similar baseline demographic information (i.e., age, gender, etc.).

Matching occurred between Raise participants and non-Raise participants at participating evaluation schools. To identify matching variables, univariate analyses were conducted of baseline demographic variables to predict variables related to Raise participation. Matching was eventually conducted with 1 nearest, with a caliper of 0.003. The identified matching variables were student experience of a difficult life event, student gender, and student disability, as well as EPOCH subscores of perseverance, optimism and connectedness, and help-seeking behaviour. The variance after matching is present below.

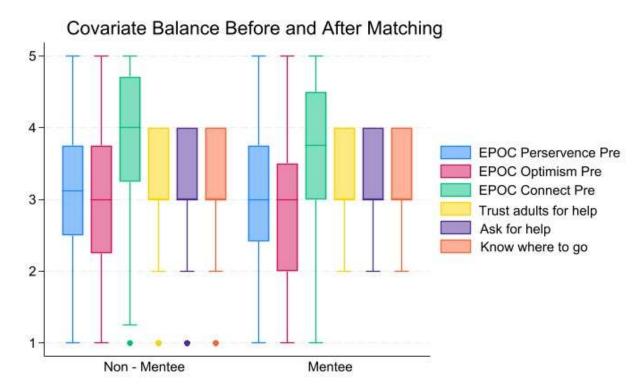


Figure 2: Covariate after matching.

Overall, 148 Raise mentees were matched with 276 non-Raise students based on demographic variables. However, as matching is based on baseline characteristics, the final sample size for each outcome data at follow up is presented in the following table.

Table 4. Sample size of each outcome after matching

	Mentee Data	Non-mentee Data
Know where to go	100	99
Ask for help	101	98
Trust in adult	101	99
EPOCH – Engagement	17	95
EPOCH – Perseverance	102	95
EPOCH – Optimism	95	83
EPOCH - Connectedness	95	83
SSE School Aspiration	17	101
School Belonging Scale	95	88
<b>Brief Resilience Scale</b>	94	83

#### 3.6.1 Defining Students who were 'Low' at Baseline

For each outcome, students who were defined as 'low' were established to examine whether these students had a differential change in their outcomes. Established risk categories and cut points for measures were used to define students as 'low', when these were available.

For measures without established cut points, the bottom third of scores for all participants who completed the baseline survey were defined as 'low'. It was decided to establish these cut points with all participants, as only using the matched cohort would less likely reflect an at-risk group of the general student population.

#### 3.7 Minimum Detectable Effect Size for the Sample Size

Given the lower-than-expected recruitment and data collection, the final sample size varied depending on outcome collected.

Given this, we conducted power calculations before the analyses to determine the minimum detectable effect size of the sample.

- For EPOCH School Engagement, which had the lowest sample size, there were 112 students overall with outcome data. Assuming a power of 0.80% and a statistically significant alpha of 5%, our final sample size had a **minimum detectable effect size of 0.28**. Therefore, it is more likely findings below this effect will not be statistically significant. However, as described in Section 3.9 below, an effect size of 0.27 could be defined as a large effect size which has significant, long-term implications for student psychological outcomes.
- For outcomes such as EPOC Connectedness, which included data collected by Raise to supplement the evaluation data, the sample size was higher at 199. For these outcomes, based on the assumption of 80% power and statistically significant alpha of 5%, the sample size enabled a minimum detectable effect size of 0.20. Based on Section 3.9 below, this effect size would meet the criteria for a medium effect size for student psychological outcomes.

#### 3.8 Interpretation of Effect Sizes

In research aiming to understand the outcomes of an intervention, a traditional approach is to focus on whether outcomes are statistically significant or not. However, increasing attention has been paid to the magnitude or size of the difference as opposed to the significance of effects. Effect size refers to the magnitude of the relation between the independent and dependent variables and is separable from statistical significance, as a highly significant finding could correspond to a small effect, and vice versa, depending on the study's sample size. For intervention research, Cohen's d is often used, with benchmarks indicating an effect size up to 0.2 as small, up to 0.5 as medium and over 0.8 as a large effect size difference.

However, more recently it has been recognised that the magnitude of difference between groups that can be expected should be aligned to the intervention's theory of change, but also to the specific domain which is being examined. More recently, Funder & Ozer (2019) found that for intervention research in psychological domains, the following benchmarks should be used:

- 0.05 indicates an effect that is very small for the explanation of single events, but potentially consequential in the not-very-long run.
- 0.10 indicates an effect that is still small at the level of single events, but potentially more ultimately consequential.
- 0.20 indicates a medium effect that is of some explanatory and practical use, even in the short run, and is therefore even more important,
- 0.30 indicates a large effect that is potentially powerful in both the short and the long run.
- 0.40 or greater, in the context of psychological research, is likely to be a gross overestimate that will rarely be found in a large sample or in a replication.

Therefore, our interpretation of the outcomes of the Raise Mentoring Program are based on these guidelines, given the psychological outcomes being measured and the type of intervention. In addition to statistical

significance, we will also examine the overall pattern of difference between the groups, reflecting the limitation of the small sample size in detecting statistically significant findings.

In the findings, these benchmarks are presented to help with interpretation, with red indicating a 0 effect size, orange an effect size of 0.2 and green representing an effect size of 0.4.

#### 3.9 Considerations for Testing of Multiple Outcomes

Examining various outcomes in a study endeavour presents a distinct array of obstacles that researchers must meticulously tackle to uphold the integrity and dependability of their results. An important concern arises from the phenomenon of alpha inflation, in which the likelihood of committing a Type I mistake escalates with the testing of each successive result. This phenomenon of inflation arises due to the fact that conventional standards of statistical significance (e.g., p < 0.05) are predicated on the premise of a solitary test.

However, when several outcomes are examined concurrently, there is an increased probably that positive findings will be identified by random chance. Therefore, there is an increased likelihood of identifying inaccurate positive results or relationships that are not genuine, potentially resulting in incorrect conclusions about an intervention's benefits. Another problem pertains to the possibility that some outcomes will interact with others, or that there is overlap in the domain which is being measured.

Therefore, a common approach to examine multiple outcomes is to consider the patterns across multiple outcomes, patterns between measures which capture similar domains in addition to the final p-value and confidence intervals.

Although the evaluation was guided by an established theory of change and program logic, it is also likely that not all outcomes will demonstrate statistical significance at the same timepoint and some changes in outcomes may be predicted by changes in other outcomes which occur earlier. For instance, students may need to have improved connectedness before changes in school engagement occur. More importantly, examination of the outcomes should consider the pattern of findings and whether outcomes in similar domains are showing findings in the same direction.

#### 3.10 Data Analysis

All analyses were conducted in STATA v17, with all output provided as Supplementary files 01 to 07.

The baseline characteristics of the participants and schools are summarised by group. Categorical variables are presented as frequency and proportion values in each category. Continuous variables are presented by means and standard deviations (SDs) for unskewed data, medians and IQRs for skewed data, and ranges.

In these analyses, the primary analysis was by intention to treat and included all matched participants where outcome data were available. The analysis used a multivariate linear regression to examine continuous outcomes, whilst logistic regression was used for categorical data. The analysis output is presented as part of Appendix H.

#### 3.10.1 To Examine Differences in Change Scores

Outcomes for the intervention students were compared with students in the control group. Both unadjusted and adjusted analyses were conducted. For adjusted analyses, two models were conducted. The first was based only on change scores, whilst the second also included student gender and family difficult event. Clustering of students within schools was accounted for in the models using regression techniques that respect these structures. Findings between groups were presented as mean differences with 95% CIs, p values and Cohen's d effect sizes.

### 3.10.2 To Examine Differences at Follow Up

Both unadjusted and adjusted analyses were conducted. For adjusted analyses, two models were conducted. The same approach to above was used, except model one accounted for baseline scores. Clustering of students within schools was accounted for in the models using regression techniques that respect these structures. Findings between groups were presented as mean differences with 95% CIs, p values and Hedge's g effect sizes.

# 4. Findings

#### 4.1 Demographic Information

The overall flow of participants is presented in the participant flow chart below.

At baseline, 418 students consented to participate and at least partially completed the survey, with 41 enrolled in Raise. These responses were across nine schools, with the remaining four schools either nonresponsive or communicating difficulties with survey facilitation or student engagement. At follow-up, 365 students completed the survey, with 28 enrolled in Raise. This was comprised of 137 students who had completed the baseline survey, representing a follow-up rate of approximately 32.78%. An additional 311 new students completing the follow-up survey (10 Raise mentees and 211 non- mentees). These responses were across eight schools, with the remaining five schools nonresponsive to multiple points of email and phone contact. This meant that overall, data from 751 unique students was available. This was supplemented with Raise administrative data for 124 additional mentees. After matching on baseline variables, there were 148 Raise mentees and 276 matched comparison students.

It is important to note that Raise administrative data provided a response option of 'Rather not say' when asking if mentees had experienced a difficult event. A conservative approach was taken for analysis, where students who selected this response option were categorised as having answered 'No/Rather not say'.

Tables 3 and 4 contain demographic information for the whole recruited cohort and matched cohort respectively. The *P* value shows the difference between mentees and non-mentees at baseline.

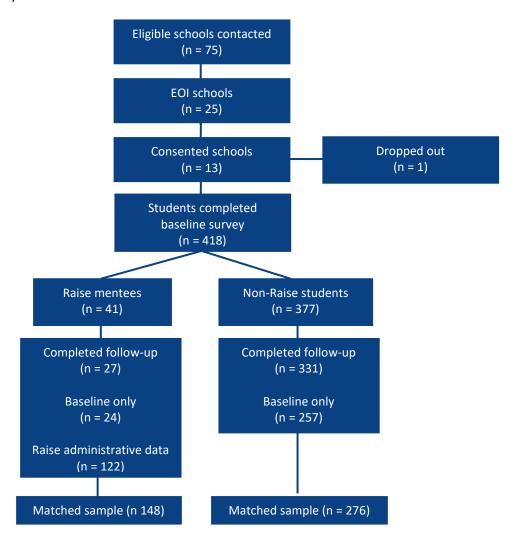


Table 5. Recruited cohort demographic information

	Mentee		Non-Mentee	Non-Mentee	
	<u>N</u>	<u>%</u>	<u>N</u>	<u>%</u>	<u>P-value</u>
Year Level Year 7 Year 8 Year 9 Year 10	11 95 45 8	6.92% 59.75% 28.30% 5.03%	2 395 199 0	0.34% 66.28% 33.39% 0.00%	0.000
Gender Male Female Other Rather not say	66 81 8 3	41.77% 51.27% 5.06% 1.90%	305 245 15 23	51.87% 41.67% 2.55% 3.91%	0.027
Born in Australia Yes No	122 36	77.22% 22.78%	460 126	78.50% 21.50%	0.729
Speak a language other than E	inglish at home				0.362
Yes, mostly Yes, sometimes No	39 45 74	24.68% 28.48% 46.84%	143 139 310	24.16% 23.48% 52.36%	
Has a disability Yes No Rather not say	9 138 7	5.84% 89.61% 4.55%	32 518 37	5.45% 88.25% 6.30%	0.707
Lives with One parent Both parents Carer or guardian Other	52 91 4 10	33.12% 57.96% 2.55% 6.37%	92 463 9 28	15.54% 78.21% 1.52% 4.73%	0.000
Carer Yes No Rather not say	49 99 6	31.82% 64.29% 3.90%	70 263 37	18.92% 71.08% 10.00%	0.001
Experienced a difficult event Yes No	82 73	52.90% 47.10%	110 263	29.49% 70.51%	0.000
Degree of impact of the difficult event 0.					
None of the time A little of the time Some of the time Most of the time All of the time	9 16 22 24 9	11.25% 20.00% 27.50% 30.00% 11.25%	11 36 39 17 6	10.09% 33.03% 35.78% 15.60% 5.50%	

Table 6. Matched cohort demographic information

Mentee	Non-Mentee

	N = 148		N = 276			
	<u>N</u>	<u>%</u>	<u>N</u>	<u>%</u>	<u>P-value</u>	
Year Level Year 7 Year 8 Year 9	11 95 43	7.38% 63.76% 28.86%	1 161 114	0.36% 58.33% 41.30%	0.000	
Gender Male Female Other Rather not say	64 75 8 2	42.95% 50.34% 5.37% 1.34%	148 108 10 8	54.01% 39.42% 3.65% 2.92%	0.082	
Born in Australia Yes No	116 33	77.85% 22.15%	221 53	80.66% 19.34%	0.494	
Speak a language other than English at home						
Yes, mostly Yes, sometimes No	35 45 69	23.49% 30.20% 46.31%	48 52 176	17.39% 18.84% 63.77%		
Has a disability Yes No Rather not say	9 134 6	6.04% 89.93% 4.03%	16 246 14	5.80% 89.13% 5.07%	0.886	
Lives with One parent Both parents Carer or guardian Other	49 85 4 10	33.11% 57.43% 2.70% 6.76%	42 219 1 14	15.22% 79.35% 0.36% 5.07%	0.000	
Carer Yes No Rather not say	48 96 5	32.21% 64.43% 3.36%	48 198 28	17.52% 72.26% 10.22%	0.000	
Experienced a difficult event Yes No	78 71	52.35% 47.65%	78 198	28.26% 71.74%	0.000	
Degree of impact of the difficult event						
None of the time A little of the time Some of the time Most of the time All of the time	9 16 21 24 7	11.69% 20.78% 27.27% 31.17% 9.09%	8 28 27 10 4	10.39% 36.36% 35.06% 12.99% 5.19%		

#### 4.2 Changes in Student Outcomes

#### 4.2.1 Help Seeking Behaviour

### Help Seeking Behaviour

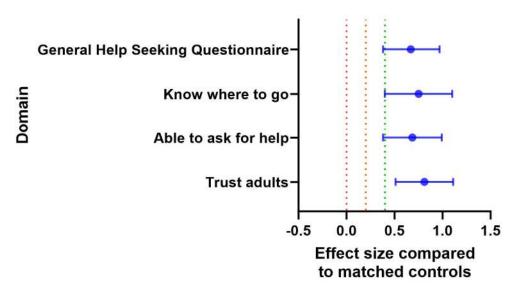


Figure 3. Effect size differences for help-seeking behaviour.

When examining help-seeking behaviour, there were consistent findings that Raise students had a statistically significant greater magnitude of change between the pre and post intervention time points. This ranged from a mean difference of 0.69 (95% CI 0.38-0.99) for being confident to ask for help when needed, to a mean difference of 0.81 (0.51-1.11) for being able to trust adults for help. Importantly, the findings were also confirmed by the validated General Health Seeking Questionnaire, which had an effect size mean difference of 0.67 (95% CI 0.38-0.97) between the two groups. These differences attenuated when accounting for difficult life event and student gender but remained statistically significant.

#### 4.2.2 Changes in Student Social-Emotional Wellbeing

# **EPOCH Wellbeing Measure**

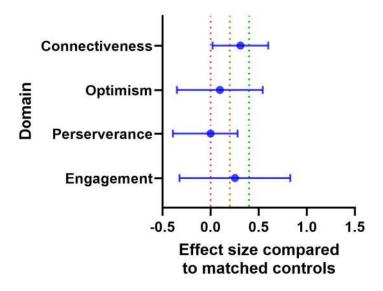


Figure 4. Effect size differences for the EPOCH measure.

Changes in student wellbeing were measured using the EPOCH, specifically the engagement, perseverance, optimism, and connectedness domains. In unadjusted analyses, there was a statistically significant difference between the Raise and matched cohort participants, with a mean difference effect size of 0.31 (95% CI 0.02 to 0.60). This finding provides evidence that Raise mentees reported having greater improvements in their connectedness, which EPOCH defines as feeling of supported and being valued by others. Although there were suggestions that Raise mentees have a greater magnitude of change in terms of optimism (effect size = 0.10, 95% CI -0.35 to 0.54) and engagement (effect size 0.25, 95% CI -0.32 to 0.83), these were not statistically significant.

#### 4.2.3 Changes in Student School Engagement

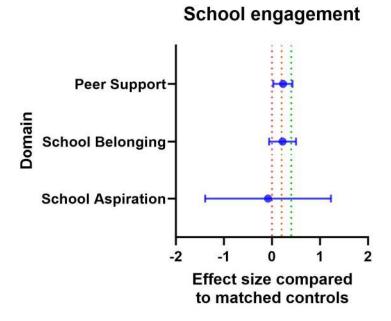


Figure 5. Effect size differences for school engagement.

Changes in student school engagement were measured using three different measures that captured school aspiration, belonging, and peer learning support. For the unadjusted analyses, there was only evidence for a statistically significant difference between Raise mentees and matched cohort students for peer support for learning. Raise mentees had a greater magnitude of positive change, with an effect size of 0.23 (95% CI -0.0.2 to 0.0.43) when compared to the matched cohort. Although school belonging had an effect size difference of 0.22, this was not statistically significant.

In addition, school aspiration, which is defined by how students feel about school, had minimal difference between the Raise and matched cohorts. Similar to above, it is important to note the smaller sample size for the school aspiration measure, with data only available for 17 mentees and 101 matched comparison students.

Overall, the findings suggest that there is evidence for Raise mentees reporting a greater change in peer learning support compared to their matched peers, with promising findings for school belonging.

#### 4.2.4 Student Resilience

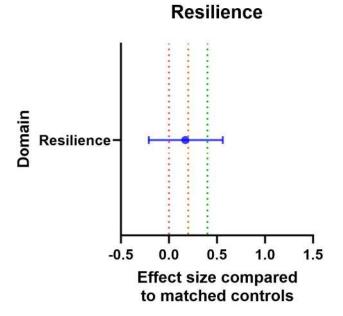


Figure 6. Effect size differences for student resilience.

The final domain examined was student resilience. Raise mentees had a greater magnitude of change when compared to the matched comparison (effect size = 0.17, 95% -0.20 to 0.56), but this difference was not statistically significant. This analysis was based on a cohort of 94 Raise mentees and 83 matched comparison students.

#### 4.3 Differential Outcomes Based on Student Baseline Demographics

Additional analyses were conducted to examine whether baseline student demographics were associated with a differential effect on the observed outcomes. This analysis explored baseline demographics of child gender and difficult life event, which was determined via the consultation with the Evaluation Advisory Group.

As the analyses involved examining the interaction of the baseline variables and Raise allocation, this meant the sample was further divided along both baseline demographic characteristics (see Table 6). Despite the sample size, there was evidence for a statistically significant interaction for some of the outcomes.

For the General Help Seeking Questionnaire, there was evidence that female Raise students had a greater magnitude of benefit compared to those who did not identify as female. The comparison to females was used, as during consultation with Raise, it was signalled that female students may have different outcomes compared to other students. Although not statistically significant, this pattern of having at least an effect size greater than 0.2 difference was also seen for engagement and school aspiration. Inversely, those who did not identify as female had a pattern of greater improvement for optimism, school belonging, and resilience.

When examining students who had a difficult life event at baseline, there was statistically significant evidence that these students had a lower magnitude of change in terms of connectedness and school belonging. A similar pattern was also seen for general help-seeking and optimism. Although not statistically significant, these Raise students had a large magnitude of difference in engagement compared to Raise mentees who did not have a difficult life event at baseline.

Table 7. Differential effects for gender (female) and difficult life event.

	Female Mean diff (95% CI)	Difficult life event Mean diff (95% CI)
General Help-Seeking Questionnaire	0.34** (0.10 to 0.59)	-0.32 (-0.54 to -0.09)
Know where to go	-0.03 (-0.59 to 0.53)	-0.06 (-0.31 to 0.18)
Ask for help	0.00 (-0.69 to 0.70)	-0.16 (-0.55 to 0.24)
Trust in adult	-0.03 (-0.59 to 0.52)	-0.04 (-0.37 to 0.28)
EPOCH – Engagement	0.60 (-0.30 to 2.50)	0.75 (-0.24 to 1.75)
EPOCH – Perseverance	-0.10 (-0.56 to 0.38)	-0.17 (-0.73 to 0.38)
EPOCH – Optimism	-0.45 (-0.93 to 0.02)	-0.41(-0.98 to 0.14)
<b>EPOCH - Connectedness</b>	0.08 (-0.38 to 0.55)	-0.48 ** (-0.90 to -0.07)
SSE School Aspiration	0.96 (-1.99 to 3.91)	-0.04 (-2.10 to 2.01)
School Belonging Scale	-0.30 (-0.80 to 0.20)	-0.63** (-1.18 to07)
Brief Resilience Scale  ** Statistically significant < 0.05	-0.31 (-0.90 to 0.32)	-0.08 (-0.65 to 0.49)

<sup>\*\*</sup> Statistically significant < 0.05

#### 4.4 Proportion of Students Showing Improved Scores

In exploratory analyses, we examined the proportion of Raise and non-Raise students whose outcomes had improved above an effect size of 0.1 between the two time points. Research has shown defining a change score as simply any improvement above 0 can lead to overestimation of benefits, as scores can increase by more than 0 (e.g. 0.05) simply due to the standardisation approach and range of schools. A similar limitation occurs when using the medium or mean score to define positive change. Therefore, the cut point of 0.1 was used to ensure that any improvements shown represent a substantial change, as well as accounting for potential influence of regression to the mean due to repeated measurements.

The differences in proportions between the Raise and non-Raise students are presented in Figure 6. Help-seeking items are not presented as these were measured via single Likert scales, which is not suitable for these types of analyses.

Similarly to above, although there appears to be a pattern that Raise students were more likely to have a score change of at least 0.1, only student connectedness had a statistically significant finding. However, Raise mentees had positive increased odds for improving scores, ranging from OR 0.95 for student mental wellbeing, to OR 2.65 for student connectedness.

## Proportion of students reporting an improved score

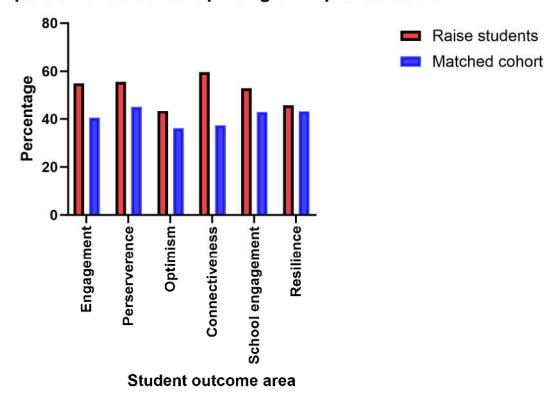


Figure 7. Proportion of students with improved scores.

Table 8. Odds ratio for mentee having an improved scores compared to matched non-mentees.

	Odds Ratio	95% CI	P-value
Engagement	1.67	0.86, 3.26	0.13
Perseverance	1.27	0.71, 2.25	0.40
Optimism	1.35	0.74, 2.46	0.31
Connectedness	2.62	1.38, 4.97	0.003
School engagement	1.39	0.49, 3.90	0.53

Resilience	1.83	0.97, 3.45	0.06
Mental well-being	0.95	0.32, 2.80	0.93

### 4.5 Changes in Student Outcomes for Those at Risk at Baseline

A previous internal evaluation conducted by Raise in 2020-2022 found that students who were at-risk in each outcome domain at program commencement had a statistically significant increase in their outcome scores at follow up. We aimed to replicate these findings for our cohort. A main different between the analyses presented here is that instead of using the mean as the cut point to define low and high student baseline outcomes, we have used the bottom third of all participants to define the low group. In examination of the differences in outcomes for those defined as low or not low at baseline, findings are presented in Figure 7 below. Furthermore, only the outcomes which were collected by both the evaluation and Raise are presented, due to the other outcomes having low sample sizes (N < 30).

Overall, the effect size mean difference suggests that Raise mentees had better change scores across all measures when compared to their matched comparisons. However, the differences were not statistically significant.

#### Outcomes for those low at baseline

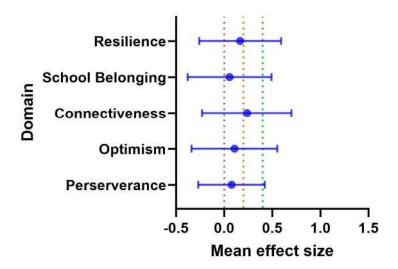


Figure 7. Differences for those low at baseline.

#### 4.6 Student Perceived Enablers and Barriers Related to Outcomes

Raise mentees were asked about the benefits and negatives of participating in the program as part of the follow-up survey. Twenty-seven mentees answered these questions (see Appendix F) and survey responses were thematically coded.

One of the key benefits identified by mentees was the opportunity to talk to someone about their issues, problems, and shared interests (n = 15). From these responses, trust, feeling listened to and having a mentor that was not a medical professional were seen as enablers to mentees' ability to talk to their mentor.

"I was able to trust my mentor and felt like I was able to improve some learning skills and had someone to talk to about my problems" (Mentee Survey Respondent).

All mentees, except one who didn't know, indicated that they had positive engagement with their mentor, which may have also supported their ability to talk about their issues and problems. Self-reflection and goal setting was seen by three respondents as the key benefit of the program.

"help you set goals for the future" (Mentee Survey Respondent).

Other benefits included improved confidence, mental health and learning skills, as well as not going to class. Providing free food was identified as an enabler to participation.

When asked about the negatives of the program, 10 mentees identified feedback to be considered for improvement. The largest portion of these responses (three mentees) thought that the Raise Program was too short. One mentee found the initial settling in period to be difficult and another mentee found talking to a stranger difficult. Other negatives identified included missing out on core school subjects and the games at the start of the period not being enjoyable.

Two further mentees completed the intended focus group/interview questions via an additional online survey (see Appendix G). Both mentees spoke frequently about their confidence as being both a key takeaway from the program and a change they had noticed in themselves through participating in the Raise program. These mentees described themselves as engaged and noted that they had tried all activities provided. Both again reiterated that the best part of the program was getting to speak with their mentor. One mentee noted that they would not change anything about the program as it was all "fun and useful". The other mentee would change the duration of the program so that it would run until the end of the year.

## 5. Discussion and Recommendations

#### 5.1 Summary of Findings

#### **Key Findings**

- Compared to similar students, those who participated in Raise had better help-seeking behaviour, connectedness, and peer support immediately at the end of the program.
- There were promising findings in relation to social-emotional wellbeing and school engagement.
- Student gender and difficult life event influenced the outcomes for Raise students.

The findings were presented to the Raise Foundation and Evaluation Advisory Group through a series of meetings to conduct sense-making activities. These aimed to determine whether there was anything unusual about the findings that required further verification, as well as understanding how findings might be similar or different to previous Raise evaluations. However, the final conclusions were based on the interpretation of the University of Melbourne Independent Evaluation team.

Overall, the findings from this evaluation provide promising, independent evidence that students who receive the Raise Mentoring Program have better outcomes at the end of the program when compared to their matched peers. Specifically, the statistically significant findings were for validated measures capturing help-seeking behaviour, students being valued and supported, and as having peer support for their learning.

Although the other outcomes were not statistically significant, evaluation of the effect size patterns found that all outcomes, except for student school aspiration and perseverance, trended towards positive outcomes for the Raise mentees compared to matched comparisons. Furthermore, similar positive patterns were demonstrated for those students who were low in each functioning domain at baseline. However, as described in section 3.9, the testing of multiple outcomes requires caution against focusing on the individual statistically significant findings due to the increased likelihood of false positive and negative findings being identified by chance. Despite this, there is a promising pattern of findings trending towards positive outcomes for Raise mentees compared to matched comparison group. This is particularly important given the guidelines by Funder & Ozer (2019) which found effect sizes between 0.2 to 0.4 to have long-term benefits for psychological based outcomes for interventions. However, the final sample size and wide confidence intervals do not fully confirm these outcomes. Therefore, replication of these findings is required to add further strength to the evidence of these benefits.

The findings from this evaluation build on the significant internal evaluations conducted by Raise over the past 15 years. Importantly, this is the first quantitative evaluation to be conducted at a national level by an independent organisation outside of Raise, and the first to use a comparison group to compare outcomes. As noted in the rapid review, there have been few school-based mentoring programs internationally which have aimed to evaluate findings using a comparison group.

The following high-level recommendations for future data collection have emerged from the conduct of the evaluation for the interim report:

#### 5.2 Recommendations

Recommendation 1: Refinement of the mentee selection criteria, with a focus on identifying students who may experience the greatest benefit from the program

The findings suggest that students who are low in functional domains at recruitment may experience greater improvements and benefits through the Raise Mentoring Program. Therefore, it is recommended that Raise consider including data on student functioning as part of the mentee selection process, which

would enable schools to prioritise which students participate in the program. This would be particularly helpful for schools in which the number of potential Raise mentees is greater than the number of potential mentors available to support students. However, it is important to caution that the use of functional data should be used in combination with other key factors which align with Raise's intervention intentions. Given that this approach does risk missing young people who may benefit from the program but are high functioning or masking difficulties that would not be reflected in functional data, it would be important to further explore the specific domains that could be used or instead identify profiles to assess student's enrolment into the Raise program.

To be able to refine the selection approach, schools could be supported to use existing and/or emerging administrative data collection tools that enable them to identify which students are most likely to benefit from Raise. For example, the Australian government funded Mental Health Checklist is due to be available for all schools in Australia once developed. The tool is designed to provide a comprehensive, but simple, measure of domains which are known to negatively impact student mental health. Another example of existing data is South Australia's Wellbeing and Engagement Collection (WEC), which enables schools to track a range of positive and negative wellbeing dimensions through student self-report. This tool includes domains such as life satisfaction, optimism, and worries. In an initial scoping approach, Raise could examine what tools schools in each state are collecting and determine to what extent this existing data collection could be utilised by schools in their student selection process.

However, it is also accepted that the limitation of administrative data is that it may not always capture all domains which are important for a selection process. Therefore, another possibility is that Raise designs a specific selection tool which can be used by schools to select students to participate in the Raise Mentoring Program. This approach would enable Raise to include items to measure specific domains of student wellbeing that Raise will be specifically aiming to address through the program. This would increase the specificity of the selection process to potentially increase the benefits of the program.

## Recommendation 2: Further examining of why demographic variables are predictive of the changes observed Raise mentees

In all the adjusted and interaction analyses, student gender (female, yes/no) and difficult family circumstances at baseline were commonly shown to be a predictor of student change outcomes for Raise mentees.

This suggests further research is required to examine the specific mechanisms for why these particular groups may have different outcomes compared to other students, and the extent these differences warrant consideration in how students are identified to participate in Raise or are supported during the program. For instance, although gender may result in some students receiving a higher benefit from the program, it would not mean that student gender should be part of the recruitment criteria. Instead, it would simply be an aspect of the program that females may be more likely to benefit more from the intervention. Furthermore, this difference may also be related to how Raise is implemented, with mentors being more likely to be female which may influence the benefits observed by students of other genders.

The benefits observed through Raise for students with a difficult family circumstance may not be as large because the family circumstances are beyond the scope of impact and influence of Raise. For instance, Raise does not address a student's exposure to negative life events. Instead, Raise supports the students at school. Therefore, the outcomes of these students are more challenging to shift in the absence of additional support that may be required within and outside of the school system to address the difficult life event more directly. The positive outcomes of Raise were still present when accounting for students who had a difficult life event, and outcomes were not negative for these students but were slightly attenuated.

## Recommendation 3: Building on the promising independent findings to conduct robust, cluster randomised controlled trials

The promising findings suggest that Raise mentees have more positive outcomes compared to matched comparisons, although not all findings were statistically significant. The limited statistically significant findings may be related to the sample size which provided outcome data, although depending on outcome the minimum detectable effect size ranged from 0.2 to 0.28 between groups. There were a significant number of outcomes with effect sizes of between 0.20 to 0.25. It is also important to note that population-based interventions aiming to address student wellbeing outcomes often have effect sizes of below 0.25 in psychosocial outcomes.

However, the positive outcomes that were observed would suggest that Raise has met the criteria for 'Level 3 – High Confidence' in terms of evidence, according to the Australian Education Research Organisation evidence standards.

Australian Education Research Organisation

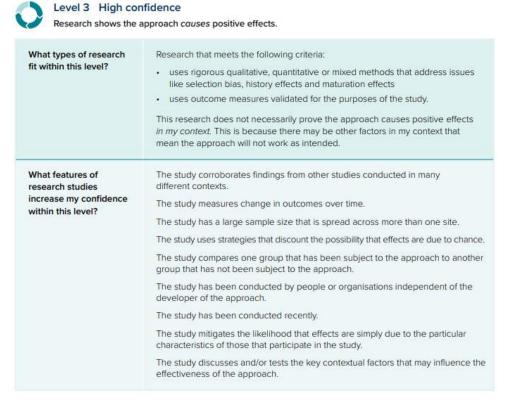


Figure 8: Australian Education Research Organization - Standards of Evidence

The promising findings from this evaluation would benefit from further independent evaluation using methods which also include a comparison group. Simply, this may require a replication quasi-experimental study, whereby the aim will be to determine whether the findings in this evaluation are replicable so as to increase the confidence in the findings.

Another approach would be for the Raise Program to be evaluated via a large-scale project that would meet the AERO criteria for 'Level 4 – Very High Confidence'. It is common for promising findings in smaller quasi-experimental studies to be used to inform the design of larger, cluster randomised controlled trials to examine the outcomes of an intervention when implemented with a broader population of students in different contexts. This is because the between-group effect sizes observed in this evaluation can be used to inform the larger study, so as to ensure it has sufficient sample size and is using the appropriate design

to examine the program. However, there are pragmatic considerations when conducting a study of this type. For instance, the student sample would be expected to be randomised after baseline data collection has occurred, with students who may benefit from Raise being randomly allocated to either receive the program or be a comparison group. Furthermore, it would be likely that all participating schools would have to be new to the program, to avoid contamination in terms of non-Raise mentees receiving part of the program unintentionally. However, this type of research often has significant financial and resourcing requirements, which is why they are not common for commercial-based programs.

Given difficulties with control group recruitment and data collection, a possible alternative would be to compare future Raise evaluation data to existing population data or administrative data. Similar to Recommendation 1, this could involve aligning outcome data collection with the Mental Health Checklist or South Australia's WEC and requesting data linkage for mentees. However, schools may collect this population data at different times and use it for different purposes. If this approach were taken, it would therefore be important to acknowledge any differences in the timeframes of data collection.

A final approach to building the evidence for Raise would be to use a longitudinal approach, whereby participants are followed up at certain intervals to capture their outcomes, and to include specific measures which enable them to reflect on the contribution of participating in Raise on those outcomes. For instance, following up cohorts the year after they were due to finish high school would capture the proportion of students who were able to complete high school, or those who have been able to secure employment by the age of 24. This would align with existing policy reports which have demonstrated the significant social and fiscal burden of students not completing high school and not being in employment by the age of 24. Although this would not necessarily involve a comparison group, the use of existing administrative data could enable a level of comparison to be conducted.

## Recommendation 4: Consider ways to increase the reach and length of the program

One theme in qualitative feedback from mentees was that they could have further benefited from the program if it was extended in duration or included more interactions. A recommendation is for Raise to explore to what extent this feedback is provided in general. However, students expressing negatively that the program has ended can be interpreted in two ways. Positively, it could be an indicator of the benefit they believe they are receiving from their participation, and not wanting the program to end. Conversely, it may reflect that they believe they require more of the program to have a positive impact on their outcomes. Existing mentoring research has consistently found that most benefits of SBM are not observed until at least one full year of regular mentoring sessions (Herrera, 2004; Lee & Cramond, 1999). It is accepted in intervention research that it is rare that all participants in a program will benefit to the same extent. Therefore, a consideration is whether there are modifications to the intervention for some participants to increase the likelihood that all participants benefit to a similar degree.

For instance, a potential approach could be to offer the program to students for a second year, targeting students who required additional mentoring support to either realise positive outcomes and/or consolidate the benefits further. This would require a review of each mentee's outcomes at the end of the year and offering additional support the following year for those identified as need this extension. However, consideration of whether the current activities are repeated or if a different manual is created needs to be considered.

Another approach could be through increasing the capacity of school staff via a professional development approach that aims to build the capacity of school staff to provide support, which is complementary to the content covered in the Raise program. This approach would enable school staff to provide support to students on the other days in which the mentoring program is not conducted, which could lead to greater benefit to students without requiring additional mentor commitment and resourcing. This approach would also enable all educators involved in the student's learning to support them.

To what extent, if any, does the current intervention model need to be extended or supplemented could be explored in future evaluation activities.	эе

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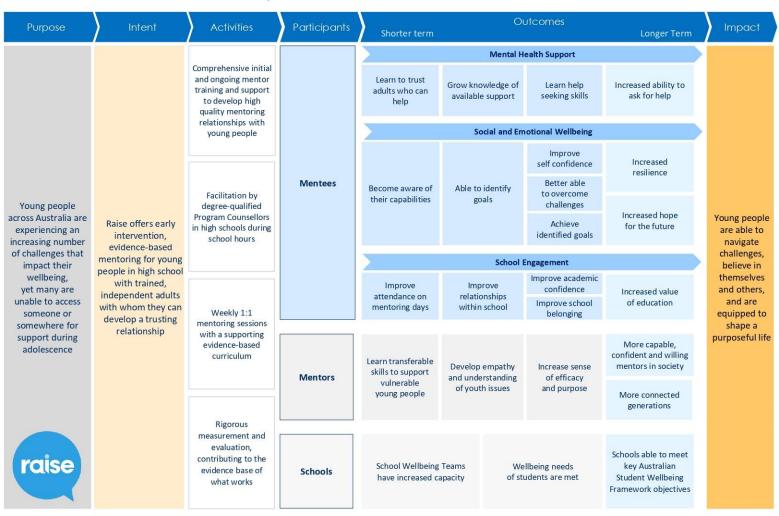
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## **Appendices**

### Appendix A. Raise Theory of Change.

Raise Foundation - How We Create Impact



## Appendix B. Raise Program Logic.

Advisory Our young people have the opportunity to benefit from having a neutral adult role model who they can trust.  Mentoring Programs for Young People  Lack of evidence based, best practice mentoring programs  Many mentoring programs  Many mentoring programs  Many mentoring programs  Many mentoring programs on ont meet definition of mentoring, and are not rigorously evaluated.  Materials  Materials  Materials  Advisory  Program  Mentor information and orientation session  Mentor sessions  Mentor information and orientation session  Mentor information and orientation session  Mentor sessions  Mentor information and orientation session  Mentor sessions  Mentor information and orientation and orientation session  Mentor sessions  Mentor information and orientation session  Mentor sessions  Mentor information and orientation session  Mentor sessions  Mentor information and orientation and orientation and orientation session  Mentor sessions  Mentor information and orientation and orientation and orientation and orientation and orientation sessions  Mentor sessions  Mentor information and orientation sessions  Mentor sessions  Mentor information and orientation	Needs	Inputs	Activities	Outputs	Intermediate Outcomes	Long Term outcomes
struggling  - Mentees - Schools - Parents/carers - Engage, recruit, train, retain and support - Ensure best practice youth safety measures in place - Ensure best practice youth safety measures in place - Ensure best practice youth safety measures in place - Ensure best practice youth safety measures in place - Ensure best practice youth safety measures in place - Raise Ram - Raise Board of Directors - Raise Ambassadors - Raise Ambassadors - Raise Ambassadors - Raise Board of Directors - Raise Ambassadors - Raise Board of Directors - Raise Ambassadors - Raise Ambassadors - Raise Ambassadors - Raise Board of Directors - Raise Ambassadors - Raise Ambassadors - Raise Board of Directors - Raise Ambassadors - Raise Ambassadors - Raise Staff Team - Raise Mentors - Raise Mentors - Raise Board of Directors - Raise Ambassadors - Raise Ambassadors - Raise Ambassadors - Raise Staff Team - Raise Mentors - Raise Board of Directors - Raise Ambassadors - Raise Staff Team - Raise Mentors - Raise Ambassadors - Raise Ambassadors - Raise Ambassadors - Raise Board of Directors - Raise Ambassadors - Raise Raim - Raise Raim						
<ul> <li>Program Material</li> <li>Raise Mentor Training Course</li> <li>Conduct mentee and mentor focus groups</li> <li>Mof evaluation by mentees, mentors, schools and staff</li> <li>Employer satisfaction</li> </ul>	struggling  Suicide is the leading cause of death for young people, more than car accidents or cancer.  1 in 4 young people are unhappy with their lives, yet only 36% of young people will ask for help.  Our young people have the opportunity to benefit from having a neutral adult role model who they can trust.  Mentoring Programs for Young People  Lack of evidence based, best practice mentoring programs  Many mentoring programs do not meet definition of mentoring, and are not	Mentees     Schools     Parents/carers  Raise Team     Raise Mentors     Raise Staff Team     Raise Board of Directors     Raise Ambassadors  Advisory     Patron's Advisory Council     Youth Advisory Council     School Advisory Council     Research Advisory Council     Research Advisory Council     Fundraising Advisory Council     Centre for Social Impact     AYMN  Funding     Government     Corporate     Community     Private Individuals     Events  Materials     Program Material	Engage, recruit, train, retain and support Ensure best practice youth safety measures in place  Schools Engage, recruit and retain school partners Engage parent/carer support  Program Mentee information and orientation sessions Mentor info session Initial 'Jitters' session Matching process for mentees with mentors Weekly mentoring Weekly evidence-based curriculum Weekly supervision Graduation celebration  Evaluation Conduct pre and post surveys with stakeholders Post-intervention school and parent/carer surveys Conduct mentee and mentor focus	<ul> <li>16 modules best practice training for all mentors (online and f2f)</li> <li>Mentee and Mentor Orientation sessions</li> <li>"Jitters" intro session</li> <li>20 weekly sessions of one-to-one mentoring</li> <li>20 weekly mentor supervision with PC</li> <li>Parent consent, match agreement, school partner forms</li> <li>Mentees set, track and achieve chosen goals</li> <li>Graduation celebration</li> <li>Mentor + mentee satisfaction with induction, training, supervision, program</li> <li>School and parent/carer satisfaction</li> <li>Dosage</li> <li>Number of programs</li> <li>Avg number of mentees</li> <li>% of retention rates</li> <li>Avg number of sessions</li> <li>Avg number of goals</li> <li>% of evaluation by mentees,</li> </ul>	<ul> <li>Support from non-parental adult (mentor)</li> <li>Increased social networks</li> <li>Mentees - Intermediate</li> <li>Confidence</li> <li>Communication skills</li> <li>Coping strategies</li> <li>Help seeking skills</li> <li>Positive relationships with friends, family and teachers</li> <li>Resilience</li> <li>Able to set and achieve goals</li> <li>Hope for the future</li> <li>Engagement with education</li> <li>Attendance</li> <li>Grades</li> <li>Employability skills</li> <li>Mentors</li> <li>Leadership skills</li> <li>Ability to mentor others</li> <li>Understanding of youth</li> <li>Sense of purpose</li> <li>Connection to community</li> <li>Parenting skills</li> <li>Increased social networks</li> </ul>	movement creating thriving communities across Australia"  Mentees  Engaged with education or employment Good mental health and psychological wellbeing Participating in community Emerging role models Leadership qualities  Mentors Increased: quality of relationships engagement with their workplace (corporate) wellbeing of workplace (corporate) corporate community networks (corporate) skills (industry mentors) skills for employment employment networks (industry and corporate mentors)

### Appendix C. Cohort Baseline and Follow-Up Survey.

What is your date  Vear level ar Vear 8 Vear 9 Vear 10  What gender do y Male Female Other Rather n  Were you born in Ves No  Do you speak a lar Yes, mos No  Who do you live w One pare Both pare Carer or Other  Are you a person w Prefer ne	re you in?  you identify as?
What is your date  What year level ar  Year 7  Year 8  Year 9  Year 10  What gender do y  Male  Female  Other  Rather n  Were you born in  Yes  No  Do you speak a lar  Yes, mos  Yes, som  No  Who do you live w  One pare  Both pare  Carer or  Other  Are you a person w  Yes  No  Prefer ne	e of birth?  re you in?  you identify as?
What year level ar	re you in?  you identify as?
O Year 7 O Year 8 O Year 9 O Year 10  What gender do you Male O Female O Other O Rather n  Were you born in O Yes O No  Do you speak a lar O Yes, mos O Yes, som O No  Who do you live wow One part O Both part Carer or O Other  Are you a person wow Prefer no	you identify as? not say
O Year 7 O Year 8 O Year 9 O Year 10  What gender do you Male O Female O Other O Rather n  Were you born in O Yes O No  Do you speak a lar O Yes, mos O Yes, som O No  Who do you live wow One part O Both part Carer or O Other  Are you a person woon Prefer no	you identify as? not say
O Year 9 O Year 10  What gender do you O Male O Female O Other O Rather n  Were you born in O Yes O No  Do you speak a lar O Yes, mos O Yes, som O No  Who do you live wo One part O Garer or O Other  Are you a person wo Yes No Prefer no	you identify as?
O Year 10  What gender do you Male O Female O Other O Rather n  Were you born in O Yes O No  Do you speak a lar O Yes, mos O Yes, som O No  Who do you live w One part O Both part Carer or O Other  Are you a person wo Yes No Prefer no	you identify as?
What gender do you hale have you born in his yes had a lare yes, some how	you identify as?
<ul> <li>Male</li> <li>Female</li> <li>Other</li> <li>Rather n</li> </ul> Were you born in <ul> <li>Yes</li> <li>No</li> </ul> Do you speak a lare <li>Yes, some</li> <li>No</li> Who do you live well on the particular of the particular	not say
<ul> <li>Male</li> <li>Female</li> <li>Other</li> <li>Rather n</li> </ul> Were you born in <ul> <li>Yes</li> <li>No</li> </ul> Do you speak a lare <li>Yes, mose</li> <li>Yes, some</li> <li>No</li> Who do you live well on the particular of the particular	not say
Other Rather n  Were you born in Yes No  Do you speak a lar Yes, mos Yes, som No  Who do you live w One par Both par Carer or Other  Are you a person v Yes No Prefer ne	not say
O Rather n  Were you born in O Yes O No  Do you speak a lar O Yes, mos O Yes, som O No  Who do you live w One par O Both par O Carer or O Other  Are you a person w Yes No Prefer no	
Were you born in  Yes  No  No  Do you speak a lar  Yes, mos  Yes, som  No  Who do you live w  One part  Both part  Carer or  Other  Are you a person w  Yes  No  Prefer no	
O Yes No  Do you speak a lar Yes, mos Yes, som No  Who do you live w One part Both part Carer or Other  Are you a person w Yes No Prefer no	Australia?
O Yes No  Do you speak a lar Yes, mos Yes, som No  Who do you live w One part Both part Carer or Other  Are you a person w Yes No Prefer no	
Do you speak a lar  Yes, mos  Yes, som  No  Who do you live w  One pare  Both pare  Carer or  Other  Are you a person w  Yes  No  Prefer no	
<ul> <li>Yes, mos</li> <li>Yes, son</li> <li>No</li> </ul> Who do you live w <ul> <li>One part</li> <li>Both part</li> <li>Carer or</li> <li>Other</li> </ul> Are you a person w <li>Yes</li> <li>No</li> <li>Prefer no</li> Do you help look as	
O Yes, mos No  Who do you live wo Done part Carer or Other  Are you a person wo Yes No Prefer no	inguage other than English at home?
<ul> <li>Yes, som</li> <li>No</li> <li>Who do you live w</li> <li>One part</li> <li>Both part</li> <li>Carer or</li> <li>Other</li> </ul> Are you a person w <ul> <li>Yes</li> <li>No</li> <li>Prefer no</li> </ul> Do you help look a	
Who do you live w One pare Both pare Carer or Other  Are you a person w Yes No Prefer no	
One par. Both par. Carer or Other  Are you a person vor Yes No Prefer no	
One par. Both par. Carer or Other  Are you a person vor Yes No Prefer no	with?
<ul> <li>Both particle</li> <li>Carer or</li> <li>Other</li> </ul> Are you a person or <ul> <li>Yes</li> <li>No</li> <li>Prefer no</li> </ul> Do you help look a	
<ul> <li>Carer or</li> <li>Other</li> </ul> Are you a person or <ul> <li>Yes</li> <li>No</li> <li>Prefer no</li> </ul> Do you help look a	
Are you a person vortes No Prefer no	r guardian
<ul><li>Yes</li><li>No</li><li>Prefer no</li></ul>	
<ul><li>Yes</li><li>No</li><li>Prefer no</li></ul>	with a disability?
<ul><li>No</li><li>Prefer no</li></ul> Do you help look a	
Do you help look a	
	not to say
	after any family members or friends who need your support because of a disability, health condition or mental
o Yes	
o No	
o Prefer n	not to say
In the past month,	n, how many school days have you been absent for?
Have you experier	
discrimination)?	nced any difficult home or personal events (e.g. mental health, family separation, financial difficulties
o Yes	nced any difficult home or personal events (e.g mental health, family separation, financial difficulties,
o No	nced any difficult home or personal events (e.g mental health, family separation, financial difficulties,

If yes, to what degree have these events impacted on your ability to engage in school activities (e.g. classroom attention, homework)?

0 0 0 0 0

#### **About your school**

The purpose of this section is to learn more about you and your school. It is important that you answer every question as honestly as possible. This information will not be used to grade you, or decide which classes you get to take. We will use the information from all students to tell your school what they are doing well to support students, and what they can do to do a better job.

On a scale of 1 to 10, how much do you agree with these statements? (With 1 being strongly disagree and 10 being strongly agree)

Please choose only one answer.

	Strongly Disagree									Strongly Agree
My family knows how I am doing in school	1	2	3	4	5	6	7	8	9	10
I like most of my teachers	1	2	3	4	5	6	7	8	9	10
If I do not know what something means, I do something to figure it out.	1	2	3	4	5	6	7	8	9	10
I study at home	1	2	3	4	5	6	7	8	9	10
I plan to pursue more education after high school.	1	2	3	4	5	6	7	8	9	10
There is someone in my family who helps me when I have trouble completing my homework.	1	2	3	4	5	6	7	8	9	10
Most days, I look forward to going to school.	1	2	3	4x	5	6	7	8	9	10
I pay attention to my teachers.	1	2	3	4	5	6	7	8	9	10
When I am doing school work, I make sure I understand what I am learning	1	2	3	4	5	6	7	8	9	10
I look for more information about things we are learning in school	1	2	3	4	5	6	7	8	9	10
My school work is important	1	2	3	4	5	6	7	8	9	10
Being successful in school will help me in the future	1	2	3	4	5	6	7	8	9	10
I am proud to be a student at this school.	1	2	3	4	5	6	7	8	9	10
When learning new things, I try to connect them to things I already know.	1	2	3	4	5	6	7	8	9	10
When I have an assignment due, I keep working until it is finished	1	2	3	4	5	6	7	8	9	10
Getting good grades is important to me.	1	2	3	4	5	6	7	8	9	10
It is important to me to be successful in a job	1	2	3	4	5	6	7	8	9	10
I talk to my family about problems I have at school.	1	2	3	4	5	6	7	8	9	10
There is a lot I can learn from my teachers.	1	2	3	4	5	6	7	8	9	10
Teachers help me to be successful at school.	1	2	3	4	5	6	7	8	9	10
I know how to study for tests	1	2	3	4	5	6	7	8	9	10
I feel like a part of my school.	1	2	3	4	5	6	7	8	9	10

About your connections	
About your connections	

	Disagree a	Disagree	Neutral	Agree	Agree a lot
I usually fit in with other kids around me					
I feel like part of a group of friends					
I have a friend I can tell everything to					
I feel like I belong at school					
I feel comfortable talking to teachers					
I can be myself at school					
I have friends at school					

If you were having a personal or emotional problem, how likely is it that you would seek help from the following people?

Please indicate your response by putting a line through the number that best describes your intention to seek help from each help source that is listed. 1 = Extremely Unlikely 3 = Unlikely 5 = Likely 7 = Extremely Likely

	Extremely Unlikely		Unlikely		Likely		Extremely Likely
	1	2	3	4	5	6	7
a. Intimate partner (e.g., girlfriend, boyfriend)							
b. Friend (not related to you)							
c. Parent							
d. Other relative/family member							
e. Mental health professional (e.g. psychologist, social worker, counsellor)							
f. Phone helpline (e.g. Lifeline)							
g. Doctor/GP							
h. Minister or religious leader (e.g. Priest, Rabbi, Chaplain)							
i. I would not seek help from anyone							
j. I would seek help from another not listed above (please list in the space provided, (e.g., work colleague							

These next few questions are about how things are going for you:

Do you feel able to trust adults who can help when you need it?	Always	Sometimes	Rarely	Never
Do you feel able to ask for help from others when you need it?	Always	Sometimes	Rarely	Never
Do you feel like you know where to get help when you need it?	Always	Sometimes	Rarely	Never

This is a survey about you! Please read each of the following statements. Circle how much each statement describes you. Please be honest - there are no right or wrong answers!

About your well-being

	Almost Never	Sometimes	Often	Very Often	Almost Always
When something good happens to me, I have people who I like to share the good news with.					
I finish whatever I begin.					
I am optimistic about my future					
I feel happy.					
When I do an activity, I enjoy it so much that I lose track of time.					
I have a lot of fun.					
I get completely absorbed in what I am doing.					
I love life.					
I keep at my schoolwork until I am done with it.					
When I have a problem, I have someone who will be there for me.					
I get so involved in activities that I forget about everything else.					
When I am learning something new, I lose track of how much time has passed.					
In uncertain times, I expect the best.					
There are people in my life who really care about me.					
I think good things are going to happen to me.					
I have friends that I really care about.					
Once I make a plan to get something done, I stick to it.					
I believe that things will work out, no matter how difficult they seem.					
I am a hard worker.					
I am a cheerful person.					

Below are some statements about your feelings and thoughts. Please indicate how often you felt or thought a certain way during the last 2 weeks.

		None of the time	Rarely	Some of the time	Often	All of the time
a.	I've been feeling optimistic about the future					
b.	I've been feeling useful					
c.	I've been feeling relaxed					
d.	I've been dealing with problems well					
e.	I've been thinking clearly					
f.	I've been feeling close to other people					
g.	I've been able to make up my own mind about things					

These questions relate to how you cope when things don't go quite right. Circle the response that describes YOU the best.

I tend to bounce back quickly after hard times	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
I have a hard time making it through stressful events	Strongly disagree	Disagree	Neutral	Agree	Strongly agree

It does not take me long to recover from a stressful event	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
It is hard for me to snap back when something bad happens	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
I usually come through difficult times with little trouble	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
I tend to take a long time to get over setbacks in my life	Strongly disagree	Disagree	Neutral	Agree	Strongly agree

,
Raise Program (only asked at follow-up)
Did you participate in the Raise mentoring program in 2023?
☐ Yes – branching – next items
□ No − branching − end survey
What has, if any, been the most significant benefit of participating in the Raise program?
What has, if any, been a negative of participating in the Raise program?
How would you describe your engagement with your mentor?
We are interested in hearing more from participants about their experience in participating in the Raise program. Would you interested in hearing more about participating in this aspect?
□ Yes
$\square$ No

## Appendix D. Mentee 'Pulse' Survey.

What school do you attend?		
What is your date of birth?		

What year level are you in?

- o Year 7
- o Year 8
- o Year 9
- O Year 10

What gender do you identify as?

- o Male
- Female
- o Other
- o Rather not say

How many Raise sessions have you attended in the past four weeks?

- 0
- 0 1
- o **2**
- o **3**
- 0 4

In the last 4 weeks have you had to match with a different mentor during the individual mentoring time?

- Yes
- o No

The following statements are about your relationship with your matched mentor in the past four weeks. Please indicate how much you agree with each statement.

	Not at all	A little	Sometimes	Mostly	Always
I feel safe with my mentor	0	0	0	0	0
My mentor helps me to share my problems	0	0	o	0	0
My mentor helps me come up with ideas to cope with my problems	o	0	0	o	o

The following statements are about you in the last four weeks. Please indicate how much you agree with each statement.

	Not at all	A little	Sometimes	Mostly	Always
I am finding the Raise program useful for me	0	0	О	0	0
I feel that I really belong in my school	0	0	0	0	0
I can bounce back quickly after hard times	0	0	O	o	O
I feel hopeful about the future	0	o	O	o	0
	o	0	О	0	o

I am able to seek hothers when I nee						
Only asked at mid	lpoint_					
Have you experien discrimination)?  O Yes  No	nced any difficult	: home or personal eve	ents (e.g mental heal	th, family separation	n, financial difficulties,	
If yes, to what deg homework)?	ree have these e	events impacted on yo	ur ability to engage i	n school activities (e	e.g. classroom attention	•
,	None of the tin	ne A little of the time o	e Some of the time o	Most of the time o	All of the time o	
			v	·	· ·	

### Appendix E. Mentor Baseline and Follow-Up Survey.

If you were having a personal or emotional problem, how likely is it that you would seek help from the following people? Please indicate your response by putting a line through the number that best describes your intention to seek help from each help source that is listed. 1 = Extremely Unlikely 3 = Unlikely 5 = Likely 7 = Extremely Likely

	Extremely Unlikely		Unlikely		Likely		Extremely Likely
	1	2	3	4	5	6	7
a. Intimate partner (e.g., girlfriend, boyfriend, husband, wife, de facto)							
b. Friend (not related to you)							
c. Parent							
d. Other relative/family member							
e. Mental health professional (e.g. psychologist, social worker, counsellor)							
f. Phone helpline (e.g. Lifeline)							
g. Doctor/GP							
h. Minister or religious leader (e.g. Priest, Rabbi, Chaplain)							
i. I would not seek help from anyone							
j. I would seek help from another not listed above (please list in the space provided, (e.g., work colleague							

Below are some statements about your feelings and thoughts. Please indicate how often you felt or thought a certain way during the last 2 weeks.

		None of the time	Rarely	Some of the time	Often	All of the time
h.	I've been feeling optimistic about the future					
i.	I've been feeling useful					
j.	I've been feeling relaxed					
k.	I've been dealing with problems well					
I.	I've been thinking clearly					
m.	I've been feeling close to other people					
n.	I've been able to make up my own mind about things					

These questions relate to how you cope when things don't go quite right. Circle the response that describes YOU the best.

I tend to bounce back quickly after hard times	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
I have a hard time making it through stressful events	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
It does not take me long to recover from a stressful event	Strongly disagree	Disagree	Neutral	Agree	Strongly agree

It is hard for me to snap back when something bad happens	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
I usually come through difficult times with little trouble	Strongly disagree	Disagree	Neutral	Agree	Strongly agree
I tend to take a long time to get over setbacks in my life	Strongly disagree	Disagree	Neutral	Agree	Strongly agree

## **Appendix F. Student Perspective Questions and Data from Evaluation Mentees.**

What has, if any, been the most significant benefit of participating in the Raise program?	What has, if any, been a negative of participating in the Raise program?	How would you describe your engagement with your mentor?
talking things on my mind out	missing out on core subjects	Great
Talking about issues with my mentor.	No	Good. I liked talking to her about life.
Not going up Class	Nothing	Good he was nice and funny
Not being in class	Could have been longer	Good. She was nice and good to talk to.
Nothing I was never here	Nothing was never here	I liked her. She can draw
talking to my mentor	going back to class after raise	very engaging
Learning skills	Nothing	Okay
Having someone in you corner	Talking to strangers	Good
Having someone to talk to who isn't a medical professional.	The games at the start of the period sometimes feel like I'm forced into them when sometimes i don't want to do them.	Very good. We had a very good connection.
I loved talking to my mentor every week and i loved the free food that they provided.	It took a week or two to get used to but then it went awesome.	Good I enjoyed getting to talk to her every week.
food?	qaiting for food	Great
Feeling listened to though other don't at times.	None	Pretty good
being able to talk to someone.	nothing	Good
Getting to speak to a person for a while about what I like to do.	No things.	We could take and engage a lot
I was able to trust my mentor and felt like I was able to improve some learning skills and had someone to talk to about my problems.	There wasn't any.	I felt I was able to warm up to her in the end and we had fun and enjoyed the time we had together.
having someone to talk to and trust	nothing	really engaged
Talking to someone I can trust	It only going for one period	We hit it off right away
got to talk to someone	nothing	easy to talk
This program has helped me and my mental health a lot I've been more happy and optimistic ever since I started this program	nothing the program has helped me so much and in no way has it been a negative effect on me	It was good we were both chill we were a bit jittery at the start but we had gotten a long swell as time passed and as we got to knowing each other
being able to speak to someone	nothing	very happy and talkative
food.	waitig fpr food.	good shes rly nice i love her
The program helped me to open up more and be more confident in myself.	Nothing	My mentor was an amazing person who has always been so kind and listens a lot.
Getting to share my responses was the most significant benefit of participating in the program.	Nothing.	I would say that it was really good.
realize who i am	realize the bad effects	i dont know
talking to the mentors.	there hasn't been any negative participating in Raise .	my engagement with my mentor is very good.
Thinking more about myself, doing more reflection		Positive
hep you set goals for the future	nothing really	good and fun

# Appendix G. Additional Student Perspective Questions and Data from Evaluation Mentees.

Tell me a little bit about why you participated in the Raise program?	Describe how you engaged in the Raise activities and sessions throughout the year.	What have been your key takeaway learnings from the program?	Describe the changes you have noticed about yourself through participating in Raise.	What was the best part of the program?	What would you change about the program?	Is there anything else you would like to add?
i joined the program because i enjoyed it and it was fun.	I engaged in all the active by participating and answering the question.	in the program i learn to be confident, set gaols for my future and to be myself no matter where i am.	since I joined raise I learn things about myself such as being more confident and i can concentrate more.	the best part of the program was talking to my mentor.	I wouldn't change anything about the program because everything is fun and useful.	there is nothing I would like to add.
To build up my confidence and meet new people	I was very engaged I tried everything they said to do	Confidence	I got more confidence	Getting to meet my mentor	For it to stay until the end of the year	No

Appendix H. Analysis Output	
Following Pages	



## **Learning Intervention**

Faculty of Education 100 Leicester Street, The University of Melbourne, 3010 VIC Email: education-li@unimelb.edu.au

name: <unnamed> log: C:\Users\quachjl\OneDrive - The University of Melbourne\Documents\1\_Projects\1\_InProgress\2022\_RAISE Evaluation\10\_Data Analysis\ALL\Stata\01\_cohort charac > teristics\_26 Mar 2024.log log type: text opened on: 26 Mar 2024, 20:31:37 \*\*\*\*\*\*\*\*\*\*\*\*\* . \*\*\* Baseline key demographics for the whole cohort \*\*\*\*\*\*\*\*\*\*\*\*\* foreach var of varlist student\_year student\_gender student\_australia student\_language student\_disability student\_live student\_support student\_absent > ifficultevent student\_impact { tab `var' 2. 3. What year | level are you in? | Freq. Percent Cum. ----- 

 Year 7 | 13
 1.72
 1.72

 Year 8 | 498
 65.96
 67.68

 Year 9 | 244
 32.32
 100.00

 Total | 755 100.00 What gender do you identify | as? | Freq. Percent Cum. ----- 

 Male |
 371
 49.73
 49.73

 Female |
 326
 43.70
 93.43

 Other |
 23
 3.08
 96.51

 Rather not say |
 26
 3.49
 100.00

 ------Total | 746 100.00 Were you born in | Australia? Freq. Percent Cum.

Do you speak a |

-----

-----

Total | 744 100.00

No | 162 21.77 21.77 Yes | 582 78.23 100.00

<pre>language other   than English     at home?</pre>		Fred	۹۰	Percent	Cı	um. 
	   	18		51.20 24.53 24.27		<b>.</b> 73
Total	•	7!	50	100.00		
Are you a per wit disabili	h a	ļ	Freq.	Percen	t	Cum.
Prefer not to	-	į	657 41 44		3	88.54 94.07 100.00
То		+ 	742	100.0	0	
Who do you l wi	ive th?		Freq.	Percen	t	Cum.
One par Both pare Carer or guard Ot	nts	    -		73.9 1.7	7 4	
То	tal	+ 	749	100.0	0	
Do you help l after any fam members friends who n your supp be	ily or eed	       	Freq.	Percen	t	Cum.
	No	•		69.0		69.08
Prefer not to	Yes say	 	119 43	22.7 8.2	1 1 :	
То	tal	+ 	524	100.0	0	
In the past   two weeks,   how many   school days   have you   been absent   for?		Freq.	Pe	ercent	Cum.	
0   .5   1		251 2 82		52.07 0.41 17.01	52.07 52.49 69.50	

2 3 3.5 4 5 6 6.5 7 9 10 11 12 13 14 15 25 26	 	54 27 1 9 21 6 1 7 3 5 1 2 2 5 1 1	11.20 5.60 0.21 1.87 4.36 1.24 0.21 1.45 0.62 1.04 0.21 0.41 1.04 0.21 0.21	80.71 86.31 86.51 88.38 92.74 93.98 94.19 95.64 96.27 97.30 97.51 97.93 98.34 99.38 99.59 99.79	
Total	+ 	 482	100.00		
Have you experienced any difficult home or personal events (e.g mental health, f	               Fr	eq.	Percent	Cum.	
No Yes	:	336 192	63.64 36.36	63.64 100.00	
Total	+ 	 528	100.00		
degree have events impa your aba		       	Freq.	Percent	Cum.
None of a A little of a Some of a Most of a All of	the time the time	+         	20 52 61 41 15	10.58 27.51 32.28 21.69 7.94	10.58 38.10 70.37 92.06 100.00
	Total	+ 	189	100.00	

sum epoc\_engagement\_pre epoc\_perser\_pre epoc\_optimism\_pre
epoc\_connect\_pre epoc\_happy\_pre helpseeking\_pre SSE\_aspiration\_pre
SSE\_productivity\_pre SSE\_belong\_p

<sup>&</sup>gt; re wemwbs brs\_pre

Variable	0bs	Mean	Std. dev.	Min	Max
epoc_enga~re	373	2.959562	.9696443	1	5
epoc_pers~re	489	3.162747	.9346528	1	5
epoc_opti~re	437	3.016209	1.017311	1	5
epoc_conn~re	437	3.793669	.9261997	1	5
epoc_happ~re	373	3.35992	1.007931	1	5
helpseeki~re	+   340	33.08529	9.273384	10	64
SSE aspir~re	415	8.097791	1.94677	1	10
SSE_produ~re	416	6.395461	1.868436	1	10
SSE_belon~re	415	5.798353	2.096109	1	10
wemwbs	358	21.67075	5.239425	7	35
brs_pre	433	3.046459	.6137383	1	5

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. log close

name: <unnamed>

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Melbourne\Documents\1\_Projects\1\_InProgress\2022\_RAISE Evaluation\10\_Data

Analysis\ALL\Stata\01\_cohort charac

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name: <unnamed>

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Analysis\ALL\Stata\02\_define\_match\_

> 26 Mar 2024.log log type: text

opened on: 26 Mar 2024, 20:31:37

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\*\*\*\*\*\*\*\*\*\*\*\*\*\*

•

foreach var of varlist student\_year student\_gender student\_australia student\_language student\_disability student\_live student\_support student\_absent student\_d

- > ifficultevent student\_impact {
  - tab student\_mentee `var', row chi
  - 3. }

Key	•
frequency	
row percentage	
++	-

IS	tn:	LS
stude	nt	а

Raise mentee?	   What year   Year 7	level are Year 8	•	Total
No	2	395	198	595
	0.34	66.39	33.28	100.00
Yes	11	103	46	160
	6.88	64.38	28.75	100.00
Total	13	498	244	755
	1.72	65.96	32.32	100.00

Pearson chi2(2) = 32.1882 Pr = 0.000

+
Key
frequency
row percentage

+----+

Is this student a Raise mentee?	     What   Male	-	you identif Other	-	Total
No	+   305   51.96	244 41.57	15 2.56	23 3.92	587   100.00
Yes	66	82	8	3	159
	41.51	51.57	5.03	1.89	100.00
Total	371	326	23	26	746
	49.73	43.70	3.08	3.49	100.00

Pearson chi2(3) = 9.5822 Pr = 0.022

+   Ke	 ≘y		+ 
 	frequ	ency	·-  
ro	ow per	centage	٦ - ٦

Is this				
student a	Were you born in			
Raise	Australia?			
mentee?	No +	Yes	Total	
No	126	459	585	
	21.54	78.46	100.00	
Yes	36	123	159	
	22.64	77.36	100.00	
Total	162	582	744	
	21.77	78.23	100.00	

Pearson chi2(1) = 0.0893 Pr = 0.765

+	+
Key	١
	١
frequency	١
row percentage	
+	+

Is this				
student a	Do you speak	a langua	age other	
Raise	than Engl	ish at h	nome?	
mentee?	No Yes	, some	Yes, most	Total
	+			+
No	310	138	143	591

	52.45	23.35	24.20	100.00
Yes	74	46	39	159
	46.54	28.93	24.53	100.00
Total	384	184	182	750
	51.20	24.53	24.27	100.00

Pearson chi2(2) = 2.4516 Pr = 0.294

+-					+
	Key				
-					
	fr	reque	enc	y	-
	row	per	cen	tag	e
<b>+</b> -					+

Is this student a Raise mentee?	   Are yo     No	ou a person disability? Yes		Total
No	517	32	37	586
	88.23	5.46	6.31	100.00
Yes	140	9	7	156
	89.74	5.77	4.49	100.00
Total	657	41	44	742
	88.54	5.53	5.93	100.00

Pearson chi2(2) = 0.7467 Pr = 0.688

+				+
Ke	3À			1
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r	ом ре	ercen	tage	ĺ
+				٠+

Is this student a Raise mentee?	       One paren	Who do you Both pare		Other	Total
No	92	462	9	28	591
	15.57	78.17	1.52	4.74	100.00
Yes	52	92	4	10	158
	32.91	58.23	2.53	6.33	100.00
Total	144	554	13	38	749
	19.23	73.97	1.74	5.07	100.00

		, =:::::				
Key	<del>-</del>					
frequency   row percent	age					
student a   Raise	need yo No	bers or fri ur support	ends who becau Prefer no	Total		
No	262 71.00	70 18.97	37   10.03			
Yes	100 64.52	49 31.61	6   3.87	155		
Total	362	119 22.71	43			
Pea	arson chi2(2	) = 13.388	0 Pr = 0.	001		
Key 	/   cage					
student a   Raise		_		In the pa	st two wee	ks, how many
school days h mentee?	0	.5	1	2	3	3.5
4 Total	5	6	6.5	7	9	10
+-						
No   7				35 3		1 1
1.92 100.00	3.29	1.37	0.27	9.59 0.82	0.82	0.27
Yes   2 117						

1.71 100.00	41.03 7.69	0.85	0.00	3.42	0.00	3.42	
Total 9 482	251 21	2 6 0.41	82 1	54 7 11.20	27 3		
for? mentee? 26	İ	12	13	14	15	been absent 25	
No 0   0.00	365 0.00	2 0.55	1 0.27	5 1.37	0 0.00		
Yes 1   0.85	117 0.85	0 0.00	1 0.85	0.00	1	1	
Total 1   0.21	1 482	2 0.41				1 0.21	
Pearson chi2(19) = 45.5107 Pr = 0.001  +							

No	262	110	372
	70.43	29.57	100.00
Yes	74	82	156
	47.44	52.56	100.00
Total	336	192	528
	63.64	36.36	100.00

Pearson chi2(1) = 25.1130 Pr = 0.000

+	+
Key	l
	İ
frequency	İ
row percentage	
+	+

Is this | student a | If yes, to what degree have these events impacted on Raise | your ability to engage in s mentee? | None of t A little | Some of t Most of t All of th | Total No | 11 | 36 | 39 | 17 | 6 | 109 | 10.09 | 33.03 | 35.78 | 15.60 | 5.50 | 100.00 | 11.25 | 20.00 | 27.50 | 30.00 | 11.25 | 100.00 | Total | 20 | 52 | 61 | 41 | 15 | 189 | 10.58 | 27.51 | 32.28 | 21.69 | 7.94 | 100.00

Pearson chi2(4) = 10.2159 Pr = 0.037

bysort student\_mentee: sum helpseeking2\_1 helpseeking2\_2
helpseeking2\_3 epoc\_engagement\_pre epoc\_perser\_pre epoc\_optimism\_pre
epoc\_connect\_pre helpseeking\_pre

> SSE\_aspiration\_pre SSE\_productivity\_pre SSE\_belong\_pre wemwbs brs\_pre

\_\_\_\_\_\_

-> student mentee = No

Variable	Obs	Mean	Std. dev.	Min	Max
helpseekin~1	304	3.078947	.8406058	1	4
helpseekin~2	304	2.940789	.8383037	1	4
helpseekin~3	303	3.171617	.8320483	1	4
epoc_enga~re	333	2.990741	.9652697	1	5
epoc_pers~re	333	3.202703	.934315	1	5
epoc opti~re	284	3.069836	1.003118	1	5

epoc_conn~re	284	3.842723	.9279131	1	5
helpseeki~re	308	33.30519	9.235485	10	64
SSE_aspir~re	374	8.17803	1.938404	1	10
SSE_produ~re	375	6.444285	1.88862	1	10
+					
SSE_belon~re	374	5.850401	2.102973	1	10
wemwbs	319	21.84752	5.180994	7	35
brs_pre	281	3.126987	.596601	1	5

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<sup>-&</sup>gt; student\_mentee = Yes

Variable	Obs	Mean	Std. dev.	Min	Max
helpseekin~1	153	3.346405	.6316255	2	4
helpseekin~2	152	3.256579	.6458908	2	4
helpseekin~3	151	3.370861	.6389146	2	4
epoc_enga~re	40	2.7	.9792724	1.5	5
epoc_pers~re	156	3.077457	.9326376	1	5
	+				
epoc_opti~re	153	2.916667	1.03912	1	5
epoc_conn~re	153	3.702614	.9190939	1	5
helpseeki~re	32	30.96875	9.519028	16	50
SSE_aspir~re	41	7.365854	1.891674	2.5	10
SSE_produ~re	41	5.948897	1.625589	2.75	8.833333
	+				
SSE_belon~re	41	5.323577	1.994579	1.333333	9.166667
wemwbs	39	20.22487	5.555305	7	35
brs_pre	152	2.897588	.619075	1.166667	5

foreach var of varlist helpseeking2\_1 helpseeking2\_2 helpseeking2\_3
epoc\_engagement\_pre epoc\_perser\_pre epoc\_optimism\_pre epoc\_connect\_pre
helpseeking\_pre SSE

3. }

Two-sample t test with equal variances

Group	0bs	Mean		Std. dev.	-	interval]
No   Yes	304 153	3.078947 3.346405	.048212 .0510639	.8406058 .6316255	2.984075 3.245519	3.17382 3.447292
Combined	457	3.16849	.0367878	.7864341	3.096195	3.240785
diff		2674579	.0770257		418828	1160877
diff =	mean(No)	- mean(Yes)				= -3.4723

H0: diff = 0 Degrees of freedom = 455

Two-sample t test with equal variances

. interval]	-		Std. err.	Mean	Obs	Group
3.035402 3.360088	2.846176 3.153069	.8383037 .6458908	.04808 .0523887	2.940789 3.256579	304 152	No Yes
3.119019	2.973086	.7928706	.0371296	3.046053	456	Combined
1635943	4679847		.077445	3157895		diff

H0: diff = 0

Two-sample t test with equal variances

Group	0bs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
No   Yes	303 151	3.171617 3.370861	.0477999 .0519941	.8320483 .6389146	3.077554 3.268125	3.26568 3.473596
Combined	454	3.237885	.036521	.7781627	3.166114	3.309657
diff		1992438	.0770332		3506315	047856

diff = mean(No) - mean(Yes) t = -2.5865H0: diff = 0Degrees of freedom = 452

Two-sample t test with equal variances

Group	0bs	Mean	Std. err.		[95% conf.	-
No   Yes	333 40	2.990741 2.7	.0528965 .1548366	.9652697 .9792724	2.886686 2.386814	3.094795 3.013186
Combined	373	2.959562	.0502063	.9696443	2.860838	3.058286
diff		.2907407	.1617771		0273743	.6088558
4÷cc -		maan (Vas)			+	_ 1 7072

diff = mean(No) - mean(Yes) t = 1.7972

H0: diff = 0Degrees of freedom = 371

Two-sample t test with equal variances

Two-sample	e t test wi	th equal var	iances			
Group		Mean	Std. err.	Std. dev.	[95% conf.	interval]
No   Yes	333 156				3.101985 2.929954	
			.0422665	.9346528	3.0797	3.245794
diff		.1252454	.0905973		0527644	.3032553
diff = H0: diff =	, ,	- mean(Yes)		Degrees	t of freedom	= 1.3824 = 487
Ha: di Pr(T < t)	iff < 0 ) = 0.9163	Pr(	Ha: diff != T  >  t ) =	0 0.1675	Ha: d Pr(T > t	iff > 0 ) = 0.0837
Two-sample	e t test wi	th equal var	iances			
	Obs		Std. err.	Std. dev.	[95% conf.	interval]
No	284	3.069836			2.952669 2.750693	
•		3.016209		1.017311	2.920563	3.111855
-		.153169			047057	.3533951
diff = H0: diff =	• •	- mean(Yes)		Degrees	t of freedom	= 1.5035 = 435
Ha: di Pr(T < t)	iff < 0 ) = 0.9333	Pr(	Ha: diff != T  >  t ) =	0 0.1334	Ha: d Pr(T > t	iff > 0 ) = 0.0667
Two-sample	e t test wi	th equal var	iances			
					[95% conf.	
No   Yes	284 153	3.842723	.0550615 .0743043	. 9279131	3.734341 3.555812	3.951105
Combined	437	3.793669	.0443061	.9261997	3.706589	
		.1401086	.0927476		0421806	.3223979
	= mean(No)	- mean(Yes)				= 1.5106
Ha: di Pr(T < t)	iff < 0 ) = 0.9342	Pr(	Ha: diff != T  >  t ) =	0 0.1316	Ha: d Pr(T > t	iff > 0 ) = 0.0658
Two-sample	e t test wi	th equal var	iances			

		Mean	Std. err.	Std. dev.	[95% conf.	interval]
No   Yes	308 32	33.30519	1.682742	9.519028	32.2697 27.53677	34.40073
Combined	340		.5029199		32.09606	34.07453
•		2.336445			-1.047263	
<pre>diff = m H0: diff = 0</pre>	, ,	- mean(Yes)		Degrees	t of freedom	= 1.3582 = 338
Ha: diff Pr(T < t) =	< 0 : 0.9123	Pr(	Ha: diff != T  >  t ) =	0 0.1753	Ha: d Pr(T > t	iff > 0 () = 0.0877
Two-sample t	test wi	th equal var	iances 			
		Mean		Std. dev.	[95% conf.	interval]
No   Yes	374 41	8.17803 7.365854	.1002325 .2954298	1.938404 1.891674	7.980939 6.768768	8.375122 7.96294
	415	8.097791			7.909942	
•		.8121766	.3181534		.1867747	1.437579
diff = m H0: diff = 0		- mean(Yes)				= 2.5528
110. UITT - 0	,			Degrees	of freedom	= 413
		Pr(	Ha: diff != T  >  t ) =	_	of freedom Ha: d Pr(T > t	
	• < 0 • 0.9945			_		
Ha: diff Pr(T < t) = Two-sample t Group	<pre></pre>	th equal var  Mean	iances  Std. err.	0 0.0110 Std. dev.	Ha: d Pr(T > t  [95% conf.	iff > 0 ) = 0.0055
Ha: diff Pr(T < t) =  Two-sample t Group   No   Yes	<pre></pre>	th equal var Mean 6.444285 5.948897	iances 	0 0.0110 	Ha: d Pr(T > t  [95% conf.  6.252513 5.435797	iff > 0 ) = 0.0055  interval]  6.636057 6.461996
Ha: diff Pr(T < t) =  Two-sample t Group  + No   Yes   Combined	0.9945 test wir Obs 375 41	Mean	iances Std. err. .0975279 .2538744	0 0.0110 	Ha: d Pr(T > t	iff > 0 ) = 0.0055 interval] 6.636057 6.461996 6.575533
Ha: diff Pr(T < t) =  Two-sample t	0.9945 test wi 0bs 375 41	Mean 6.444285 5.948897 6.395461	iances Std. err0975279 .2538744	0 0.0110 	Ha: d Pr(T > t	iff > 0 ) = 0.0055  interval]  6.636057 6.461996  6.575533  1.09836
Ha: diff Pr(T < t) =  Two-sample t	0.9945 test wir obs 375 41 416	Mean 6.444285 5.948897 6.395461	iances Std. err. .0975279 .2538744 	0 0.0110 	Ha: d Pr(T > t	iff > 0 ) = 0.0055  interval] 6.636057 6.461996 6.575533 1.09836 = 1.6150
Ha: diff Pr(T < t) =  Two-sample t  Group    No   Yes    Combined    diff    diff = m  H0: diff = 0	0.9945 test wi 0bs 375 41 416	Mean 6.444285 5.948897 6.3954614953882 mean(Yes)	iances Std. err. .0975279 .2538744  .0916076 	0 0.0110  Std. dev 1.88862 1.625589 1.868436 Degrees	Ha: d Pr(T > t	iff > 0 ) = 0.0055  interval] 6.636057 6.461996 6.575533 1.09836 = 1.6150 = 414
Ha: diff Pr(T < t) =  Two-sample t  Group    No   Yes    Combined    diff    diff = m H0: diff = 0  Ha: diff Pr(T < t) =	0.9945 test wi 0bs 375 41 416 nean(No) 0 0.9465	th equal var Mean 6.444285 5.948897 6.395461 4953882 Pream(Yes)	iances	0 0.0110  Std. dev 1.88862 1.625589 1.868436 Degrees 0 0.1071	Ha: d Pr(T > t	iff > 0 interval]  6.636057 6.461996  6.575533  1.09836  = 1.6150 = 414 diff > 0 e) = 0.0535
Ha: diff Pr(T < t) =  Two-sample t	0.9945 test wir Obs 375 41 416 nean(No)	Mean	iances	0 0.0110  Std. dev 1.88862 1.625589 1.868436 Degrees 0 0.1071	Ha: d Pr(T > t	iff > 0 ) = 0.0055  interval] 6.636057 6.461996 6.575533 1.09836 1.09836 1.09836 1.09836 1.09836 1.09836 1.09836

Yes   41	5.323577	.311501	1.994579	4.69401	5.953144		
+	5.798353	.1028939	2.096109	5.596094	6.000613		
diff				 - <b>.</b> 1499293			
<pre>diff = mean(No) H0: diff = 0</pre>	- mean(Yes)		Degrees	t of freedom	= 1.5302 = 413		
Ha: diff < 0 Pr(T < t) = 0.9366	Pr(	Ha: diff != T  >  t ) =	0 0.1267	Ha: d Pr(T > t	iff > 0 () = 0.0634		
Two-sample t test wi	th equal var	iances					
Group   Obs		Std. err.	Std. dev.	[95% conf.	interval]		
No   319 Yes   39	21.84752	.29008					
Combined   358		.2769122	5.239425	21.12617	22.21534		
diff		.8858694		1195432	3.364847		
<pre>diff = mean(No) H0: diff = 0</pre>	- mean(Yes)		Degrees	t of freedom	= 1.8317 = 356		
Ha: diff < 0 Pr(T < t) = 0.9661	Pr(	Ha: diff != T  >  t ) =	0 0.0678	Ha: d Pr(T > t	iff > 0 () = 0.0339		
Two-sample t test wi	th equal var	iances					
Group   Obs	Mean	Std. err.	Std. dev.	[95% conf.	interval]		
No   281 Yes   152	2.897588	.0502136	.619075	2.798376			
Combined   433					3.104429		
diff	.2293992	.0608717		.1097569	.3490416		
<pre>diff = mean(No) H0: diff = 0</pre>	- mean(Yes)		Degrees	t of freedom	= 3.7686 = 431		
Ha: diff < 0 Pr(T < t) = 0.9999	Pr(	Ha: diff != T  >  t ) =	0 0.0002	Ha: d Pr(T > t	iff > 0 () = 0.0001		
<pre> . ************************** . ********</pre>							

for propensity score matching
. logistic student\_mentee student\_year student\_gender student\_australia

# $\verb|student_language| student_disability| student_difficult event|$

Logistic regression	LR	chi2(6)	s = 513 = 52.35 = 0.0000				
Log likelihood = -286.4324	Ps	eudo R2	= 0.0837				
student_mentee   Odds ratio Std. err. interval]	z	P> z	[95% conf.				
 student_year   .4404501 .0895811 .6561733	-4.03	0.000	.2956479				
student_gender   1.074111 .1558456 1.427421	0.49	0.622	.8082502				
student_australia   1.055102 .288042 1.801656	0.20	0.844	.6178983				
student_language   1.487713 .2060708 1.951747	2.87	0.004	1.134004				
student_disability   .6279462 .1401576 .972547	-2.08	0.037	.4054472				
student_difficultevent   2.929617 .6285888 4.461158	5.01	0.000	1.923862				
_cons   1.290872 .7760974 4.194132	0.42	0.671	.3973053				
Note: _cons estimates baseline odds.							
<pre> ** Found that student_year and student_disability were related to decrease in likelihood of being a mentee . ** Found student_language, and difficult_event were related with increased likelihood of being a mentee</pre>							
<ul> <li>** Identify baseline outcomes which are related to treatment groups for propensity score matching</li> <li>logistic student_mentee epoc_perser_pre epoc_optimism_pre epoc_connect_pre</li> </ul>							
Logistic regression	LR	chi2(3)	s = 437 = 2.77				
Log likelihood = -281.57969			= 0.4283 = 0.0049				
student_mentee   Odds ratio Std. err. z interval]	P> z	[95%	conf.				
epoc_perser_pre   1.006256 .1447863 0.04 1.334089							

epoc_optimism_pu	re   .910149	2 .1326196	-0.65	0.518	.6840413			
epoc_connect_pi	re   .904529	4 .1285767	-0.71	0.480	.6845835			
· -	ns   1.02259							
Note: _cons est:		e odds.						
<pre>. logistic stude SSE_total_pre note: SSE_total_</pre>	_		-		SSE_aspira	tion_pre		
Logistic regress	sion			LR ch	r of obs = i2(3) =	5.89		
Log likelihood =	-130.86026				> chi2 = > R2 =			
student_me interval]	entee   Odds r				-			
SSE_productivity 1.521115	/_pre   1.07	9082 .1890	261 0	.43 0.664	.7655	03		
SSE_belong 1.236653	g_pre   .954	0516 .1262	905 -0	.36 0.722	.73603	06		
SSE_aspiration	n_pre   .815	3582 .0948	993 -1	.75 0.079	.64904	.89		
SSE_tota 1.366612	l_pre   _cons   .437	1 (omitte 0582 .2542	ed) 173 -1	.42 0.155	.13977	62		
Note: _cons est:	imates baselin	e odds.						
. logistic stude	ent_mentee hel	pseeking2_1	helpseekin	ng2_2 helps	eeking2_3			
Logistic regress	sion			LR ch	r of obs = i2(3) =	17.61		
Log likelihood =	Prob > chi2 = 0.0005 Log likelihood = -279.94048 Pseudo R2 = 0.0305							
student_mentee	<u>-</u>	Std. err.	z I	P> z  [	95% conf. i	nterval]		
helpseeking2_1	1.242168	.2234111	1.21	a.228	.873146	1.767152		
helpseeking2_2	1.503957	.2757762	2.23	0.026 1	.049912	2.154357		
helpseeking2_3 _cons	.9965004   .0706041				7040994 0239555			
Noto: sons :-1	·							

Note: \_cons estimates baseline odds.

. \*\* Include these as confounders in the propensity matching analyses. Only difficult life event and year remained after modelling . \*\*\* Creating the matched cohort . kmatch ps student\_mentee student\_difficultevent student\_year student\_language student\_disability\_epoc\_perser\_pre\_epoc\_optimism\_pre\_epoc\_connect\_pre helpseeking2\_1 hel > pseeking2 2 helpseeking2 3, att nn(1) caliper(0.001) generate( treated) Propensity-score nearest-neighbor matching Number of obs = 424Neighbors: min = Treatment : student mentee = 1 Covariates : student\_difficultevent student\_year student\_language student\_disability epoc\_perser\_pre epoc\_optimism\_pre epoc\_connect\_pre helpseeking2 1 ... PS model : logit (pr) Matching statistics Matched Controls Caliper Yes No Total Used Unused Total Treated | 52 96 148 | 45 231 276 .001 Stored variables Variable Storage Display Value type format label Variable label \_treated byte Treatment indicator %8.0g Number of matched controls \_KM\_nc byte %10.0g Number of times used as a match \_KM\_nm byte %10.0g double %10.0g Matching weight \_KM\_mw KM ps double %10.0g Propensity score \_KM\_strata byte %8.0g Matching stratum . kmatch sum Raw Matched(ATT) Means | Treated Untreated StdDif | Treated Untreated StdDif

	+-				+-		
<pre>student_difficultevent .0403433</pre>		.527027	.2826087	.5127537		.3269231	.3076923
student_year 1440615		2.216216	2.40942	3618324		2.269231	2.346154
student_language		.777027	.5362319	.3046688		.5961538	.6346154
<pre>0486639     student_disability .1263475</pre>	I	.1283784	.1594203	0679825		.1730769	.1153846
epoc_perser_pre		3.103604	3.192029	0951542		3.326923	3.370192
0465618 epoc_optimism_pre 2526691		2.927365	3.066425	1353395		3.086538	3.346154
epoc_connect_pre		3.689189	3.838164	1612765		3.903846	4.009615
114503 helpseeking2_1		3.344595	3.061594	.3750685		3.288462	3.423077
1784096 helpseeking2_2	1	3.256757	2.949275	.4095684		3.173077	3.269231
1280779 helpseeking2_3 1034123	I	3.378378	3.177536	.270004		3.307692	3.384615
	I		Raw	I		М	atched(ATT)
Variances Ratio	•	Treated	Untreated	·	•	Treated	Untreated
	•		Untreated	·	•	Treated	Untreated
Ratio student_difficultevent	-+-		Untreated		+-	Treated	Untreated
Ratio student_difficultevent 1.029835 student_year	· ·+- 		Untreated		+- 	Treated	Untreated
Ratio student_difficultevent 1.029835 student_year 1.374879 student_language	· -+-   	.2509653	Untreated	1.233376	+-   	Treated 	Untreated
Ratio   student_difficultevent  1.029835  student_year  1.374879  student_language  .9491059  student_disability	· -+-     	.2509653 .3202795	Untreated2034783 .2499473	1.233376   1.281388	+-   	Treated224359 .3182504	Untreated2178591 .2314753
Ratio	· -+-     	.2509653 .3202795 .6506251	Untreated	1.233376   1.281388   1.086762	+-	Treated224359 .3182504 .5984163	Untreated2178591 .2314753 .6305053
Ratio	· -+- 	.2509653 .3202795 .6506251 .1806858	.2034783 .2499473 .5986825 .2363109	1.233376   1.281388   1.086762   .7646104	+-     	Treated .224359 .3182504 .5984163 .2635747	Untreated .2178591 .2314753 .6305053 .1437265
Ratio	· +-	.2509653 .3202795 .6506251 .1806858 .865516	Untreated	1.233376   1.281388   1.086762   .7646104   1.004513	· +- 	Treated .224359 .3182504 .5984163 .2635747 .9351433	Untreated2178591 .2314753 .6305053 .1437265 .7067891
Ratio	· +-	.2509653 .3202795 .6506251 .1806858 .865516 1.090352	Untreated .2034783 .2499473 .5986825 .2363109 .8616271 1.021127	1.233376   1.281388   1.086762   .7646104   1.004513   1.067792	· +-               .	Treated .224359 .3182504 .5984163 .2635747 .9351433 1.477658	Untreated2178591 .2314753 .6305053 .1437265 .7067891 .9468927
Ratio   student_difficultevent  1.029835  student_year  1.374879  student_language .9491059  student_disability  1.833863  epoc_perser_pre  1.323087  epoc_optimism_pre  1.560534  epoc_connect_pre  1.027914	· +-	.2509653 .3202795 .6506251 .1806858 .865516 1.090352 .8452943	Untreated	1.233376   1.281388   1.086762   .7646104   1.004513   1.067792   .9814861	· +-               .	Treated	Untreated2178591 .2314753 .6305053 .1437265 .7067891 .9468927 .7276152
Ratio	. + -	.2509653 .3202795 .6506251 .1806858 .865516 1.090352 .8452943 .4042563	Untreated .2034783 .2499473 .5986825 .2363109 .8616271 1.021127 .8612392 .7343742	1.233376 1.281388 1.086762 .7646104 1.004513 1.067792 .9814861 .5504773	. +	Treated	Untreated2178591 .2314753 .6305053 .1437265 .7067891 .9468927 .7276152 .6036512

\_\_\_\_\_\_

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. label define mentee\_lab 0 "Non - Mentee" 1 "Mentee"

. label values \_treated mentee\_lab

. tab \_treated if complete==1

Treatment   indicator	Freq.	Percent	Cum.
Non - Mentee   Mentee	95 102	48.22 51.78	48.22 100.00
Total	197	100.00	

. tab \_treated student\_mentee, m

  Treatment		student a mentee?	
indicator	No	Yes	Total +
Non - Mentee	276	0	276
Mentee	0	148	148
.	321	27	348
Total	597	175	772

- . \*\*replace \_treated=1 if student\_mentee==1
- . tab \_treated student\_mentee, m

   Treatment		student a mentee?	
indicator	No	Yes	Total +
Non - Mentee	276	0	276
Mentee   .	9 321	148 27	148   348
+ Total	597	175	+   772

. gen \_notmatched=.
(772 missing values generated)

- . replace \_notmatched=1 if \_treated==. & student\_mentee==1
  (27 real changes made)
- . tab \_notmatched

_notmatched	Freq.	Percent	Cum.

```
1 | 27 100.00 100.00
    Total | 27 100.00
 *******
       Boxplots of covariates before and after matching
**************
. graph box epoc_perser_pre epoc_optimism_pre epoc_connect_pre helpseeking2_1
helpseeking2_2 helpseeking2_3, over(_treated) ///
```

> title("Covariate Balance Before and After Matching")

\*\*\*\*\*\*\*\*\*\*\*\*\*\* . \*\*\* Demographics for the matched cohort \*\*\*\*\*\*\*\*\*\*\*\*

foreach var of varlist student\_year student\_gender student\_australia student\_language student\_disability student\_live student\_support student\_absent student d

- > ifficultevent student impact {
  - tab student\_mentee `var', row chi 2.
  - 3.

+·    -	Key	+     
	frequency	
	row percentage	

Is this | student a İ

Raise mentee?	What year   Year 7	level are Year 8	•	Total
No	2	395	198	595
	0.34	66.39	33.28	100.00
Yes	11	103	46	160
	6.88	64.38	28.75	100.00
Total	13	498	244	755
	1.72	65.96	32.32	100.00

Pearson chi2(2) = 32.1882 Pr = 0.000

++   Key	
   frequency	
row percentage	

Is this student a Raise mentee?	     What   Male	Female		y as? Rather no	•
No	305	244	15	23	587
	51.96	41.57	2.56	3.92	100.00
Yes	66	82	8	3	159
	41.51	51.57	5.03	1.89	100.00
Total	371	326	23	26	746
	49.73	43.70	3.08	3 <b>.</b> 49	100.00

Pearson chi2(3) = 9.5822 Pr = 0.022

+	+- 
	- İ
frequency	ļ
row percentage	
<b>_</b>	_ 1

Is this student a Raise	   Were you   Austr	born in	
mentee?	No	Yes	Total
No	126	459	585
	21.54	78.46	100.00
Yes	36	123	159
	22.64	77.36	100.00
Total	162	582	744
	21.77	78.23	100.00

Pearson chi2(1) = 0.0893 Pr = 0.765

+
Key
frequency
row percentage
+

Is this				
student a	Do you sp	eak a langu	age other	
Raise	than	English at	home?	
mentee?	No No	Yes, some	Yes, most	Total
No	310	138	143	591
	52.45	23.35	24.20	100.00

				+
Yes	74	46	39	159
I	46.54	28.93	24.53	100.00
+				+
Total	384	184	182	750
	51.20	24.53	24.27	100.00

Pearson chi2(2) = 2.4516 Pr = 0.294

+		+
Key	•	
f	requency	
row	percentage	
+		+

Is this						
student a	Are yo	Are you a person with a				
Raise		disability?				
mentee?	No No	Yes	Prefer no	Total		
	+			+		
No	517	32	37	586		
	88.23	5.46	6.31	100.00		
	+			+		
Yes	140	9	7	156		
	89.74	5.77	4.49	100.00		
	+			+		
Total	657	41	44	742		
	88.54	5.53	5.93	100.00		

Pearson chi2(2) = 0.7467 Pr = 0.688

+	+
Key	
	1
frequency	
row percentage	
+	4

Is this student a Raise mentee?	-	Who do you Both pare		Other	Total
No	92	462	9	28	591
	15.57	78.17	1.52	4.74	100.00
Yes	52	92	4	10	158
	32.91	58.23	2.53	6.33	100.00
Total	144	554	13	38	749
	19.23	73.97	1.74	5.07	100.00

Pearson chi2(3) = 27.5660 Pr = 0.000

+	+
Key	
frequency	
row percentage	
+	+

+----+

Is this	Do you help look after any						
student a	family mem	family members or friends who					
Raise	need yo	ur support	becau				
mentee?	No	Yes	Prefer no	Total			
	<b></b>			+			
No	262	70	37	369			
	71.00	18.97	10.03	100.00			
	   100	40		+			
Yes	100	49	6	155			
	64.52	31.61	3.87	100.00			
Total	 l 362	119	 43	+   524			
IULai			_	:			
	69.08	22.71	8.21	100.00			

Pearson chi2(2) = 13.3880 Pr = 0.001

Key	!					
   frequen   row perce	cy					
+	+					
Is this student a Raise	į			In the r	nast two wee	ks, how many
	have you be	en absent fo	or?	In the p	dase ewo wee	ks, now many
mentee?	0	.5	1			3.5
4 Total	5	6	6.5	7	9	10
	+					
	1 202	2	67	25	17	1
7	203 12			35	17 3	1 1
365				_	-	•
		0.55				
1.92 100.00	3.29	1.37	0.27	0.82	0.82	0.27
	+					
	I 40	0	15	10	10	0
Yes 2	9 48	0 1	0	19 4	_	0 4
117	_	_	-	-	-	- 1
	41.03	0.00	12.82	16.24	8.55	0.00

100.00		0.85				3.42
Total 9 482	· 	2 6 0.41	82 1	54 7 11.20	27 3	1 5   0.21
for? mentee? 26	   In the   11 Total		13	14	15	been absent 25
No 0   0.00	365 0.00	2 0.55	1 0.27	5 1.37	0 0.00	
Yes 1   0.85	117   0.85	0.00	1 0.85	0 0.00	1 0.85	1
Total	1   482	2 0.41	2	5	1	1 0.21
0.21   100.00  Pearson chi2(19) = 45.5107 Pr = 0.001  ++    Key						

Is this student a	Have you exper   any difficult   personal event   mental healt   No	home or ts (e.g	Total
No	+   262	110	372

+----+

	70.43	29.57	100.00
Yes	74	82	156
	47.44	52.56	100.00
Total	336	192	528
	63.64	36.36	100.00

Pearson chi2(1) = 25.1130 Pr = 0.000

<del>+</del>	+
Key	ļ
	ļ
frequency	ļ
row percentage	
+	+

Pearson chi2(4) = 10.2159 Pr = 0.037

bysort \_treated: sum helpseeking2\_1 helpseeking2\_2 helpseeking2\_3
epoc\_engagement\_pre epoc\_perser\_pre epoc\_optimism\_pre epoc\_connect\_pre
helpseeking\_pre SSE\_a

> spiration\_pre SSE\_productivity\_pre SSE\_belong\_pre wemwbs brs\_pre

-----

\_\_\_\_\_

-> \_treated = Non - Mentee

Variable	Obs	Mean	Std. dev.	Min	Max
helpseekin~1	   276	3.061594	.8569563	1	4
helpseekin~2	276	2.949275	.8470051	1	4
helpseekin~3	276	3.177536	.8405641	1	4
epoc_enga~re	276	2.996075	.9619843	1	5
epoc_pers~re	276	3.192029	.9282387	1	5
	<b></b>				
epoc_opti~re	276	3.066425	1.010508	1	5
epoc_conn~re	276	3.838164	.9280297	1	5

helpseeki~re	242	33.16116	9.013295	10	64
SSE_aspir~re	276	8.203804	1.920465	1	10
SSE_produ~re	276	6.53688	1.861054	1	10
SSE_belon~re	276	5.936775	2.090089	1	10
wemwbs	262	21.76145	5.263501	7	35
brs_pre	273	3.122161	.6004375	1	5

-----

-----

## -> \_treated = Mentee

Variable	0bs	Mean	Std. dev.	Min	Max
helpseekin~1	148	3.344595	.6358115	2	4
helpseekin~2	148	3.256757	.6401703	2	4
helpseekin~3	148	3.378378	.6325137	2	4
epoc_enga~re	36	2.736111	1.012325	1.5	5
epoc_pers~re	148	3.103604	.9303311	1	5
+					
epoc_opti~re	148	2.927365	1.044199	1	5
epoc_conn~re	148	3.689189	.9193989	1	5
helpseeki~re	29	30.82759	9.562086	16	50
SSE_aspir~re	37	7.439189	1.893341	2.5	10
SSE_produ~re	37	6.055985	1.646507	2.75	8.833333
+					
SSE_belon~re	37	5.445946	2.037857	1.333333	9.166667
wemwbs	35	20.55629	5.725057	7	35
brs_pre	147	2.900907	.6246454	1.166667	5

-----

-----

<sup>-&</sup>gt; \_treated = .

Variable	0bs	Mean	Std. dev.	Min	Max
helpseekin~1	33	3.272727	.6261353	2	4
helpseekin~2	32	2.90625	.7770654	1	4
helpseekin~3	30	3.1	.7588558	2	4
epoc_enga~re	61	2.92623	.976115	1	5
epoc_pers~re	65	3.173077	.9792008	1	5
epoc_opti~re	13	2.961538	.8281893	1.75	4.5
epoc_conn~re	13	4.038462	.9119927	2.25	5
helpseeki~re	69	33.76812	10.02371	11	64
SSE_aspir~re	102	8.049837	2.005776	1	10
SSE_produ~re	103	6.13846	1.935571	1	10
+-					
SSE_belon~re	102	5.551634	2.115606	1	10
wemwbs	61	21.92066	4.83998	14.75	35
brs_pre	13	3.102564	.4978586	2	4.166667

```
foreach var of varlist helpseeking2_1 helpseeking2_2 helpseeking2_3
epoc_engagement_pre epoc_perser_pre epoc_optimism_pre epoc_connect_pre
helpseeking pre SSE
> _aspiration_pre SSE_productivity_pre SSE_belong_pre wemwbs brs_pre {
       ttest `var', by(_treated)
 3.
Two-sample t test with equal variances
______
 Group | Obs Mean Std. err. Std. dev. [95% conf. interval]
Non - Me | 276 3.061594 .0515827 .8569563 2.960047 3.163141
Mentee | 148 3.344595 .0522634 .6358115 3.24131 3.447879
------
Combined | 424 3.160377 .0387346 .7975944 3.084241 3.236514
 diff | -.2830004 .0801819
                                  -.440606 -.1253948
______
                             t = -3.5295
  diff = mean(Non - Me) - mean(Mentee)
H0: diff = 0
                             Degrees of freedom = 422
  Ha: diff < 0
                 Ha: diff != 0
                                     Ha: diff > 0
Pr(T < t) = 0.0002 Pr(|T| > |t|) = 0.0005 Pr(T > t) = 0.9998
Two-sample t test with equal variances
-----
Group | Obs Mean Std. err. Std. dev. [95% conf. interval]
Non - Me | 276 2.949275 .0509837 .8470051 2.848907 3.049643
Mentee | 148 3.256757 .0526217 .6401703 3.152764 3.360749
------
Combined | 424 3.056604 .0385577 .793951 2.980815 3.132392
-------
  diff | -.3074814 .0795899
                                  -.4639234 -.1510394
______
  diff = mean(Non - Me) - mean(Mentee)
                                  t = -3.8633
                            Degrees of freedom =
H0: diff = 0
                   Ha: diff != 0
  Ha: diff < 0
                                      Ha: diff > 0
Pr(T < t) = 0.0001 Pr(|T| > |t|) = 0.0001 Pr(T > t) = 0.9999
Two-sample t test with equal variances
______
 Group | Obs Mean Std. err. Std. dev. [95% conf. interval]
Non - Me | 276 3.177536 .050596 .8405641 3.077932 3.277141
Mentee | 148 3.378378 .0519923 .6325137 3.275629 3.481127
Combined | 424 3.247642 .037854 .779461 3.173236 3.322047
diff | -.2008421 .0789037 -.3559353 -.045749
diff = mean(Non - Me) - mean(Mentee)
                                       t = -2.5454
```

Two-sample	+	test	with	fauna	variances
I WO JUILDIC	_		WICII	Cuuai	vai Talices

Group	Obs	Mean	Std. err.		[95% conf.	_
Non - Me   Mentee	276 36	2.996075 2.736111	.0579047 .1687208	.9619843 1.012325	2.882082 2.39359	3.110068 3.078633
Combined	312	2.966079	.054905	.9698163	2.858047	3.074111
diff		.2599638	.1714971		0774817	
diff =	mean(Non -	 - Me) - mean	(Mentee)		 : †	= 1.5158

diff = mean(Non - Me) - mean(Mentee) t = 1.5158H0: diff = 0 Degrees of freedom = 310

#### Two-sample t test with equal variances

Group	0bs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
Non - Me   Mentee	276 148	3.192029 3.103604	.0558734 .0764727	.9282387 .9303311	3.082035 2.952476	3.302023 3.254732
Combined	424	3.161164	.0451079	.9288285	3.0725	3.249827
diff		.0884254	.0946451		0976092	.27446

diff = mean(Non - Me) - mean(Mentee) t = 0.9343 H0: diff = 0 Degrees of freedom = 422

#### Two-sample t test with equal variances

0bs	Group	Mean			[95% conf.	_
276 148	Non - Me   Mentee	3.066425 2.927365	.0608255 .0858326	1.010508 1.044199	2.946682 2.75774	3.186168 3.09699
424	Combined	3.017885	.0496966	1.023315	2.920202	3.115568
	diff	.1390603	.1041611		065679	.3437995
ean(Non	diff = n	- Me) - mean	(Mentee)		 t	= 1.3350

Ha: diff < 0 Ha: diff != 0 Ha: diff > 0

Ha: diff != 0

Pr(T < t) = 0.9883 Pr(|T| > |t|) = 0.0234 Pr(T > t) = 0.0117

Ha: diff > 0

Two-sample t test with equal variances

Ha: diff < 0

Group			Std. err.	Std. dev.	[95% conf. interval
Non - Me   Mentee	276 37	6.53688 6.055985	.1120223 .2706841		6.316349 6.7574 5.507012 6.60495
•	313	6.480033		1.841134	6.275271 6.68479
•		.4808951			152079 1.11386
diff = n H0: diff = 0	•	- Me) - mean	(Mentee)	Degrees	t = 1.494 s of freedom = 31
Ha: diff Pr(T < t) =	f < 0 = 0.9320	Pr(	Ha: diff != T  >  t ) =	0.1360	Ha: diff > 0 Pr(T > t) = 0.068
Two-sample t	test wi	th equal var	iances		
Group			Std. err.	Std. dev.	[95% conf. interval
Non - Me	276	5.936775 5.445946	.1258086 .3350216	2.090089 2.037857	5.689105 6.18444 4.766491 6.12540
Combined	313	5.878754	.1179536	2.086813	5.646669 6.11083
diff		.4908294	.3648694		2270954 1.20875
diff = n	nean(Non	 - Me) - mean	(Mentee)		t = 1.345
H0: diff = 0	-	,	`	Degrees	of freedom = 31
H0: diff = 0	)				Ha: diff > 0 Pr(T > t) = 0.089
H0: diff = 0	F < 0 = 0.9102	Pr(	Ha: diff != T  >  t ) =		
H0: diff = 0  Ha: diff Pr(T < t) =  Two-sample t  Group	f < 0 = 0.9102 = test wi 	Pr(  th equal var  Mean	Ha: diff != T  >  t ) = iances 	0 0.1795	
H0: diff = 0  Ha: diff Pr(T < t) =  Two-sample t Group   Non - Me   Mentee	f < 0 = 0.9102 t test wi  Obs  262 35	Pr(  th equal var  Mean  21.76145 20.55629	Ha: diff != T  >  t ) =  iances Std. err3251802 .9677113	5.263501 5.725057	Ha: diff > 0 Pr(T > t) = 0.089  [95% conf. interval 21.12114 22.4017 18.58966 22.5229
H0: diff = 0  Ha: diff Pr(T < t) =  Two-sample t  Group  + Non - Me   Mentee   Combined	6 < 0 = 0.9102 = test wirder Obs 	Pr(  th equal var Mean  21.76145 20.55629 	Ha: diff != T  >  t ) =  iances Std. err3251802 .96771133089295	5.323995	Ha: diff > 0 Pr(T > t) = 0.089  [95% conf. interval 21.12114 22.4017 18.58966 22.5229
H0: diff = 0  Ha: diff Pr(T < t) =  Two-sample t  Group    Non - Me   Mentee   Combined	6 < 0 = 0.9102 = test wi Obs 	Pr(  th equal var Mean 21.76145 20.55629 21.61943	Ha: diff != T  >  t ) =  iances Std. err3251802 .96771133089295	5.263501 5.725057 5.323995	Ha: diff > 0 Pr(T > t) = 0.089  [95% conf. interval 21.12114
H0: diff = 0  Ha: diff Pr(T < t) =  Two-sample t  Group    Non - Me   Mentee   Combined   diff	6 < 0 = 0.9102 = test wir Obs 	Pr(  th equal var Mean 21.76145 20.55629 21.61943	Ha: diff != T  >  t ) =  iances Std. err3251802 .96771133089295	5.263501 5.725057 5.323995	Ha: diff > 0 Pr(T > t) = 0.089  [95% conf. interval 21.12114 22.4017 18.58966 22.5229
H0: diff = 0  Ha: diff Pr(T < t) =  Two-sample t  Group	6 < 0 = 0.9102 = test wi Obs 	Pr(  th equal var Mean 21.76145 20.55629 21.61943 1.205165	Ha: diff != T  >  t ) =  iances Std. err3251802 .967711330892959571991 (Mentee)	5.263501 5.725057 5.323995	Ha: diff > 0 Pr(T > t) = 0.089  [95% conf. interval 21.12114 22.4017 18.58966 22.5229 21.01145 22.2276786396 3.08896
H0: diff = 0  Ha: diff Pr(T < t) =  Two-sample t  Group	0	Pr(  th equal var Mean 21.76145 20.55629 21.61943 1.205165  Me) - mean	Ha: diff != T  >  t ) =  iances Std. err3251802 .967711330892959571991 (Mentee)  Ha: diff != T  >  t ) =  iances	5.263501 5.725057 5.323995 Degrees	Ha: diff > 0 Pr(T > t) = 0.089  [95% conf. interval] 21.12114
H0: diff = 6  Ha: diff Pr(T < t) =  Two-sample t  Group	6 < 0 = 0.9102	Pr(  th equal var	Ha: diff != T  >  t ) =  iances Std. err3251802 .96771133089295 (Mentee)  Ha: diff != T  >  t ) =  iances Std. err.	5.0 0.1795 Std. dev. 5.263501 5.725057 5.323995 Degrees 0 0.2090	Ha: diff > 0 Pr(T > t) = 0.089  [95% conf. interval 21.12114

```
Non - Me | 273 3.122161 .0363401 .6004375 3.050617 3.193705
Mentee | 147 2.900907 .0515199 .6246454 2.799086 3.002728
Combined | 420 3.044722 .0301252 .6173832 2.985507 3.103938
.0987894 .3437189
          .2212541 .0623022
-----
   diff = mean(Non - Me) - mean(Mentee)
                                                 t = 3.5513
H0: diff = 0
                                     Degrees of freedom = 418
   Ha: diff < 0
                        Ha: diff != 0
                                              Ha: diff > 0
Ha: diff < 0 Ha: diff != 0 Ha: diff > 0 Pr(T < t) = 0.9998 Pr(|T| > |t|) = 0.0004 Pr(T > t) = 0.0002
**************
. *** Demographics for those who could not be matched
       foreach var of varlist student_year student_gender student_australia
student_language student_disability student_live student_support student_absent
student d
> ifficultevent student impact {
         tab student_mentee if _notmatched==1
 3.
   Is this |
 student a
    Raise |
  mentee? | Freq. Percent Cum.
-----
     Yes | 27 100.00 100.00
    Total | 27 100.00
   Is this |
 student a l
    Raise
   mentee? |
            Freq. Percent Cum.
     Yes | 27 100.00 100.00
               Total | 27 100.00
   Is this |
 student a |
    Raise
   mentee? | Freq. Percent Cum.
     Yes | 27 100.00 100.00
               27
    Total |
                     100.00
   Is this |
 student a
```

Raise

mentee?	Freq.	Percent	Cum.	
Yes		100.00		
Total	27			
Is this   student a   Raise   mentee?		Doncont	Cum	
+		Percent		
Yes		100.00		
Total	27	100.00		
Is this   student a   Raise   mentee?		Percent	Cum.	
+ Yes	27	100.00	100.00	
+ Total	 27	100.00		
Is this   student a   Raise   mentee?		Percent	Cum.	
+ Yes		100.00	100.00	
   Total	 27	100.00		
Is this   student a   Raise		Percent	Cum.	
		100.00	100.00	
Total	27	100.00		
Is this   student a   Raise   mentee?	-	Percent		
Yes	27	100.00	100.00	
Total		100.00		
Is this   student a   Raise				

Cum.	Percent	•	mentee?
100.00		27	Yes
	100.00		Total

> e SSE\_belong\_pre wemwbs brs\_pre if \_notmatched==1

Variable	Obs .	Mean	Std. dev.	Min	Max
helpseekin~1	   5	3.4	.5477226	3	4
helpseekin~2	j 4	3.25	.9574271	2	4
helpseekin~3	3	3	1	2	4
epoc_enga~re	4	2.375	.595119	2	3.25
epoc_pers~re	8	2.59375	.8957987	1.75	4.25
	+				
epoc_opti~re	5	2.6	.9117291	1.75	4
epoc_conn~re	5	4.1	.9117291	3	5
SSE_aspir~re	4	6.6875	2.003902	5.25	9.5
SSE_produ~re	4	4.958333	1.119069	3.666667	6.25
SSE_belon~re	4	4.191667	1.147743	2.6	5.333333
	+				
wemwbs	4	17.325	2.636848	14.75	20.73
brs_pre	5	2.8	.4624812	2	3.166667

### . log close

name: <unnamed>

log: C:\Users\quachjl\OneDrive - The University of Melbourne\Documents\1\_Projects\1\_InProgress\2022\_RAISE Evaluation\10\_Data Analysis\ALL\Stata\02\_define\_match\_

> 26 Mar 2024.log log type: text

closed on: 26 Mar 2024, 20:31:41

sum helpseeking2\_1 helpseeking2\_2 helpseeking2\_3 epoc\_engagement\_pre epoc\_perser\_pre epoc\_optimism\_pre epoc\_connect\_pre SSE\_aspiration\_pre SSE\_productivity\_pr

-----

name: <unnamed>

log: C:\Users\quachjl\OneDrive - The University of

 $\label{locality} {\tt Melbourne\Documents\1\_Projects\1\_InProgress\2022\_RAISE\ Evaluation\10\_Data}$ 

Analysis\ALL\Stata\03\_matched\_analy

> sis\_26 Mar 2024.log
log type: text

opened on: 26 Mar 2024, 20:31:41

•

. \*\* Conducting the analyses

\*\*\*\*\*\*\*\*\*\*\*

•

. \*\*\* All outcomes examined

. sum epoc\_engagement\_change epoc\_perser\_change epoc\_optimism\_change epoc\_connect\_change epoc\_happy\_change

Variable	0bs	Mean	Std. dev.	Min	Max
epoc_enga~ge	125	.1813333	.9163651	-2.25	3.5
epoc_pers~ge	214	.1674455	.8044684	-2.5	3.25
epoc_opti~ge	182	.2408425	.7633557	-1.5	2.75
epoc_conn~ge	182	.1694139	.7203256	-2.25	2.5
epoc_happ~ge	125	.1226667	.7671702	-2.25	3

. sum helpseeking2\_1\_change helpseeking2\_2\_change helpseeking2\_3\_change

Variable	0bs	Mean	Std. dev.	Min	Max
hel~1_change	209	.2296651	.8575908	-2	3
hel~2_change	208	.2355769	.8610092	-2	2
hel~3_change	208	.3028846	.8395946	-2	2

. sum belonging\_change

Variable	0bs	Mean	Std. dev.	Min	Max
+					
belonging~ge	192 .	1012153	.6496153	-2.333333	2.142857

. sum SSE\_aspiration\_change SSE\_belong\_change SSE\_productivity\_change SSE\_total\_change

Variable	0bs	Mean	Std. dev.	Min	Max
SSE_aspir~ge	137	-1.354015	1.709641	-8	3
SSE_belon~ge	137	.1493917	1.542948	-4.166667	4.333333
SSE produ~ge	137	0917441	1.426916	-5.416667	5

. sum brs\_change wembs\_change

Variable			Std. dev.		Max
brs_change					
wembs_change	115	.2803478	4.439234	-16.41	12.11

. sum z\_epoc\_engagement\_change z\_epoc\_perser\_change z\_epoc\_optimism\_change z\_epoc\_connect\_change z\_helpseeking2\_1\_change z\_helpseeking2\_2\_change z\_helpseeking2\_3\_change

> z\_belonging\_change z\_SSE\_total\_change z\_brs\_change z\_wembs\_change

Variable	0bs	Mean	Std.	dev.	Min	Max
z_epoc_en~ge   z_epoc_pe~ge   z_epoc_op~ge   z_epoc_co~ge   z_h~1_change	125 214 182 182 209	4.89e-09 -5.29e-09 5.49e-09 5.32e-10 -8.27e-09		1 1 1 1	-2.653237 -3.315786 -2.280513 -3.358778 -2.599917	3.621555 3.83179 3.28701 3.235462 3.230369
z_h~2_change   z_h~3_change   z_belongi~ge   z_SSE_tot~ge   z_brs_change	208 208 192 137 181	6.02e-09 -5.59e-09 2.47e-09 -3.40e-10 6.40e-09		1 1 1 1 1	-2.596461 -2.742853 -3.747677 -3.795455 -3.075493	2.04925 2.021351 3.142848 3.062357 2.952776
z_wembs_ch~e	115	-3.24e-11		1	-3.759736	2.664796

. \* sum z\_epoc\_engagement\_change z\_epoc\_perser\_change z\_epoc\_optimism\_change z\_epoc\_connect\_change z\_belonging\_change z\_belonging\_change z\_SSE\_total\_change z\_brs\_change

> z\_wembs\_change

•

\*\*\*\*\*\*\*\*\*

. \*\*\*\* How many in matched cohort

. tab student\_mentee

Is this |
student a |
Raise |
mentee? | Freq. Percent Cum.
No | 597 77.33 77.33
Yes | 175 22.67 100.00

```
Total |
             772 100.00
. drop if _treated==.
(348 observations deleted)
. count
 424
codebook student_mentee
student_mentee
                                                      Is this student a Raise
mentee?
                 Type: Numeric (byte)
                Label: student_mentee_
                Range: [0,1]
                                                    Units: 1
        Unique values: 2
                                                Missing .: 0/424
           Tabulation: Freq. Numeric Label
                         276
                                0 No
                                 1 Yes
                         148
. rename student_gender gender
. gen student_gender=gender
(2 missing values generated)
. recode student_gender 1=0 2=1 3=0 4=0
(422 changes made to student_gender)
. save "raise_matchedclean.dta", replace
file raise_matchedclean.dta saved
. **** Change scores
. sum epoc_engagement_change epoc_perser_change epoc_optimism_change
epoc_connect_change helpseeking2_1_change helpseeking2_2_change
helpseeking2_3_change belonging_cha
> nge SSE_aspiration_change SSE_belong_change SSE_productivity_change
SSE_total_change brs_change wembs_change
                                       Std. dev.
   Variable |
                     0bs
                                Mean
                                                       Min
```

Max

epoc_enga~ge	112	.1443452	.9086795	-2.25	3.5
epoc_pers~ge	197	.1298646	.7616857	-2.5	2
epoc_opti~ge	178	.2223783	.756937	-1.5	2.75
epoc_conn~ge	178	.170412	.7276606	-2.25	2.5
hel~1_change	200	.235	.8622601	-2	3
	+				
hel~2_change	199	.2512563	.84518	-2	2
hel~3_change	199	.3115578	.8429249	-2	2
belonging~ge	183	.1076243	.6553943	-2.333333	2.142857
SSE_aspir~ge	118	-1.32839	1.751112	-8	3
SSE_belon~ge	118	.1019774	1.571846	-4.166667	4.333333
	+				
SSE_produ~ge	118	1415062	1.44602	-5.416667	5
SSE_total~ge	118	4559729	1.331939	-5.305556	3.5
brs_change	177	.0422787	.7221972	-2.166667	2.166667
wembs_change	103	.3671845	4.329821	-16.41	12.11

\*\*\*\*\*\*\*\*\*

-----

<sup>-&</sup>gt; \_treated = Non - Mentee

Variable	0bs	Mean	Std. dev.	Min	Max
epoc_enga~ge	95	.1087719	.8917531	-1.5	3.5
epoc_pers~ge	95	.15	.7140311	-1.5	2
epoc_opti~ge	83	.184739	.7331033	-1.5	2.75
epoc_conn~ge	83	.0502008	.7037381	-1.75	2.25
hel~1_change	99	.0606061	.8429562	-2	3
	<b></b>				
hel~2_change	99	.0909091	.8217814	-2	2
hel~3_change	98	.1938776	.8453409	-2	2
belonging~ge	88	.0332792	.6604462	-2.333333	2.142857
SSE_aspir~ge	101	-1.316832	1.670829	-8	3
SSE_belon~ge	101	.2640264	1.50156	-4.166667	4.333333
SSE_produ~ge	101	.0399715	1.315767	-5.416667	5
SSE_total~ge	101	3376113	1.237613	-5.305556	3.5
brs_change	83	0251004	.5150208	-1.166667	1.333333
wembs_change	87	.5434483	3.998817	-10.06	12.11

-----

<sup>. \*\*</sup> Looking at change scores overall

<sup>\*\*\*\*\*\*\*\*\*\*\*\*\*\*\*\*\*</sup> 

<sup>.</sup> bysort \_treated: sum epoc\_engagement\_change epoc\_perser\_change
epoc\_optimism\_change epoc\_connect\_change helpseeking2\_1\_change
helpseeking2\_2\_change helpseeking2\_3\_cha

<sup>&</sup>gt; nge belonging\_change SSE\_aspiration\_change SSE\_belong\_change
SSE\_productivity\_change SSE\_total\_change brs\_change wembs\_change

-----

-> \_treated = Mentee

Variable	Obs .	Mean	Std. dev.	Min	Max
epoc_enga~ge	17	.3431373	1.003415	-2.25	2
epoc_pers~ge	102	.1111111	.8066123	-2.5	1.75
epoc_opti~ge	95	.2552632	.7795414	-1.5	2
epoc_conn~ge	95	.2754386	.7355858	-2.25	2.5
hel~1_change	101	.4059406	.8506259	-2	2
	+				
hel~2_change	100	.41	.8420154	-2	2
hel~3_change	101	.4257426	.8288128	-2	2
belonging~ge	95	.1764912	.6465225	-1.666667	1.5
SSE_aspir~ge	17	-1.397059	2.229172	-5	2.25
SSE_belon~ge	17	8607843	1.680659	-3.833333	2.266666
SSE_produ~ge	   17	-1.219697	1.739806	-4 <b>.</b> 583333	.8333335
SSE total~ge	17	-1.15918	1.669116	-4.138889	.8611107
brs_change	94	.1017731	.8634537	-2.166667	2.166667
wembs_change	16	5912498	5.885818	-16.41	8.040001

```
. foreach var of varlist epoc_engagement_change epoc_perser_change
epoc_optimism_change epoc_connect_change helpseeking2_1_change
helpseeking2 2 change helpseeking2 3 c
> hange belonging_change SSE_aspiration_change SSE_belong_change
SSE_productivity_change SSE_total_change brs_change wembs_change {
  2.
          bootstrap, reps(500) seed(7582) nodots: regress `var' _treated,
cluster(school)
. *** Outcomes adjusting for student gender and difficult live event
          bootstrap, reps(500) seed(7582) nodots: regress `var' _treated
student_gender student_difficultevent, cluster(school)
 4.
. *** Interaction between student gender and mentoring
          bootstrap, reps(500) seed(7582) nodots: regress `var'
_treated##student_gender, cluster(school)
. *** Interaction between difficult life event and mentoring
         bootstrap, reps(500) seed(7582) nodots: regress `var'
treated##student difficultevent, cluster(school)
```

Linear regression

```
Number of obs = 112
Replications = 500
Wald chi2(1) = 0.74
Prob > chi2 = 0.3893
R-squared = 0.0086
Adj R-squared = -0.0004
Root MSE = 0.9088
```

(Replications	based	on	8	clusters	in	school)	
<b>,</b> -F						/	

	Observed	Bootstrap			Noon	al-based
epoc_enga~ge	coefficien			P> z		
_treated	.2343653 .1087719	.2722129 .104751	1.04	0.299	0965362	.3140801
Linear regress	sion		(Po	nlicatio	Replication Wald chi2(3 Prob > chi2 R-squared Adj R-squar Root MSE	bbs = 112 s = 500 s) = 3.18 s = 0.3641 = 0.0188 ed = -0.0085 = 0.9125
school)			( ne			8 clusters in
Normal-based	1	Observed	Bootstrap			
<pre>epoc_engagemer interval]</pre>	_	coefficient				_
	·					
.7872113	_treated	.1553808	.3223684	0.48	3 0.630	4764497
	nt_gender	.1394512	.1977703	0.71	0.481	2481714
student_diffic	cultevent	.1125428	.12246	0.92	2 0.358	1274744
.2541246	_cons	.0164986	.12124	0.14	1 0.892	2211275
Linear regress	sion				R-squared	s = 488 s) = 4.63 d = 0.2006 = 0.0252 red = -0.0019
in school)			(Ro	eplicati	ions based o	on 8 clusters
Noomel b	I	0bserved	Bootstra	р		
Normal-based epoc_engageme interval]	ent_change	coefficient	std. err	. Z	z P> z	[95% conf.

	•				
_treated	•	920250		0 926	1 90025
Mentee 1.445271	1819892	.830250	-0.22	0.826	-1.80925
1.student_gender .4338623	.0771017	.182024	0.42	0.672	279659
_treated#student_gender Mentee#1 2.293841	   .5506761	.889386	3 0.62	0.536	-1.192489
2.293041	1				
_cons .2473834	•		3 0.97		
Note: One or more paramet					
Linear regression			R W P R A	rob > chi -squared	$\begin{array}{rcl} \text{ons} &=& 500 \\ \text{(3)} &=& 3.46 \\ \text{(2)} &=& 0.3254 \\ &=& 0.0310 \\ \text{ored} &=& 0.0041 \end{array}$
			(Rep	lications	based on 8
clusters in school)					
		Observed	Bootstrap		
Normal-based epoc_engagement_ [95% conf. interval]	_	efficient	std. err.	Z	P> z
= =					
<b>—</b>	treated   Mentee   -	.2227011	.435966	-0.51	0.609
student_difficu	ltevent   Yes	.028425	.0894595	0.32	0.751
1469124 .2037624	1				
_treated#student_difficu	:	.6950599	.4680111	1.49	0.138
222225 1.612345	1				
0910223 .2864246	_cons	.0977011	.0962892	1.01	0.310

------

Number of obs = 197Replications = 499Wald chi2(1) = 0.08Linear regression Prob > chi2 = 0.7790R-squared = 0.0007 Adj R-squared = -0.0045Root MSE = 0.7634(Replications based on 13 clusters in school) \_\_\_\_\_\_ Observed Bootstrap Normal-based epoc\_pers~ge | coefficient std. err. z P>|z| [95% conf. interval] \_treated | -.0388889 .1385915 -0.28 0.779 -.3105232 .3238042 Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications. Number of obs = 197Linear regression Replications = Wald chi2(3) = 2.39Prob > chi2 = 0.4946R-squared = 0.0116 Adj R-squared = -0.0037Root MSE = 0.7631(Replications based on 13 clusters in school)

\_\_\_\_\_\_ | Observed Bootstrap epoc\_perser\_change | coefficient std. err. z P>|z| [95% conf. interval] -----\_treated | -.0708086 .1600951 -0.44 0.658 -.3845892 student\_gender | .0731195 .1049018 0.70 0.486 -.1324842 .2787233 student\_difficultevent | .1359355 .1019035 1.33 0.182 -.0637918 .3356628 \_cons | .0716573 .0688934 1.04 0.298 -.0633714 .2066859

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

197 Linear regression Number of obs = Replications = 499 Wald chi2(3) = 0.74

Prob > chi2 = 0.8642

R-squared = 0.0045 Adj R-squared = -0.0110 Root MSE = 0.7658

(Replications based on 13 clusters

in school)		(Repl	lications	based on	13 clusters
Normal-based	Observed	Bootstrap			
epoc_perser_change	coefficient	std. err.	7	P> z	[95% conf.
interval]					_
	+				
	1				
_treated Mentee	  0253813	1509091	-0.17	0.866	3211577
.270395	1 .0233013	.1303031	0.17	0.000	. 3211377
1.student_gender	.1300098	.1694673	0.77	0.443	20214
.4621595					
_treated#student_gender					
Mentee#1	- E	.1950849	-0.39	0.700	4576457
.307073	•				
	101000	0=0=4=0			02=2004
_cons .2470679	.1048387	.0/256/2	1.44	0.149	03/3904
.2470075					
standard-error est Linear regression	imates include	only compl	Num Rep Wal Pro R-s	nber of ol plication: .d chi2(3 ob > chi2 quared	bs = 197 s = 499 ) = 2.99 = 0.3931 = 0.0116
				i R-squar ot MSE	ed = -0.0038 = 0.7631
			NOC	ic MSL	- 0.7031
clusters in school)			(Repli	cations	based on 13
	l 01-	D			
Normal-based	l Op	served Bo	ootstrap		
	_change   coef	ficient st	td. err.	Z	P> z
[95% conf. interval]	-				
_	treated				
	Mentee   .0	072402 .2	2119066	0.03	0.973
408089 .4225695	ı				
student_difficu	ltevent				

0458461	.4841033	Yes	.2	191286	.1351937	7 1.62	0.105
_treated#stud	ent_difficul Mente .311726		1	406972	.2308324	4 -0.61	0.542
1109563	. 2402666	_cons	.0	646552	.0895993	3 0.72	0.471
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.							
Linear regres	sion		(Ran]	ications	hased or	Wald chi2(1 Prob > chi2 R-squared Adj R-squar Root MSE	s = 499 ) = 0.16 = 0.6859 = 0.0022 ed = -0.0035
	   Observed						al-based
epoc_opti~ge	•		•		P> z		
	.0705242 .184739						.4122529 .4320184
Note: One or i	more paramet d-error esti						
Linear regres	sion			(Pa		Prob > chi R-squared Adj R-squa Root MSE	ns = 499 3) = 7.26 2 = 0.0640 = 0.0298 red = 0.0131 = 0.7520
school)				(ке	plication	is based on	13 clusters in
Normal-based epoc_optimi	 sm_change	Observ coeffici	ed ent		. z		[95% conf.
.3876596	_treated					1 0.970	3728642
	nt_gender	.21913	46	.1190902	1.84	1 0.066	0142778

.452547 student_difficultevent   .341099cons   .3026959	.1196739 .061791				1017512 1791139
Note: One or more parame standard-error est					
Linear regression  in school)		(Repl	R W P R A	Replication Prob > chi Resquared Resquared Resquared Resquared	obs = 178 ons = 499 (3) = 9.38 .2 = 0.0247 = 0.0360 ared = 0.0194 = 0.7496
Normal-based epoc_optimism_change interval]	coefficient				_
	+				
_treated	   .1700758	.1870564	0.91	0.363	196548
1.student_gender .7329369	.4218615	.1587149	2.66	0.008	.110786
_treated#student_gender Mentee#1 .0166343		.1860402	-1.87	0.061	71263
_cons .3131219	.0424242	.1381136	0.31	0.759	2282734
Note: One or more parame standard-error est					
Linear regression			R W F R	lald chi2( Prob > chi R-squared	$\begin{array}{rcl} \text{ons} & = & 499 \\ \text{(3)} & = & 13.45 \\ \text{(2)} & = & 0.0038 \\ & = & 0.0212 \\ \text{ored} & = & 0.0043 \end{array}$

(Replications based on 13

		1	0bserved	Bootstrap	1	
[95% conf. in	oc_optimism_cl terval]					
		+				
2511354	_ Mei	eated   ntee	.2025935	.2314986	0.88	0.381
	ent_difficult	event   Yes	.3098239	.0893996	3.47	0.001
.134604 .48	850439	ı				
_treated#stud	Mentee		3202406	.2186856	-1.46	0.143
7488565	.1083753	I				
234843 .:	-	_cons	.0578231	.1493222	0.39	0.699
Linear regres	sion				Replicatio Wald chi2(	1) = 4.53
			<b></b>		R-squared Adj R-squa Root MSE	red = 0.0184 = 0.7209
			(Replications	s based on	R-squared Adj R-squa Root MSE 13 cluster	= 0.0240 red = 0.0184 = 0.7209 s in school)
epoc_conn~ge	Observed   coefficient	Boots std.	trap err, z	s based on  P> z	R-squared Adj R-squa Root MSE  13 cluster Norm [95% con	= 0.0240 red = 0.0184 = 0.7209  s in school) al-based f. interval]
epoc_conn~ge  _treated	l Ohserved	Boots std.	trap err. z 836 2.13	s based on  P> z   0.033	R-squared Adj R-squa Root MSE  13 cluster Norm [95% con0178031	= 0.0240 red = 0.0184 = 0.7209  s in school) al-based f. interval]4326725
epoc_conn~getreatedconsNote: One or i	Observed   coefficient +   .2252378   .0502008	Boots std. .105 .0919	trap err. z 	s based on   P> z    0.033   0.585   imated in 1	R-squared Adj R-squa Root MSE  13 cluster Norm [95% con01780311299425 bootstrap	= 0.0240 red = 0.0184 = 0.7209  s in school) al-based f. interval]4326725 .2303441 replicate;

school)					13 clusters in
Normal-based epoc_connect_change   interval]	Observed coefficient	Bootstrap	Z	P> z	[95% conf.
.4090168 student_gender .2931982 student_difficultevent .2635677		.1117049 .0891702 .0952245	1.70 1.33 0.81	0.089 0.184 0.419	
.1482931 Note: One or more parame standard-error est	eters could no	t be estimat	ted in 1 lete repl	bootstrap	o replicate;
Linear regression			R W F R	Replication Wald chi2 Prob > chi R-squared	ons = 499 (3) = 6.77 i2 = 0.0798 = 0.0313 ared = 0.0146
in school)		(Repl	lications	based or	n 13 clusters
Normal-based epoc_connect_change interval]	coefficient				-
	167803		1.08	0.282	1381108
_treated#student_gender Mentee#1 .401493	.0614719	.1734833	0.35	0.723	2785492

\_cons |

.2212002

.019697

.1028097 0.19 0.848 -.1818063

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications. Linear regression Number of obs = 178 Replications = Wald chi2(3) = 28.54Prob > chi2 = 0.0000R-squared = 0.0417 Adj R-squared = 0.0252Root MSE = 0.7184(Replications based on 13 clusters in school) Observed Bootstrap Normal-based epoc\_connect\_change | coefficient std. err. z P>|z| [95% conf. interval] \_treated | Mentee .3754252 .1329186 2.82 0.005 .1149095 .6359409 student\_difficultevent | .2761605 .0604873 Yes 4.57 0.000 .1576076 .3947134 \_treated#student\_difficultevent | Mentee#Yes | -.3510718 .1521102 -.6492022 -.0529414 cons -.0629252 .0716884 -0.88 0.380 -.2034318 .0775815 Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications. Linear regression Number of obs = 200 Replications = 499 Wald chi2(1) = 6.19Prob > chi2 = 0.0129R-squared = 0.0403Adj R-squared = 0.0355Root MSE = 0.8468 (Replications based on 13 clusters in school) Normal-based Observed Bootstrap  $hel\sim1_{change} \mid coefficient std. err. z P>|z|$  [95% conf. interval]

```
______
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
    standard-error estimates include only complete replications.
Linear regression
                                    Number of obs =
                                                200
                                    Replications =
                                                499
                                    Wald chi2(3) = 7.43
                                    Prob > chi2 = 0.0595
                                    R-squared = 0.0416
                                    Adj R-squared = 0.0269
                                    Root MSE = 0.8506
                            (Replications based on 13 clusters in
school)
   .....
               Observed
                        Bootstrap
Normal-based
helpseeking2_1_change | coefficient std. err. z > |z| [95% conf.
interval]
-----
         _treated | .3387029 .1483499 2.28 0.022 .0479424
.6294634
     student_gender | -.0070076 .1187719 -0.06 0.953 -.2397963
.2257811
student_difficultevent | .0639368 .1362832
                                0.47 0.639 -.2031734
.331047
           .253922
______
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
    standard-error estimates include only complete replications.
                                    Number of obs =
Linear regression
                                                200
                                    Replications =
                                                499
                                    Wald chi2(3) = 7.89
                                    Prob > chi2 = 0.0484
                                    R-squared = 0.0427
                                    Adj R-squared = 0.0280
                                    Root MSE = 0.8501
                             (Replications based on 13 clusters
in school)
-----
                  Observed Bootstrap
Normal-based
 helpseeking2_1_change | coefficient std. err. z > |z| [95% conf.
interval]
```

	•				
_treated Mentee	.2662	338 .15657	24 1.70	0.089	0406425
.57311 1.student_gender .2973632	0909	091 .19810	17 -0.46	0.646	4791814
_treated#student_gender Mentee#1	     .1744	442 .27994	72 0.62	0.533	3742421
.7231305					
_cons .2043692	.0909				022551
Note: One or more parameters standard-error estimates time ar regression			omplete repi	lications.	
			\   	Wald chi2( Prob > chi R-squared	$\begin{array}{rcl} \cos s & = & 499 \\ (3) & = & 8.02 \\ 2 & = & 0.0456 \\ & = & 0.0457 \\ \text{ared} & = & 0.0311 \\ & = & 0.8488 \end{array}$
					based on 13
Normal-based	I	Observed	Bootstrap		
helpseeking2_1_ [95% conf. interval]	- 0 .				P> z
	+				
<del>-</del>	treated   Mentee	.2393548	.15033	1.59	0.111
student_difficu	   Itevent   Yes	0536181	.1487743	-0.36	0.719
3452104 .2379741	1				
	ltevent   ee#Yes	.2238142	.1585258	1.41	0.158
0868906 .5345191	cons	.0806452	.1238129	0.65	0.515
1620237 .3233141					

Note: One or more parameters could not be estimated in 1 bootstrap replicate;

standard-error estimates include only complete replications.

Linear regression

Number of obs = 199
Replications = 499
Wald chi2(1) = 5.18
Prob > chi2 = 0.0228
R-squared = 0.0358
Adj R-squared = 0.0309
Root MSE = 0.8320

## (Replications based on 13 clusters in school)

hel~2_change	std. err.		P> z	Normal [95% conf.	
_treated	.1401629	2.28	0.023	.0443766	.5938052
_cons	.0843516	1.08	0.281	074417	.2562352

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 199
Replications = 499
Wald chi2(3) = 12.75
Prob > chi2 = 0.0052
R-squared = 0.0449
Adj R-squared = 0.0302
Root MSE = 0.8323

(Replications based on 13 clusters in

s	c	h	o	o	1	)

	Observed	Bootstrap			
Normal-based helpseeking2_2_change interval]			z	P> z	[95% conf.
	T				
_treated	.2745818	.1554207	1.77	0.077	0300371
.5792008					
student gender	1480357	.1301637	1.14	0.255	1070804
.4031519	1 12:00337	. 1301037		0.233	.20,000
	1 0506633	4466004	0 50	0 645	4600405
student_difficultevent	.0586632	.1166224	0.50	0.615	1699125
.2872389					
cons	.0196392	.0996582	0.20	0.844	1756872
.2149657	•				
. 2177071					

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 199 Replications = 499

Wald chi2(3) = 11.01 Prob > chi2 = 0.0117 R-squared = 0.0474 Adj R-squared = 0.0327 Root MSE = 0.8312

(	Replications	based	on	13	clusters
١	INCPETCUCEOUS	Duscu	011		CIGOCCI

		(Repl	ications	based on	13 clusters
in school)					
Name 1 has a	Observed	Bootstrap			
Normal-based helpseeking2_2_change interval]	•				_
	+				
_treated					
	.1861472	.2164618	0.86	0.390	2381102
.6104046		2206704	0.10	0.050	4242070
1.student_gender .5152169	.0454545 	.2396/91	0.19	0.850	4243078
_treated#student_gender	İ				
Mentee#1	.2098821	.3383048	0.62	0.535	4531831
.8729473	ı				
_cons	   0757576	.0767315	0 99	0 323	- 07/633/
.2261485	.0737370	.0707313	0.55	0.525	0740334
standard-error est Linear regression  clusters in school)			Ni Ri Wi Pi R Ai	umber of eplicatio ald chi2( rob > chi -squared dj R-squa oot MSE	obs = 199 ns = 499 3) = 5.44 2 = 0.1424 = 0.0386 red = 0.0238 = 0.8350 based on 13
Normal-based helpseeking2_2 [95% conf. interval]	•	ficient sto			
	treated	662936 .1			

	ent_difficulte .3887948	event   Yes	.0274629	.1843564	0.15	0.882		
_treated#stude	ent_difficulte Mentee# .5209243	:	.0961865	.216707	0.44	0.657		
1024278	.2637181	cons	.0806452	.0934063	0.86	0.388		
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.								
Linear regres	sion	(Re	nlications	: hasad on	Wald chi2( Prob > chi R-squared Adj R-squa Root MSE	$\begin{array}{rcl} \cos & = & 499 \\ (1) & = & 2.26 \\ (2) & = & 0.1330 \\ & = & 0.0190 \\ \sin & = & 0.0140 \end{array}$		
	   Observed	Bootstra				nal-based		
hel~3_change	coefficient	std. err	z	P> z	[95% cor	nf. interval]		
_treated _cons			1.50 1.57			.5343706 .4360208		
	more parameter d-error estima							
Linear regres	sion				Wald chi2( Prob > chi R-squared	$\begin{array}{rcl} \cos s & = & 499 \\ (3) & = & 5.69 \\ (2) & = & 0.1275 \\ & = & 0.0287 \\ (3) & = & 0.0138 \end{array}$		
school)			(Re	plication	s based on	13 clusters in		
		Observed	Bootstra	р				
interval]	_3_change   cc					_		
	_treated					1177359		

.5333928		424000		0.004	0==4004
student_gender .2798017	.011305	.1369903	0.08	0.934	2571906
student_difficultevent .430281	.166714	.1344752	1.24	0.215	096852
_cons	.127127	'3 .1311182	0.97	0.332	1298597
.3841143					
Note: One or more parameter standard-error es					
Linear regression			,	Replicatio Wald chi2( Prob > chi R-squared	obs = 199 ons = 499 (3) = 4.71 i2 = 0.1945 = 0.0252 ared = 0.0102 = 0.8386
in school)		(Re	plication	s based or	n 13 clusters
	Observ	ved Bootstra	р		
Normal-based helpseeking2_3_changointerval]	•				_
	+				
_treated Mentee	•	.2202418	0.47	0.641	3291019
.5342301	1 .10230	.2202410	0.47	0.041	3291019
1.student_gende	r  10955	.1918069	-0.57	0.568	4854918
_treated#student_gende	 r				
Mentee#1 .8100187	.26774	.2766732	0.97	0.333	2745203
_con:	s   .23076	.1 <b>0</b> 9055	2.12	0.034	.0170253
Note: One or more parameters of standard-error es					

Linear regression Number of obs = 199 Replications = 499 Wald chi2(3) = 3.59 Prob > chi2 = 0.3090 R-squared = 0.0293 Adj R-squared = 0.0144 Root MSE = 0.8368

clusters in s	chool)					
[95% conf. in	 pseeking2_3_ch terval]	ange   coef		std. err.	z	
1524276	 _tre Men	ated   tee   .:				
stud 1904982		:	227293	.1598129	0.77	0.443
_treated#stud 257729	ent_difficulte Mentee# .431094		866825	.1757234	0.49	0.622
1346415	_	cons   .:	147541	.1439733	1.02	0.305
standar	 more parameter d-error estima			mplete rep	lications.	
Linear regres	sion				Prob > chi R-squared Adj R-squa	ns = 499 1) = 2.32 2 = 0.1278
		(Repl	ications	based on	13 cluster	s in school)
belonging~ge	Observed   coefficient	Bootstrap std. err.	Z			al-based f. interval]
	.143212				0/112/	227540

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 183 Replications = 499 Wald chi2(3) = 11.40Prob > chi2 = 0.0098 R-squared = 0.0420

(Replications based on 13 clusters in

school)					
	Observed	Bootstrap			
Normal-based belonging_change interval]	•				_
	•				
_treated .3000479	.0950238	.1046061	0.91	0.364	1100004
<pre>student_gender .253501</pre>	.1092203	.073614	1.48	0.138	0350605
student_difficultevent .3964344	.1924556	.1040727	1.85	0.064	0115232
	0758305	.1084406	-0.70	0.484	2883701
Linear regression		(Rep)	R W P R A	rob > chi -squared dj R-squa oot MSE	
in school)					
Normal-based	Observed	Bootstrap			
belonging_change interval]	·				_
_treated Mentee	•	.1643784	1.19	0.236	1272667
.5170847 1.student_gender .3999002	.2367913	.0832203	2.85	0.004	.0736825
_treated#student_gender Mentee#1 .1353143	:	.168629	-1.16	0.247	5256992

```
.2054811
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
     standard-error estimates include only complete replications.
Linear regression
                                              Number of obs =
                                                              183
                                              Replications =
                                                             499
                                              Wald chi2(3) = 10.88
                                              Prob > chi2 = 0.0124
                                              R-squared = 0.0595
                                              Adj R-squared = 0.0437
                                              Root MSE = 0.6409
                                           (Replications based on 13
clusters in school)
                          | Observed Bootstrap
Normal-based
            belonging_change | coefficient std. err. z P>|z|
[95% conf. interval]
-----
                   _treated |
                   Mentee .3003638 .115886 2.59 0.010
.0732313 .5274962
       student_difficultevent |
                     Yes .4216465 .1325165 3.18
                                                       0.001
.1619189
          .681374
_treated#student_difficultevent |
                Mentee#Yes -.4100098 .1831179 -2.24
                                                       0.025
-.7689143 -.0511054
                     _cons | -.1296296 .1294782 -1.00 0.317
-.3834023 .124143
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
     standard-error estimates include only complete replications.
                                             Number of obs =
Linear regression
                                                            118
                                             Replications =
                                                              500
                                             Wald chi2(1) = 0.01
                                             Prob > chi2 = 0.9046
                                             R-squared = 0.0003
                                             Adj R-squared = -0.0084
                                             Root MSE
                                                       = 1.7584
                            (Replications based on 8 clusters in school)
```

   SSE_aspir~ge	Observed coefficient		Z	P> z		al-based f. interval]
_ :		.6695173 .0646512				
Linear regress	ion				Replication Wald chi2( Prob > chi2 R-squared Adj R-squa	obs = 118 ns = 500 3) = 8.61 2 = 0.0349 = 0.0262 red = 0.0006 = 1.7506
school)			(Re <sub>l</sub>	plicatio	ns based on	8 clusters in
Normal-based SSE_aspirationinterval]		Observed coefficient	std. err.			_
		3281643			0.661	-1.795498
1.128641 student_diffic		.5736843				.0187275
.8504874 -1.183123	_cons	-1.543977	.1841127	-8.39	0.000	-1.904831
Linear regress	ion				Wald chi2( Prob > chi R-squared Adj R-squa	obs = 118 ns = 488 3) = 9.29 2 = 0.0256 = 0.0332 red = 0.0077 = 1.7443
in school)						n 8 clusters
Normal-based	1		Bootstra			
		coefficient				[95% conf.
	+-					

2.057097	_treated   Mentee	9330	0882	1.52563	3 -0.	61	0.541	-3.923273
	ent_gender	.459	9336	.218000	6 2.	11	0.035	.0320628
_treated#stude	ent_gender   Mentee#1	.9614	4973	1.50724	4 0.	64	0.524	-1.992646
-1.26447	_cons	-1.46	6912	.103288	6 -14.	20	0.000	-1.669354
Note: One or more parameters could not be estimated in 12 bootstrap replicates; standard-error estimates include only complete replications.								
Linear regres	sion					Repl Wald Prob R-sq Adj	ications chi2(3 > chi2 uared	bs = 118 s = 500 ) = 0.29 = 0.9624 = 0.0032 ed = -0.0230 = 1.7712
clusters in s	chool)				(	Repli	cations	based on 8
Normal-hased			Ob	served	Bootstr	ар		
Normal-based	_aspiration_ terval]		coef	ficient	std. er	r.		
Normal-based SSE [95% conf. in	_aspiration_ terval]   _t		coef +	ficient	std. er	r. 		
Normal-based SSE [95% conf. in 	_aspiration_ terval]   _t M	reated lentee	coef +   	ficient	std. er			
Normal-based	_aspiration_ terval] t 2.343291 ent_difficul 437116 ent_difficul	reated Mentee .tevent Yes	coef +         .2	ficient  109375	std. er	7. 	-0.09	0.930
Normal-based SSE [95% conf. in	_aspiration_ terval]  _t 2.343291 ent_difficul 437116 ent_difficul Mente	reated Mentee .tevent Yes	coef   	ficient  109375 014358	std. er	3 1	-0.09 0.53	<ul><li>0.930</li><li>0.595</li></ul>

Linear regression

Number of obs = 118 Replications = 500

Wald chi2(1) = 6.53 Prob > chi2 = 0.0106 R-squared = 0.0637 Adj R-squared = 0.0556 Root MSE = 1.5275

## (Replications based on 8 clusters in school)

		(110	pricacions	basea oi	i o ciuscei	3 111 3011001)
SSE_belon~ge	coefficie		Z	P> z		nal-based nf. interval]
_treated	-1.12481		-2.56			32619887 3 .5144371
Linear regress	sion		·		Replication Wald chi2( Prob > chi R-squared Adj R-squa Root MSE	obs = 118 ons = 500 (3) = 7.60 .2 = 0.0550 = 0.0731 ared = 0.0487 = 1.5331
school)			(Re <sub>l</sub>	plication	ns based or	n 8 clusters in
interval]3096718	_treated   nt_gender	coefficient1.254747 .1520841	.4821901 .2848278 .2978009	-2.60 0.53 0.86	0.009 0.593	-2.199822 4061681 3262224
Linear regress	sion				Number of Replication Wald chi2( Prob > chi R-squared Adj R-squared Root MSE	ons = 488 (3) = 11.45

(Replications based on 8 clusters

in school)

	Observed	Bootstra	р				
Normal-based SSE_belong_change interval]					_		
_treated Mentee	-1.845098	1.506282	-1.22	0.221	-4.797356		
1.10716							
1.student_gender .7053105	.0579322	.3303011	0.18	0.861	589446		
_treated#student_gender							
Mentee#1 4.193509	.9892901	1.634836	0.61	0.545	-2.214929		
	ļ						
_cons	.2450981	.1857715	1.32	0.187	1190074		
Note: One or more parameters could not be estimated in 12 bootstrap replicates; standard-error estimates include only complete replications.							
Linear regression			Nı	ımher of	obs = 118		
Linear regression			Re Wa Pr R - Ac	eplication ald chi2( nob > chi squared Ij R-squa	ons = 500 3) = 8.24 .2 = 0.0414 = 0.0812 ared = 0.0570 = 1.5264		
			(Repl	ications.	based on 8		
clusters in school)							
Normal-based	01	oserved	Bootstrap				
SSE_belong [95% conf. interval]	g_change   coe-						
	·						
-3.641759 .0997455	treated   Mentee   -1	.771007	.9544831	-1.86	0.064		
student_difficu 5993355 .894859	:	1477618	.3811791	0.39	0.698		
_treated#student_difficu		9345615	.9052998	1.03	0.302		

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2295674	.6493591	_cons	.2098959	.2242201	0.94	0.349
Linear regres	sion				Replication Wald chi2 Prob > chi R-squared Adj R-squ	obs = 118 ons = 500 (1) = 6.19 i2 = 0.0128 = 0.0944 ared = 0.0866 = 1.3820
			(Replication:	s based or	n 8 cluste	rs in school)
SSE_produ~ge		d Bootst nt std.e		P> z		mal-based nf. interval]
_treated _cons						82674784 9 .2512869
Linear regres	sion		(Re	eplication	Replication Wald chi2 Prob > chi R-squared Adj R-squared Root MSE	obs = 118 ons = 500 (3) = 9.59 i2 = 0.0224 = 0.1192 ared = 0.0960 = 1.3748 n 8 clusters in
Normal-based SSE_productiv interval]		coefficie		. Z		_
	+-					
3031912	_treated	-1.43040	6 .5751203	-2.49	0.013	-2.557621
stude	nt_gender	.127475	3 .2922661	0.44	0.663	4453557
.7003064 student_diffi	cultevent	.43603	8 .3323518	1.31	0.190	2153596
1.087436 .2556364	_cons		6 .2127855		0.448	5784676
Linear regres	sion				Number of Replication	

Wald chi2(3) = 16.04 Prob > chi2 = 0.0011 R-squared = 0.1182 Adj R-squared = 0.0950 Root MSE = 1.3756

(Replications based on 8 clus
-------------------------------

		(Rep.	lications	s based o	n 8 clusters
in school)					
	Observed	Bootstrap			
Normal-based SSE_productivity_change interval]					[95% conf.
	+				
_treated	I				
<del>_</del>	-2.157498	1.39876	-1.54	0.123	-4.899016
.5840208					
1.student_gender .5409166	.0251928 	.2631292	0.10	0.924	490531
_treated#student_gender	İ				
Mentee#1	1.258393	1.293767	0.97	0.331	-1.277344
3.794129	ı				
cons	   0317402	1518995	A 21	0 834	2659773
.3294576	1 .0317402	.1310333	0.21	0.054	.2033773
Note: One or more parame standard-error est					-
Linear regression			Re Wa Pr R - Ac	eplicatio ald chi2( rob > chi -squared	2 = 0.0378 = 0.1329 red = 0.1101
clusters in school)			(Rep]	lications	based on 8
Normal-based SSE_productivity [95% conf. interval]	_change   coef	ficient st			
	treated   Mentee   -2.	033302 1.	100483	-1.85	0.065

	ent_difficult@ 1.163715		.3042767	.4384971	0.69	0.488		
_treated#stude	ent_difficult@ Mentee 3.170734	:	1.063607	1.075085	0.99	0.323		
6033038	-	_cons   -	.0714962	.2713354	-0.26	0.792		
Linear regression  Number of obs = 118 Replications = 500 Wald chi2(1) = 2.57 Prob > chi2 = 0.1087 R-squared = 0.0473 Adj R-squared = 0.0391 Root MSE = 1.3056  (Replications based on 8 clusters in school)								
	   Observed	R)  Bootstra		s based or		s in school)  al-based		
SSE_total~ge	coefficient			P> z		f. interval]		
_treated _cons	•	.5122109		0.109 0.000				
Linear regression  Number of obs = 118 Replications = 500 Wald chi2(3) = 3.79 Prob > chi2 = 0.2849 R-squared = 0.0698 Adj R-squared = 0.0453 Root MSE = 1.3014								
school)			·			8 clusters in		
Normal-based SSE_tota interval]	 al_change   co			. Z		[95% conf.		
	_treated   ·				0.073	-2.104057		
.0951788 stude	nt_gender					190157		
.7589863 student_diffi						1957962		

.7303798	_cons	5284576	.1690305	-3.13	0.002	8597513
Linear regression				R W P R A	ald chi2 rob > ch: -squared	ons = 488 (3) = 5.93 i2 = 0.1149 = 0.0774 ared = 0.0531
in school)			(Rep	lication	s based o	on 8 clusters
	I	Observed	Bootstrap			
Normal-based SSE_total_ interval]	change	coefficient	std. err.	z	P> z	[95% conf.
	+					
M	reated   entee	-1.645228	1.472535	-1.12	0.264	-4.531344
1.240887 1.student_ .546713	gender	.1808203	.1866834	0.97	0.333	1850724
_treated#student_ Men 3.894114	gender   tee#1	1.069727	1.44104	0.74	0.458	-1.75466
1560094	_cons	3966912	.1227991	-3.23	0.001	6373729
Note: One or more standard-er	-	ers could no mates includ				
Linear regression				R W P R A	eplication ald chi2 rob > ch: -squared dj R-squa	obs = 118 ons = 500 (3) = 4.23 i2 = 0.2374 = 0.0668 ared = 0.0423 = 1.3035
clusters in schoo	1)			(Rep	lications	s based on 8
		0	bserved Bo	otstrap		

Normal-based					. z	
-3.410937	_ Me	eated   ntee   -1	.304561	1.074701	-1.21	0.225
	dent_difficult		2178247	.3142248	0.69	0.488
3980446	.8336941 dent_difficult	 event				
-1.220339	Mentee		6519412	.9552628	0.68	0.495
7896376		_cons	4174085	.1899163	-2.20	0.028
Linear regres	ssion				Number of o Replication Wald chi2(1 Prob > chi2 R-squared Adj R-squar Root MSE	s = 499 ) = 0.81 = 0.3672 = 0.0077 ed = 0.0021
		(Rep	lications 	based on	13 clusters	in school)
	Observed   coefficient		Z		[95% conf	l-based . interval]
_treated	.1268735  0251004	.1406874	0.90	0.367	1488687	
	more paramete rd-error estim				•	replicate;
Linear regres	ssion		(Re	plication		s = 499 ) = 12.64 = 0.0055 = 0.0410
•						
	1	Observed	Bootstra	р		

Normal-based brs_change	coefficient	std. err.	z	P> z	[95% conf.
interval]					-
_treated   .4289899	.1056141	.1649907	0.64	0.522	2177616
student_gender   .4573457	.1798321	.1415912	1.27	0.204	0976814
<pre>student_difficultevent   .0783524</pre>	215458	.149906	-1.44	0.151	5092684
_cons   .211116	.0024932	.1064421	0.02	0.981	2061295
Note: One or more parame standard-error est	eters could no	ot be estima	ted in 1	bootstra	p replicate;
Linear regression				Replicatio Wald chi2 Prob > ch: R-squared	obs = 177 ons = 499 (3) = 11.51 i2 = 0.0092 = 0.0250 ared = 0.0081 = 0.7193
in school)					n 13 clusters
			<b></b>		<b></b>
Normal-based	Observed	Bootstrap			
	coefficient	std ann	7	DNIT	[95% conf

Normal-based	Observed	Bootstrap			
brs_change	coefficient	std. err.	Z	P> z	[95% conf.
interval]	+				
	•				
_treated	1				
Mentee	1903178	.2483372	0.77	0.443	296414
.6770497					
1.student_gender	.2849026	.1015931	2.80	0.005	.0857838
.4840214					
_treated#student_gender	ļ				
Mentee#1 .232881	2269643	.2346192	-0.97	0.333	6868096
	I				
_cons	1212121	.0855852	-1.42	0.157	2889561
.0465319					
	<b> </b>			<del></del> -	

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Replications = 499 Wald chi2(3) = 5.41Prob > chi2 = 0.1441R-squared = 0.0269 Adj R-squared = 0.0100Root MSE = 0.7186

(Replications based on 13

clusters in	school)						
			Observed	Bootstrap			
Normal-based		rs change l	coefficient	•	Z	P> z	
[95% conf. i							
1795038	.5201175	_treated   Mentee	.1703069	.1784781	0.95	0.340	
1/95056	.52011/5	I					
stu	ıdent_diffi	cultevent     Yes	1683874	.1213966	-1.39	0.165	
4063203	.0695456	ı					
_treated#stu		ultevent   ntee#Yes	0564353	.2103345	-0.27	0.788	
4686835	.3558128						
		_cons	.0438776	.1040949	0.42	0.673	
1601448	.2478999						

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression Number of obs = 103

Replications = 500 Wald chi2(1) = 0.68Prob > chi2 = 0.4106R-squared = 0.0091 Adj R-squared = -0.0007

= 4.3314 Root MSE

(Replications based on 8 clusters in school)

   wembs_change		std. err.		P> z	Normal [95% conf.	
_treated	-1.134698		-0.82	0.411 0.513	-3.837364 -1.083658	1.567968 2.170555

Replications = 500 Wald chi2(3) = 3.61 Prob > chi2 = 0.3070 R-squared = 0.0364 Adj R-squared = 0.0072 Root MSE = 4.3142

school)					n 8 clusters in
Normal-based	Observed	Bootstrap			
	<b></b>				
_treated   1.751084	-1.46272	1.639726	-0.89	0.372	-4.676525
student_gender 1.670439	.0124515	.8459275	0.01	0.988	-1.645536
<pre>student_difficultevent 3.104632</pre>	1.453767	.8422934	1.73	0.084	1970976
_cons 2.104747	0456941	1.097184	-0.04	0.967	-2.196135
Linear regression		(Rer	R W P R A	eplication ald chi2( rob > chi -squared dj R-squa oot MSE	i2 = 0.6990 = 0.0397 ared = 0.0106
in school)					
Normal-based wembs_change interval]	coefficient	Bootstrap std. err.			[95% conf.
	1				
_treated Mentee 2.912936	   -4.028526	3.541627	-1.14	0.255	-10.96999
1.student_gender .8021777	4555264	.6416976	-0.71	0.478	-1.713231
_treated#student_gender Mentee#1	     4.436254	3.92822	1.13	0.259	-3.262916

12.13342	1							
2.408759	_cons	.700	5264	.871563	32	0.80	0.422	-1.007706
Note: One or m	nore parameto I-error esti							
Linear regress	sion					Re Wa Pr R- Ad Ro	eplication ald chi2 nob > chi squared alj R-squa not MSE	obs = 103 ons = 500 (3) = 6.16 i2 = 0.1040 = 0.0627 ared = 0.0343 = 4.2549
clusters in sc	chool)					(керт	.icacion:	s based on 8
Normal-based			Ob	served	Boots	trap		
[95% conf. int	erval]	_					z	
-9.038357 1	_ Mo	reated entee		809487	2.66	784	-1.43	0.153
	ent_difficul <sup>.</sup> 2.538548	tevent Yes	:	508465	.8610	878	0.99	0.323
_treated#stude -2.050736 1	ent_difficul Mente .0.00371			976487	3.07	517	1.29	0.196
-1.637412 2	2.039719	_cons	.2	011538	.9380	608	0.21	0.830
<pre>bysort _treated: sum z_epoc_engagement_change z_epoc_perser_change z_epoc_optimism_change z_epoc_connect_change z_helpseeking2_1_change z_helpseeking2_2_change z_help &gt; seeking2_3_change z_helpseeking_change z_belonging_change z_SSE_aspiration_change z_SSE_belong_change z_SSE_productivity_change z_SSE_total_change z_brs_change z_wemb &gt; s_change</pre>								

-> \_treated = Non - Mentee

Obs.	Mean	Std. dev.	Min	Max
95	0791839	.9731418	-1.834786	3.621555
95	0216857	.8875813	-2.07273	2.277969
83	0734959	.9603692	-2.280513	3.28701
83	1654989	.9769721	-2.664648	2.888397
99	1971325	.9829352	-2.599917	3.230369
+				
99	1680212	.9544397	-2.596461	2.04925
98	129833	1.006844	-2.742853	2.021351
78	.1128577	.9285732	-1.916379	3.033852
88	1045789	1.016673	-3.747677	3.142848
101	.021749	.9772979	-3.887356	2.546741
+				
101	.0742959	.973176	-2.797281	2.711655
101	.0923079	.9221056	-3.731771	3.568357
101	.0736057	.9638592	-3.795455	3.062357
83	0962769	.7164656	-1.684354	1.793493
87	.0592671	.9007898	-2.329309	2.664796
	95 95 83 83 99 99 98 78 88 101 101 101 101 101	950791839 950216857 830734959 831654989 991971325 	950791839 .9731418 950216857 .8875813 830734959 .9603692 831654989 .9769721 991971325 .9829352 	950791839 .9731418 -1.834786 950216857 .8875813 -2.07273 830734959 .9603692 -2.280513 831654989 .9769721 -2.6644648 991971325 .9829352 -2.599917 

## -> \_treated = Mentee

Variable	0bs	Mean	Std. dev.	Min	Max
z_epoc_en~ge	17	.1765715	1.094995	-2.653237	1.984653
z_epoc_pe~ge	102	0700268	1.002665	-3.315786	1.967205
z_epoc_op~ge	95	.0188912	1.021203	-2.280513	2.304506
z_epoc_co~ge	95	.1471899	1.021185	-3.358778	3.235462
z_h~1_change	101	.2055474	.9918785	-2.599917	2.064312
z_h~2_change	100	.2025798	.97794	-2.596461	2.04925
z_h~3_change	101	.1463301	.9871583	-2.742853	2.021351
z_h~g_change	12	4884278	1.418722	-3.566456	1.256846
z_belongi~ge	95	.1158777	.9952391	-2.721429	2.153251
z_SSE_asp~ge	17	0251773	1.303883	-2.132602	2.108053
z_SSE_bel~ge	17	6547053	1.089252	-2.581244	1.372227
z_SSE_pro~ge	17	7904832	1.219277	-3.147761	.6483058
z_SSE_tot~ge	17	5662363	1.299916	-2.886849	1.007177
z_brs_change	94	.0802218	1.201184	-3.075493	2.952776
z_wembs_ch~e	16	1963396	1.325863	-3.759736	1.747971

<sup>\*\*\*\*\*\*\*</sup> Help Seeking Behaviourforeach var of varlist z\_helpseeking2\_1\_change z\_helpseeking2\_2\_change

```
z_helpseeking2_3_change z_helpseeking_change {
       bootstrap, reps(500) seed(7582) nodots: regress `var' _treated,
cluster(school)
. *** Outcomes adjusting for student gender and difficult live event
        bootstrap, reps(500) seed(7582) nodots: regress `var' treated
student_gender student_difficultevent, cluster(school)
. *** Interaction between student gender and mentoring
        bootstrap, reps(500) seed(7582) nodots: regress `var'
_treated##student_gender, cluster(school)
. *** Interaction between difficult life event and mentoring
    bootstrap, reps(500) seed(7582) nodots: regress `var'
_treated##student_difficultevent, cluster(school)
Linear regression
                                            Number of obs = 200
                                            Replications = 499
                                            Wald chi2(1) = 6.19
                                            Prob > chi2 = 0.0129
R-squared = 0.0403
                                            Adj R-squared = 0.0355
                                            Root MSE = 0.9875
                          (Replications based on 13 clusters in school)
______
_treated | .4026799 .1619137 2.49 0.013 .0853349 .7200248   _cons | -.1971325 .1216942 -1.62 0.105 -.4356487 .0413837
-----
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
    standard-error estimates include only complete replications.
                                            Number of obs = 200
Linear regression
                                            Replications = 499
                                            Wald chi2(3) = 7.43
                                            Prob > chi2 = 0.0595
                                            R-squared = 0.0416
                                            Adj R-squared = 0.0269
                                            Root MSE = 0.9918
                                  (Replications based on 13 clusters in
school)
-----
                 Observed Bootstrap
Normal-based
z_helpseeking2_1_cha~e | coefficient std. err. z P>|z| [95% conf.
interval]
```

	.394947	.1729845	2.28	0.022	.0559035					
.7339904 student_gender	0081713	.1384949	-0.06	0.953	2796162					
.2632737	074554	450044	0.47	0.630	2260440					
<pre>student_difficultevent   .3860198</pre>	.074554	.158914	0.47	0.639	2369118					
_cons	2222724	.1278377	-1.74	0.082	4728297					
.028285										
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.										
Linear regression $ \begin{array}{lllllllllllllllllllllllllllllllllll$										
		(Repl:	ications	based or	n 13 clusters					
in school)		, .								
	Observed	Bootstrap								
Normal-based z_helpseeking2_1_change interval]	coefficient				_					
	+									
_treated Mentee	•	.1825724	1.70	0.089	0473915					
.6682792 1.student_gender	l - 1060052	.230998	-0.46	0.646	5587529					
.3467425	.1000032 	.230330	0.40	0.040	. 3307 323					
_treated#student_gender Mentee#1	     .2034119	.3264344	0.62	0.533	4363878					
.8432117	I									
_cons 0294964	  1617974	.0675018	-2.40	0.017	2940985					

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Prob > chi2 = 0.0456 R-squared = 0.0457

Adj R-squared = 0.0311 Root MSE = 0.9897

(Replications based on 13

clusters in school)							
Normal-based  z_helpseeking2_1_change   c  [95% conf. interval]	coefficient						
treated   treated   Mentee   0644672 .6226701	.2791015	.1752933	1.59	0.111			
student_difficultevent   Yes  4025351 .2774915	0625218	.1734794	-0.36	0.719			
_treated#student_difficultevent   Mentee#Yes  1013195 .6232799	.2609802	.1848502	1.41	0.158			
_cons   4567316 .1092001	1737658	.144373	-1.20	0.229			
Note: One or more parameters could not be estimated in 1 bootstrap replicate;							

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 199
Replications = 499
Wald chi2(1) = 5.18
Prob > chi2 = 0.0228
R-squared = 0.0358
Adj R-squared = 0.0309
Root MSE = 0.9663

(Replications based on 13 clusters in school)

z_h~2_change		std. err.	P>   z	Normal [95% conf.	-based interval]
	.370601 1680212	.1627891	0.023 0.086	.0515402 3600355	.6896619

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 199Replications = 499Wald chi2(3) = 12.75

Prob > chi2 = 0.0052 R-squared = 0.0449 Adj R-squared = 0.0302 Root MSE = 0.9667

(Replications based on 13 clusters in

h1\		(Repli	cations	based on	13 clusters in			
school)								
	Observed	Bootstrap						
Normal-based z_helpseeking2_2_cha~e   interval]					_			
+								
_treated   .6726998	.318907	.1805099	1.77	0.077	0348859			
student_gender   .4682318								
<pre>student_difficultevent   .3336072</pre>								
_cons   0239385	250796	.1157458	-2.17	0.030	4776536			
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.								
Linear regression  Number of obs = 19  Replications = 49  Wald chi2(3) = 11.0  Prob > chi2 = 0.011  R-squared = 0.047  Adj R-squared = 0.032  Root MSE = 0.965								
in school)		(Repl	ications	s based o	n 13 clusters			
Normal-based	Observed	Bootstrap						
<pre>z_helpseeking2_2_change interval]</pre>					<u>-</u>			
	+							
_treated Mentee	   .2161965	.2514048	0.86	0.390	2765478			
.7089408 1.student_gender .5983872	.0527922	.2783699	0.19	0.850	4928028			
_treated#student_gender Mentee#1	     .2437629	.3929166	0.62	0.535	5263395			

0109504	_cons						0.037	
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.								
Linear regression	o1)					F W F F	Wald chi2( Prob > chi R-squared Adj R-squa Root MSE	obs = 199 ons = 499 (3) = 5.44 i2 = 0.1424 = 0.0386 ared = 0.0238 = 0.9698 based on 13
					Doo+c	+		
Normal-based z_helpsee [95% conf. interv	/al]		coef <sup>.</sup>		std.	err.		P> z
0349004 .65	_†	treated					1.76	
student_ 3877646 .451	_difficu:	ltevent   Yes	.0	318962	.2141	.166	0.15	0.882
_treated#student_ 3815886 .60	Mente	ltevent   ee#Yes	.1	117137	.2516	895	0.44	0.657
392568 .0326	5839	_cons	:	179942	.1084	1846	-1.66	0.097
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.								
Linear regression	1					F W F F A	Replicatio Wald chi2( Prob > chi R-squared Wdj R-squa	obs = 199 ons = 499 (1) = 2.26 i2 = 0.1330

## (Replications based on 13 clusters in school)

z_h~3_change		std. err.		P> z	-	-based interval]
	.276163		1.50	0.133 0.378	0841364 4182379	.6364625 .1585719

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 199
Replications = 499
Wald chi2(3) = 5.69
Prob > chi2 = 0.1275
R-squared = 0.0287
Adj R-squared = 0.0138
Root MSE = 0.9970

(Replications based on 13 clusters in

school)

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Observed	Bootstrap			
		Z	P> z	[95% conf.
.2475342	.1978423	1.25	0.211	1402295
.0134655	.1631625	0.08	0.934	306327
.1985655	.1601668	1.24	0.215	1153556
2093359	.1561685	-1.34	0.180	5154205
	.0134655 .1985655	.2475342 .1978423 .0134655 .1631625 .1985655 .1601668	coefficient std. err. z  .2475342 .1978423 1.25  .0134655 .1631625 0.08  .1985655 .1601668 1.24	coefficient std. err. z P> z   .2475342 .1978423 1.25 0.211  .0134655 .1631625 0.08 0.934  .1985655 .1601668 1.24 0.215

Note: One or more parameters could not be estimated in 1 bootstrap replicate;

standard-error estimates include only complete replications.

Linear regression

Number of obs = 199
Replications = 499
Wald chi2(3) = 4.71
Prob > chi2 = 0.1945
R-squared = 0.0252
Adj R-squared = 0.0102
Root MSE = 0.9988

(Replications based on 13 clusters

in school)

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Naggar I hadad	Observed	Bootstrap			
Normal-based z_helpseeking2_3_change interval]	coefficient				_
_treated Mentee	•	.2623192	0.47	0.641	3919771
.6362953 1.student_gender .3172692	1304881	.2284518	-0.57	0.568	5782454
_treated#student_gender Mentee#1 .9647735		.3295319	0.97	0.333	3269676
_cons .1686868	0858931	.1298901	-0.66	0.508	340473
Note: One or more parame standard-error est					replicate;
Linear regression			Re Wa Pr Re Ac Ro	eplicatio ald chi2( rob > chi -squared dj R-squa oot MSE	obs = 199 ns = 499 3) = 3.59 2 = 0.3090 = 0.0293 red = 0.0144 = 0.9967
clusters in school)			(керт	cations	based on 13
Namal based	0	bserved Bo	otstrap		
Normal-based z_helpseeking2_3 [95% conf. interval]	3_change	fficient st			P> z
181549 .5923639	_treated   Mentee   .	2054075 .1	974304	1.04	0.298
student_diffic		1461768 .1	903453	0.77	0.443
2268931 .5192467	I				
_treated#student_difficu Ment 3069684 .5134549		1032432 .2	092955	0.49	0.622
.5055004	_cons	1850222 .1	714795	-1.08	0.281

-----

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Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 90
Replications = 500
Wald chi2(1) = 1.90
Prob > chi2 = 0.1686
R-squared = 0.0407
Adj R-squared = 0.0298
Root MSE = 1.0030

## (Replications based on 8 clusters in school)

z_h~g_change		std. err.		P> z	-	
_treated	6012855 .1128577	.4367318	-1.38	0.169 0.269	-1.457264 0873702	.2546931 .3130856

Linear regression

Number of obs = 90 Replications = 500 Wald chi2(3) = 3.74 Prob > chi2 = 0.2912 R-squared = 0.0494 Adj R-squared = 0.0163 Root MSE = 1.0100

(Replications based on 8 clusters in

					-		
c	$\boldsymbol{\mathcal{C}}$	h	റ	^	-	١	

-----Observed Bootstrap Normal-based z helpseeking change | coefficient std. err. z P>|z| [95% conf. interval] ------\_treated | -.5254393 .5040915 -1.04 0.297 -1.51344 .4625618 student gender | -.1887341 .2688191 -0.70 0.483 -.7156098 .3381416 student\_difficultevent | -.0390563 .2402643 -0.16 0.871 -.5099656 .431853 \_cons | .1888718 .1060127 1.78 0.075 -.0189094 .3966529 \_\_\_\_\_\_

Linear regression

Number of obs = 90 Replications = 488

Wald chi2(3) = 8.66 Prob > chi2 = 0.0341 R-squared = 0.1226 Adj R-squared = 0.0920 Root MSE = 0.9703

(Replications	based	on	8	clusters
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		(Rep	lications	based o	n 8 clusters
in school)					
	Observed	Bootstrap			
Normal-based z_helpseeking_change interval]		std. err.			_
	+				
	   -1.596192	.7336329	-2.18	0.030	-3.034086
158298 1.student_gender 0721484	4309838	.1830827	-2.35	0.019	7898193
_treated#student_gender Mentee#1 3.545608	     1.71614	.9334193	1.84	0.066	1133283
_cons .4030316	.2509935	.0775719	3.24	0.001	.0989555
Note: One or more parame standard-error est			ete repli	cations.	
Linear regression			Re Wa Pr R- Ad	plicatio ld chi2( ob > chi squared j R-squa	obs = 90 ns = 499 3) = 3.65 2 = 0.3020 = 0.0584 red = 0.0256 = 1.0052
clusters in school)			(Repl	ications	based on 8
Normal-based z_helpseeking [95% conf. interval]	_change   coef				
	treated   Mentee   -1.	085648 .6	547556	-1.66	0.097

```
student_difficultevent
                        Yes
                               -.1807671 .2669523 -0.68
                                                             0.498
-.7039841
           .3424498
treated#student difficultevent |
                  Mentee#Yes |
                                .7995461 .8102666
                                                      0.99
                                                             0.324
-.7885473
           2.387639
                                 .184701 .0682154
                        cons
                                                     2.71
                                                             0.007
.0510013
          .3184008
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
     standard-error estimates include only complete replications.
 **********
 ***** EPOCH MEASURE
 ******
         foreach var of varlist z epoc engagement change z epoc perser change
z_epoc_optimism_change z_epoc_connect_change {
         bootstrap, reps(500) seed(7582) nodots: regress `var' _treated,
cluster(school)
 *** Outcomes adjusting for student gender and difficult live event
         bootstrap, reps(500) seed(7582) nodots: regress `var' _treated
student_gender student_difficultevent, cluster(school)
 *** Interaction between student gender and mentoring
         bootstrap, reps(500) seed(7582) nodots: regress `var'
_treated##student_gender, cluster(school)
. *** Interaction between difficult life event and mentoring
         bootstrap, reps(500) seed(7582) nodots: regress `var'
treated##student difficultevent, cluster(school)
 6.
                                                  Number of obs =
Linear regression
                                                                    112
                                                  Replications =
                                                                    500
                                                  Wald chi2(1) =
                                                                   0.74
                                                  Prob > chi2 = 0.3893
                                                  R-squared = 0.0086
                                                  Adj R-squared = -0.0004
                                                  Root MSE = 0.9918
                               (Replications based on 8 clusters in school)
              _____
               Observed Bootstrap
                                                        Normal-based
z_epoc_en~ge | coefficient std. err. z P>|z| [95% conf. interval]
   _treated |
              .2557554 .2970572 0.86 0.389 -.326466 .8379769
```

_cons  079183	39 .1143114	-0.69	0.488	3032302	.1448623
Linear regression			F F F	Replication Wald chi2(3 Prob > chi2 R-squared Ndj R-squar	bs = 112 s = 500 ) = 3.18 = 0.3641 = 0.0188 ed = -0.0085 = 0.9958
school)				is based on	8 clusters in
Normal-based	Observed	•		D. II	F05% 5
z_epoc_engagement_ch~e   interval]					_
	.1695621	.3517904	0.48	0.630	5199344
student_gender   .5751788	.1521786	.2158204	0.71	0.481	2708216
student_difficultevent   .3847375	.1228143	.1336367	0.92	0.358	1391088
	1798789	.1323053	-1.36	0.174	4391927
Linear regression			F F F F	R-squared Adj R-squar Root MSE	s = 488 ) = 4.63 = 0.2006 = 0.0252 ed = -0.0019 = 0.9925
in school)		(Re	plicatio	ons based o	n 8 clusters
Normal based	Observed	Bootstrap	)		
Normal-based z_epoc_engagement_cha~e interval]					[95% conf.
treated Mentee	· 	.9060256		2 0.826	-1.974377
1.577178 1.student_gender	.0841386	.1986371	0.42	0.672	305183

.4734602					
_treated#student_gender   Mentee#1   2.503196	.6009353	.970559	0.62	2 0.536	-1.301325
1	108411	.0920881	-1.18	8 0.239	2889004
Note: One or more parameter standard-error estima					p replicates;
Linear regression			,,	Number of Replicatio Wald chi2( Prob > chi R-squared	obs = 112 ns = 500 3) = 3.46 2 = 0.3254 = 0.0310 red = 0.0041 = 0.9896
clusters in school)			(Re	eplications	based on 8
Normal-based z_epoc_engagement_ch [95% conf. interval]	ange   coef		std. err		
	eated   etee  2				
student_difficulte	•	310193	.0976243	0.32	0.751
_treated#student_difficulte Mentee# 2425071 1.7595		584967	.5107256	1.49	0.138
297213 .1146827	_cons  0	912651	.1050774	-0.87	0.385
Linear regression			F V F F	Number of o Replication Wald chi2(1 Prob > chi2 R-squared Adj R-squar	s = 499 ) = 0.08 = 0.7790

## (Replications based on 13 clusters in school)

z_epoc_pe~ge	-	std. err.		P> z	-	interval]
_treated	0483411 0216857	.1722771	-0.28	0.779 0.844	385998 2377342	.2893158 .1943628

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 197
Replications = 499
Wald chi2(3) = 2.39
Prob > chi2 = 0.4946
R-squared = 0.0116
Adj R-squared = -0.0037
Root MSE = 0.9486

(Replications based on 13 clusters in

c	c	h	^	$\sim$	1	١
3	L	11	U	U	_	,

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Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 197
Replications = 499
Wald chi2(3) = 0.74
Prob > chi2 = 0.8642
R-squared = 0.0045
Adj R-squared = -0.0110
Root MSE = 0.9520

(Replications based on 13 clusters

in school)

		Observ	ed Bootstr	ran		
Normal-based		l observ	eu bootsti	ар		
	rser_change	coeffici	ent std. er	rr. z	P> z	[95% conf.
interval]		<b></b>				
	_treated Mentee		05 187588	36 -0.1	7 0.866	3992173
.3361164						
	dent_gender	.16160	95 .210657	75 0.7	7 0.443	2512715
.5744906						
_treated#stud						
.3817092	Mentee#1	09358	52 .242501	16 -0.3	9 0.700	5688796
13017032						
.0989751	_cons	07782	38 .090205	52 -0.8	6 0.388	2546227
.0909731						
Notes One on			mat ha aati	مد امعادسا	1 600+0+0	
Note: One or standar	more parame rd-error est:					
			,			
Linear regres	ssion					obs = 197 ons = 499
					Wald chi2(	(3) = 2.99
						12 = 0.3931 = 0.0116
					•	red = -0.0038
					Root MSE	= 0.9486
				(Re	plications	based on 13
clusters in s	school)			•	•	
		1	Observed	Bootstra	р	
Normal-based	_epoc_perser	change	coefficient	std ann	. 7	D\ 7
[95% conf. ir	nterval]					
		+-				
		•				
		treated				
	-1	treated   Mentee	.009		0.03	
5072778	-1					
	-1	Mentee   		.2634119	0.03	0.973
stud	.5252779 dent_difficu	Mentee   		.2634119	0.03	
	.5252779	Mentee           Ltevent	.009	.2634119	0.03	0.973
stud	.5252779  dent_difficu: .601768  dent_difficu:	Mentee     Ltevent     Yes     Ltevent	.009	.2634119 .1680534	0.03	0.973 0.105
stud 0569893	.5252779  dent_difficu  .601768  dent_difficu  Ment	Mentee           Yes	.009	.2634119 .1680534	0.03	0.973 0.105

```
_cons | -.1277742 .1113771 -1.15 0.251
-.3460692
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
    standard-error estimates include only complete replications.
                                            Number of obs = 178
Linear regression
                                            Replications =
                                                            499
                                            Wald chi2(1) = 0.16
                                            Prob > chi2 = 0.6859
R-squared = 0.0022
                                             Adj R-squared = -0.0035
                                             Root MSE = 0.9933
                           (Replications based on 13 clusters in school)
          | Observed Bootstrap
                                                  Normal-based
z_epoc_op~ge | coefficient std. err. z P>|z| [95% conf. interval]
reated | .0923871 .2284055 0.40 0.686 -.3552794
_cons | -.0734959 .1652772 -0.44 0.657 -.3974333
   _treated |
                                                        .5400535
Note: One or more parameters could not be estimated in 1 bootstrap replicate;
     standard-error estimates include only complete replications.
Linear regression
                                             Number of obs = 178
                                             Replications = 499
                                             Wald chi2(3) = 7.26
                                             Prob > chi2 = 0.0640
                                             R-squared = 0.0298
                                             Adj R-squared = 0.0131
                                             Root MSE = 0.9851
                                   (Replications based on 13 clusters in
school)
______
                  Observed Bootstrap
Normal-based
z_epoc_optimism_change | coefficient std. err. z P>|z| [95% conf.
------
           treated | .009691 .2541604 0.04 0.970 -.4884541
.5078361
      student_gender | .2870675 .1560087 1.84 0.066 -.018704
.592839
student difficultevent | .1567734 .1479966 1.06 0.289
                                                      -.1332946
.4468415
             _cons | -.2345584 .1610166 -1.46 0.145 -.5501451
.0810282
```

-----

Note:	One or	more	parameters	could	not	be	estimated	in	1	bootstrap	replicate;
	standa	rd-err	ror estimat	es inc	lude	on]	y complet	e re	p.	lications.	

Linear regression				F W F F	Prob > chi R-squared	ns = 499 3) = 9.38 2 = 0.0247 = 0.0360 red = 0.0194
in school)			(Rep.	lications	based on	13 clusters
Normal-based	I	0bserved	Bootstrap			
z_epoc_optimism_cha interval]	nge		std. err.		P> z	[95% conf.
	ted   ee	.2228002	.2450449	0.91	0.363	2574789
.7030792 1.student_gen .9601513	der	.5526408	.2079173	2.66	0.008	.1451303
_treated#student_gen Mentee .021791		455879	.2437137	-1.87	0.061	9335491
_c	ons	259929	.1809295	-1.44	0.151	6145443
Note: One or more pa standard-error						replicate;
Linear regression				F W F F	Prob > chi R-squared	ns = 499 3) = 13.45 2 = 0.0038 = 0.0212 red = 0.0043
clusters in school)				(Repl	ications	based on 13
Normal-based z_epoc_opti	mism_	Ob _change   coef		ootstrap	z	P> z

	iterval]		-+				
3289887	.859786	_treated Mentee		553986	.3032644	0.88	0.381
	lent_diffi 6354101	cultevent Yes		105871	.1171139	3.47	0.001
_treated#stud	lent_diffi	cultevent ntee#Yes		195169	.2864793	-1.46	0.143
6231505	.1436379	_cons	23	397563	.1956129	-1.23	0.220
Note: One or standar					imated in i		
Linear regres	sion					Number of Replication	obs = 178 ons = 499
			(Renli	cations	s hased on	Wald chi2 Prob > ch R-squared Adj R-squared Root MSE	(1) = 4.53 $i2 = 0.0333$ $= 0.0240$ $ared = 0.0184$ $= 1.0008$
z_epoc_co~ge	•	ved Boo ient std	 tstrap			Wald chi2 Prob > ch R-squared Adj R-squ Root MSE  13 cluste	(1) = 4.53 i2 = 0.0333 = 0.0240 ared = 0.0184 = 1.0008 rs in school) 
_treated	coeffic:  31268 16549	ient std  889 .14 989 .12	tstrap . err.  69279 75972	z 2.13 -1.30	P> z  0.033 0.195	Wald chi2 Prob > chi R-squared Adj R-squa Root MSE  13 cluste Norn [95% coi .0247154	(1) = 4.53 i2 = 0.0333 = 0.0240 ared = 0.0184 = 1.0008 rs in school) 
treated cons 	coeffic:   .3126  16549   more para	ient std  889 .14 989 .12  meters co	tstrap . err.  69279 75972 	z 2.13 -1.30 be esti	P> z  0.033 0.195	Wald chi2 Prob > chi R-squared Adj R-squared Root MSE  13 cluste Nori [95% coi .02471544155844	(1) = 4.53 i2 = 0.0333 = 0.0240 ared = 0.0184 = 1.0008 rs in school) 
cons  Note: One or	coeffic:   .3126  1654   more para	ient std  889 .14 989 .12  meters co	tstrap . err.  69279 75972 	z 2.13 -1.30 be esti	P> z  0.033 0.195	Wald chi2 Prob > chi R-squared Adj R-squi Root MSE  13 cluste Norm [95% com .024715 .415584 .415584 .150lications Number of Replication Wald chi2 Prob > chi R-squared Adj R-squi	(1) = 4.53 i2 = 0.0333 = 0.0240 ared = 0.0184 = 1.0008 rs in school) 

| Observed Bootstrap

Normal-based z_epoc_connect_change interval]	-					-
<del>_</del>		.2638795	.1550756	1.70	0.089	040063
.5678221	1	1644007	1227015	1 22	A 194	Q702101
student_gender .4070356	I	.1044087	.1237915	1.33	0.184	0782181
student_difficultevent .3659007	I	.1068004	.1321965	0.81	0.419	1522999
_cons	-	.2647117	.1200994	-2.20	0.028	5001021
Note: One or more para standard-error e						
Linear regression	3 CIME	ics includ	C Only Comp		Number of Replicati Wald chi2 Prob > ch R-squared Adj R-squ	obs = 178 ons = 499 (3) = 6.77 i2 = 0.0798 = 0.0313 ared = 0.0146 = 1.0028
					NOOC MSL	- 1.0020
in school)			(Rep	olicatior 	s based o	n 13 clusters
Normal-based	l	Observed	Bootstrap	o		

Normal-based z_epoc_connect_change   interval]			z	P> z	-
_treated   Mentee   .6576427 1.student_gender   .2937566	.2329544	.2166817	1.08 1.46	0.282 0.144	1917339 0426978
_treated#student_gender   Mentee#1   .557377	.085339	.2408401	0.35	0.723	386699
_cons   _cons	2078462	.1427267	-1.46	0.145	4875853

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Replications = 499 Wald chi2(3) = 28.54 Prob > chi2 = 0.0000 R-squared = 0.0417 Adj R-squared = 0.0252 Root MSE = 0.9974

(Replications based on 13

```
clusters in school)
                           Observed Bootstrap
Normal-based
       z_epoc_connect_change | coefficient std. err. z P>|z|
[95% conf. interval]
-----
______
                  treated
                  Mentee
                           .5211881 .1845257 2.82
                                                   0.005
.1595244 .8828519
      student difficultevent |
                    Yes |
                           .3833828 .0839721
                                             4.57
                                                   0.000
.2188004
         .5479652
_treated#student_difficultevent |
               Mentee#Yes
                          -.4873793 .2111686
                                             -2.31
                                                   0.021
-.9012622
         -.0734964
                         -.3225473
                                  .0995222
                                            -3.24
                                                   0.001
                   _cons
```

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

```
5.
. *** Interaction between difficult life event and mentoring
      bootstrap, reps(500) seed(7582) nodots: regress `var'
treated##student difficultevent, cluster(school)
Linear regression
                                     Number of obs =
                                     Replications =
                                                 500
                                     Wald chi2(1) = 0.01
                                     Prob > chi2 = 0.9046
                                     R-squared = 0.0003
                                     Adj R-squared = -0.0084
                                     Root MSE
                                           = 1.7584
                       (Replications based on 8 clusters in school)
-----
      Observed Bootstrap
                                         Normal-based
SSE_aspir~ge | coefficient std. err. z P>|z| [95% conf. interval]
_cons | -1.316832 .0646512 -20.37 0.000 -1.443546 -1.190118
    ______
                                     Number of obs =
Linear regression
                                     Replications = 500
                                     Wald chi2(3) = 8.61
                                     Prob > chi2 = 0.0349
                                     R-squared = 0.0262
                                     Adj R-squared = 0.0006
                                     Root MSE = 1.7506
                             (Replications based on 8 clusters in
school)
 -----
                  Observed
                         Bootstrap
Normal-based
SSE aspiration change | coefficient std. err. z > |z| [95% conf.
interval]
------
         _treated | -.3281643 .7486532 -0.44 0.661 -1.795498
1.139169
     student gender | .5736843
                         .2831465 2.03 0.043
                                            .0187275
1.128641
student difficultevent | .1083807
                         .3786328 0.29 0.775 -.6337259
.8504874
           _cons | -1.543977 .1841127 -8.39 0.000
                                             -1.904831
-1.183123
                                     Number of obs =
Linear regression
                                                  118
                                     Replications =
                                                  488
```

Wald chi2(3) = 9.29 Prob > chi2 = 0.0256 R-squared = 0.0332 Adj R-squared = 0.0077 Root MSE = 1.7443

		(Rep	lications	based o	n 8 clusters
in school)					
	Observed	Bootstrap			
Normal-based SSE_aspiration_change interval]	coefficient				_
	+				
_treated Mentee 2.057097	  9330882	1.525633	-0.61	0.541	-3.923273
1.student_gender .8866093	.459336	.2180006	2.11	0.035	.0320628
_treated#student_gender Mentee#1 3.915641	     .9614973	1.507244	0.64	0.524	-1.992646
_cons -1.26447	-1.466912	.1032886	-14.20	0.000	-1.669354
Note: One or more parame standard-error est					p replicates;
Linear regression			Rep Wal Pro R-s Adj	lication d chi2(3 b > chi2 quared	bs = 118 s = 500 ) = 0.29 = 0.9624 = 0.0032 ed = -0.0230 = 1.7712
clusters in school)			(Repl	ications	based on 8
Normal-based SSE_aspiration [95% conf. interval]	_change   coef				
	treated   Mentee	109375 1.	251383	-0.09	0.930

	ent_difficulte 437116	•	2014358	.3787191	0.53	0.595		
_treated#stude -2.102123 2	ent_difficulte Mentee# 2.017433		0423449	1.050926	-0.04	0.968		
- <b>1.718</b> 565 -1	<del>-</del>	_cons	390625	.1673195	-8.31	0.000		
Linear regression  Number of obs = 183 Replications = 499 Wald chi2(1) = 2.32 Prob > chi2 = 0.1278 R-squared = 0.0120 Adj R-squared = 0.0065 Root MSE = 1.0056  (Replications based on 13 clusters in school)								
	Observed	Bootstrap	)		Norr	nal-based		
z_belongi~ge	coefficient	std. err.	Z	P> z	[95% co	nf. interval]		
_treated _cons						2 .5042185 3 .184566		
Note: One or m	-							
standard-error estimates include only complete replications.  Linear regression  Number of obs = 183 Replications = 499 Wald chi2(3) = 11.40 Prob > chi2 = 0.0098 R-squared = 0.0420 Adj R-squared = 0.0259 Root MSE = 0.9957								
school)			·	•		13 clusters in		
Normal-based z_belongin	 ng_change   co		Bootstra	p . z	P> z			
.4618856	_treated   _tgender		.1610277	0.91	0.364			
Scuder	rc_gender	. 100130/	• 1133134	1.40	0.130	.0337/11		

.3902326 student_difficultevent   .6102603cons   .0546382	.2962609 2725395							
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.								
Linear regression in school)		(Repl	R W P R A	eplication  ald chi2( rob > chi -squared dj R-squa oot MSE	obs = 183 ons = 499 (3) = 9.46 i2 = 0.0238 = 0.0262 ared = 0.0099 = 1.0039			
	Ohserved	Bootstrap						
Normal-based z_belonging_change interval]	coefficient	std. err.			_			
_treated Mentee .7959861	.3000376	.2530396	1.19	0.236	1959108			
1.student_gender .6155954	.3645101	.1281071	2.85	0.004	.1134248			
_treated#student_gender Mentee#1 .2082992	    3004739	.2595829	-1.16	0.247	809247			
_cons	  2205594	.1944237	-1.13	0.257	6016228			
Note: One or more parame standard-error est								
Linear regression			R W P R A	eplicatio lald chi2( rob > chi -squared dj R-squa	obs = 183 ons = 499 (3) = 10.88 i2 = 0.0124 = 0.0595 ared = 0.0437 = 0.9866			

(Replications based on 13

clusters in school)								
Normal based		I	Observed	Boot	strap			
[95% conf. int	z_belonging_chaterval]							
	 	+-						
.1127303 .8	_ Ment	ated   tee	.4623718	.178	3918	2.59	0.010	
stude	ent_difficulte		.649071	. 203	39923	3.18	0.001	
.2492535 1	.048889	1						
_treated#stude			6311579	.281	18866	-2.24	0.025	
746007 .6	_(	cons	3553563	.199	3152	-1.78	0.075	
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.  Linear regression  Number of obs = 179 Replications = 500 Wald chi2(1) = 5.03 Prob > chi2 = 0.0250 R-squared = 0.0134							179 500 5.03 0.0250 0.0134 0.0078	
						Root MSE	= (	∂.9957
		(	Replicatio	ns base	ed on	12 cluster	s in s	chool)
z_sei_change	Observed   coefficient			P>	z		al-base f. inte	
	.2307811 1016074							
Linear regress	sion					Number of Replicatio Wald chi2( Prob > chi R-squared Adj R-squa Root MSE	ns = 3) = 2 = 0 = 0 red = 0	500 10.58 0.0142 0.0247 0.0080

(Replications based on 12 clusters in

school)					
	Observed	Bootstrap			
Normal-based z_sei_change			Z	P> z	[95% conf.
interval]					_
	.1752212	.1252313	1.40	0.162	0702277
.4206701 student_gender   .4610196	.1637878	.1516517	1.08	0.280	133444
student_difficultevent   .4410554	.1274524	.1600044	0.80	0.426	1861505
_cons   .0168387					
Linear regression			R W P R A	eplication lald chi2 erob > chi e-squared	(3) = 11.71 i2 = 0.0084 = 0.0244 ared = 0.0077
in school)		(Repl	ications	based or	n 12 clusters
	l Observed	Bootstrap			
Normal-based z_sei_change interval]	•	·	z	P> z	[95% conf.
	+				
_treated Mentee		.1659085	1.78	0.074	0290464
.6213028 1.student_gender .6312545	.3144989	.161613	1.95	0.052	0022566
_treated#student_gender Mentee#1 .135372		.197761	-1.28	0.202	639837
_cons .0045098	2028255	.1057853	-1.92	0.055	4101608

Replications = 500 Wald chi2(3) = 10.30 Prob > chi2 = 0.0162 R-squared = 0.0238 Adj R-squared = 0.0071 Root MSE = 0.9960

(Replications based on 12

```
clusters in school)
                                 Observed Bootstrap
Normal-based
                 z_sei_change | coefficient std. err. z P>|z|
[95% conf. interval]
______
                     treated
                                 .3454073 .1167408 2.96
                      Mentee
                                                             0.003
.1165996 .5742149
        student difficultevent |
                        Yes
                                 .2995774 .1933259 1.55
                                                             0.121
-.0793345
            .6784892
_treated#student_difficultevent |
                  Mentee#Yes -.2946773 .2869138 -1.03
                                                             0.304
-.8570181 .2676634
                        _cons | -.2186836 .1425066 -1.53
                                                             0.125
         *******
         ****** Resilience
         ***********
         foreach var of varlist z_brs_change z_wembs_change{
         bootstrap, reps(500) seed(7582) nodots: regress `var' _treated,
cluster(school)
. *** Outcomes adjusting for student gender and difficult live event
         bootstrap, reps(500) seed(7582) nodots: regress `var' _treated
student_gender student_difficultevent, cluster(school)
. *** Interaction between student gender and mentoring
         bootstrap, reps(500) seed(7582) nodots: regress `var'
_treated##student_gender, cluster(school)
. *** Interaction between difficult life event and mentoring
         bootstrap, reps(500) seed(7582) nodots: regress `var'
```

```
_treated##student_difficultevent, cluster(school)
  6. }
```

Linear regression

Number of obs = 177
Replications = 499
Wald chi2(1) = 0.81
Prob > chi2 = 0.3672
R-squared = 0.0077
Adj R-squared = 0.0021
Root MSE = 1.0036

## (Replications based on 13 clusters in school)

z_brs_change	•	std. err.		P> z	Normal [95% conf.	
	.1764986	.1957157	0.90	0.367 0.310	2820022	.5600943 .0894485

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 177
Replications = 499
Wald chi2(3) = 12.64
Prob > chi2 = 0.0055
R-squared = 0.0410
Adj R-squared = 0.0244
Root MSE = 0.9923

(Replications based on 13 clusters in

school)		, , ,			
  Normal-based	Observed	Bootstrap			
z_brs_change	coefficient	std. err.	z	P> z	[95% conf.
interval]					
_treated   .5967845	.1469239	.229525	0.64	0.522	3029367
student_gender	.2501715	.196973	1.27	0.204	1358885
<pre>.6362315 student_difficultevent  </pre>	- 299732	.2085401	-1.44	0.151	7084632
.1089991		.2005 101	2.11	0.131	.7001032
_cons   .232333	0578903	.1480758	-0.39	0.696	3481135

Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.

Linear regression

Number of obs = 177

Replications = 499
Wald chi2(3) = 11.51
Prob > chi2 = 0.0092
R-squared = 0.0250
Adj R-squared = 0.0081
Root MSE = 1.0006

in school)		(Repl:	ications	based on	13 clusters
Normal-based z_brs_change interval]	Observed   coefficient	std. err.			_
.9418703 1.student_gender .673341 _treated#student_gender	.2647586 .3963391	.1413301	2.80	0.005	.1193372
.3239698	3157389    2299816				
Note: One or more parame standard-error est					replicate;
Linear regression			Re Wa Pr R- Ac Rc	eplicatio ald chi2( rob > chi -squared dj R-squa oot MSE	
clusters in school)			(керт	cations	based on 13
[95% conf. interval]	0b _change	ficient sto			
	treated   Mentee   .2	369205 .24	482878	0.95	0.340

2497147	.7235557	1						
	ent_difficultevent Yes .0967476	2342502	.1688795	-1.39	0.165			
_	ent_difficultevent Mentee#Yes .4949851	0785094	.2926046	-0.27	0.788			
2841423	_cons .2835045	0003189	.1448105	-0.00	0.998			
Note: One or more parameters could not be estimated in 1 bootstrap replicate; standard-error estimates include only complete replications.								
Linear regress	sion	(Replication	Re Wa Pr R- Ad Ro	ld chi2(1 ob > chi2 squared j R-squar ot MSE	s = 500 ) = 0.68 = 0.4106 = 0.0091 red = -0.0007			
	   Observed Boot	 :strap			al-based			
z_wembs_ch~e	coefficient std.		P> z					
_treated _cons	2556067 .316   .0592671 .187			8644203 3072616				
Linear regress	sion		R W P R A	eplicatio ald chi2( rob > chi -squared dj R-squa	obs = 103 ns = 500 3) = 3.61 2 = 0.3070 = 0.0364 ared = 0.0072 = 0.9718			
school)					8 clusters in			
	Obser	rved Bootstra	p					
interval]	os_change   coeffic				_			
.3944564	_treated  3294							

.3762898	student_gender	.0028049	.1905571	0.01	0.988	3706801
	_difficultevent	.3274815	.1897384	1.73	0.084	044399
.410971	_cons	0734455	.2471562	-0.30	0.766	5578627
	-					
Linear	regression			R W P R A	eplicational chi2 chi2 chi2 chi2 chi2 chi2 chi2 chi2	obs = 103 ons = 488 (3) = 1.43 i2 = 0.6990 = 0.0397 ared = 0.0106 = 0.9702
			(Rep	lication	s based o	on 8 clusters
in scho	01)					
		Observed	Bootstrap			
Normal-	based z_wembs_change	l coofficient	ctd onn	-	ובו	[OF% conf
interva		Coefficient	stu. em.	۷	F> 2	[93% COIII.
	<del>-</del>	+				
.6561798	1.student_gender	9074822	.7978014		0.255 0.478	
_treated	d#student_gender Mentee#1 5	     .9993286	.8848869	1.13	0.259	7350178
.479454	_cons		.1963319			
Si	 ne or more parame tandard-error est regression	ters could not		ed in 12 ete repl N R W P R A	bootstraications lumber of leplications lald chi2 rob > chi -squared ldj R-squared loot MSE	ap replicates; obs = 103 ons = 500 (3) = 6.16 i2 = 0.1040 = 0.0627 ared = 0.0343 = 0.9585
				(кер	TICATION:	s based on 8

clusters in school)

		I	Observed	Bootstrap		
Normal-based				a+d	_	p. I=1
[95% conf. i	z_wembs_change .nterval] 	-+		sta. err.		P> z
		•				
-2.036017	_treated Mentee .3197361		8581405	.6009685	-1.43	0.153
2.030017	.3137301	I				
stu	dent_difficultevent. Yes		.1916651	.1939721	0.99	0.323
1885133	.5718435	ı	.1910031	.1939721	0.55	0.323
_treated#stu	dent_difficultevent Mentee#Yes		.8957595	.6927253	1.29	0.196
4619572		'			_,_,	
4320023	_cons		0178396	.2113114	-0.08	0.933
•						
	:*************************************				*******	******
•						
	:*************************************				*******	******
•						
. log close name:	<unnamed></unnamed>					
	C:\Users\quachjl\0 cuments\1_Projects\			-		10 Data
	.\Stata\03_matched_a			ZZ_NAISE EV	aluacion (	(10_Data
> sis_26 Mar	_					
<pre>log type: closed on:</pre>	text 26 Mar 2024, 20:36	:2	1			

```
name: <unnamed>
     log: C:\Users\quachjl\OneDrive - The University of
Melbourne\Documents\1_Projects\1_InProgress\2022_RAISE Evaluation\10_Data
Analysis\ALL\Stata\04_low_analysis_
> 26 Mar 2024.log
 log type: text
opened on: 26 Mar 2024, 20:36:21
        **********
. *** Part 2 - Outcomes for those low at baseline
 *********
 *********
 *** EPOCH
******
. bootstrap, reps(500) seed(7582) nodots: regress z_epoc_engagement_change
_treated if epoc_engagement_pre_low==1
Linear regression
                                              Number of obs =
                                                               55
                                                              500
                                               Replications =
                                              Wald chi2(1) = 0.50
                                               Prob > chi2 = 0.4815
                                               R-squared = 0.0081
                                               Adj R-squared = -0.0106
                                               Root MSE = 0.9418
_treated | .2021454 .2871517 0.70 0.481 -.3606616 _cons | .3350616 .1379358 2.43 0.015 .0647124
                                                           .7649524
                                                           .6054108
. bootstrap, reps(500) seed(7582) nodots: regress z_epoc_perser_change _treated
if epoc_perser_pre_low==1
                                              Number of obs =
Linear regression
                                                                80
                                               Replications =
                                                                500
                                              Wald chi2(1) = 0.17
                                               Prob > chi2 = 0.6822
                                               R-squared = 0.0021
                                               Adj R-squared = -0.0107
                                               Root MSE = 0.8357
                                  Normal-based
     Observed Bootstrap
z_epoc_pe~ge | coefficient std. err. z P>|z| [95% conf. interval]
```

<del>_</del>	.0751014 .2890785				284365 .4345677 .0122338 .5659231				
<pre> bootstrap, reps(500) seed(7582) nodots: regress z_epoc_optimism_change _treated if epoc_optimism_pre_low==1</pre>									
Linear regress	sion				Number of obs = 71 Replications = 500 Wald chi2(1) = 0.20 Prob > chi2 = 0.6580 R-squared = 0.0029 Adj R-squared = -0.0115 Root MSE = 0.9656				
_				P> z	Normal-based [95% conf. interval]				
	.1039187	.2347255			3561349 .5639723 .1266963 .7979251				
bootstrap, reps(500) seed(7582) nodots: regress z_epoc_connect_change _treated if epoc_connect_pre_low==1  Linear regression  Number of obs = 71 Replications = 500 Wald chi2(1) = 0.88 Prob > chi2 = 0.3492 R-squared = 0.0129 Adj R-squared = -0.0015 Root MSE = 1.0267									
z_epoc_co~ge	Observed coefficient	Bootstrap std. err.	z	P> z	Normal-based [95% conf. interval]				
_treated _cons	.2328696 .388034	.2487438 .1917534		0.349 0.043	2546594 .7203985 .0122042 .7638638				
<pre> ************ . ** Belonging . *********** . bootstrap, reps(500) seed(7582) nodots: regress z_belonging_change _treated if belonging_pre_low==1</pre>									

Linear regression

Replications = 500 Wald chi2(1) = 0.05 Prob > chi2 = 0.8245 R-squared = 0.0007 Adj R-squared = -0.0120 Root MSE = 1.0014

z_belongi~ge			P> z	Normal [95% conf.	-based interval]
_treated _cons	.2322982	0.22 2.38	0.824 0.017	4037802 .0813939	.5068121 .8429678

```
· *************
```

. \*\*\* Resilience . \*\*\*\*\*\*\*\*

. bootstrap, reps(500) seed(7582) nodots: regress z\_brs\_change \_treated if brs\_pre\_low==1

Linear regression

Number of obs = 65 Replications = 500 Wald chi2(1) = 0.65 Prob > chi2 = 0.4217 R-squared = 0.0070 Adj R-squared = -0.0088 Root MSE = 0.9417

z_brs_change		std. err.		P> z	Normal [95% conf.	-based interval]
_treated	.1680959	.2092132	0.80	0.422	2419544	.5781462
_cons		.1530266	3.53	0.000	.2403824	.8402356

\*\*\*\*\*\*\*\*\*\*\*\*\*

\*\*\*\*\*\*\*\*\*\*

. \*\*\* EPOCH . \*\*\*\*\*\*\*\*\*\*

. bootstrap, reps(500) seed(7582) nodots: regress z\_epoc\_engagement\_change
\_treated if epoc\_engagement\_pre\_low==0

Linear regression

Number of obs = 57
Replications = 497
Wald chi2(1) = 0.26
Prob > chi2 = 0.6105

R-squared = 0.0076 Adj R-squared = -0.0104 Root MSE = 0.8781

z_epoc_en~ge	-	std. err.	P> z	Normal [95% conf.	
	2672208	.5245761 .1140662	0.610 0.000	-1.295371 6452989	.7609294 1981675

Note: One or more parameters could not be estimated in 3 bootstrap replicates; standard-error estimates include only complete replications.

. bootstrap, reps(500) seed(7582) nodots: regress z\_epoc\_perser\_change \_treated
if epoc\_perser\_pre\_low==0

Linear regression

Number of obs = 117
Replications = 500
Wald chi2(1) = 0.37
Prob > chi2 = 0.5447
R-squared = 0.0030
Adj R-squared = -0.0057
Root MSE = 0.9420

z_epoc_pe~ge	•	std. err.	P> z	[95% conf.	-based interval]
_treated _cons	1024641	.1691495 .1052893	0.545	4339911 4540594	.2290629

. bootstrap, reps(500) seed(7582) nodots: regress z\_epoc\_optimism\_change
\_treated if epoc\_optimism\_pre\_low==0

Linear regression

Number of obs = 107
Replications = 500
Wald chi2(1) = 0.08
Prob > chi2 = 0.7812
R-squared = 0.0008
Adj R-squared = -0.0087
Root MSE = 0.8426

z_epoc_op~ge		std. err.	P> z	-	interval]
	.0473975	.1706783	0.781	2871259 6337915	.3819208

```
. bootstrap, reps(500) seed(7582) nodots: regress z_epoc_connect_change _treated
if epoc_connect_pre_low==0
Linear regression
                                         Number of obs = 107
                                         Replications = 500
                                         Wald chi2(1) = 3.88
                                         Prob > chi2 = 0.0489
R-squared = 0.0316
                                         Adj R-squared = 0.0223
                                         Root MSE = 0.8332
      ------
      Observed Bootstrap
                                             Normal-based
z_{epoc_co~ge} \mid coefficient std. err. z P>|z| [95% conf. interval]
_treated | .2981606 .1514022 1.97 0.049 .0014178 .5949034    _cons | -.4954897 .1023319 -4.84 0.000 -.6960565 -.2949229
***********
. ** Belonging
. bootstrap, reps(500) seed(7582) nodots: regress z_belonging_change _treated if
belonging pre low==0
                                         Number of obs = 102
Linear regression
                                         Replications = 500
                                         Wald chi2(1) = 2.33
                                         Prob > chi2 = 0.1266
R-squared = 0.0229
                                         Adj R-squared = 0.0131
                                         Root MSE = 0.8409
     Observed Bootstrap
                                             Normal-based
z_belongi~ge | coefficient std. err. z P>|z| [95% conf. interval]
______
***********
. *** Resilience
. bootstrap, reps(500) seed(7582) nodots: regress z_brs_change _treated if
brs_pre_low==1
                                        Number of obs =
Linear regression
                                                       65
```

500

Replications =

Wald chi2(1) = 0.65 Prob > chi2 = 0.4217 R-squared = 0.0070 Adj R-squared = -0.0088 Root MSE = 0.9417

z_brs_change	std. err.		P> z	Normal [95% conf.	 -based interval]
_treated _cons	.2092132	0.80 3.53	0.422 0.000	2419544 .2403824	.5781462 .8402356

. log close

name: <unnamed>

log: C:\Users\quachjl\OneDrive - The University of

 $\label{locality} {\tt Melbourne \setminus Documents \setminus 1\_Projects \setminus 1\_InProgress \setminus 2022\_RAISE\ Evaluation \setminus 10\_Data}$ 

Analysis\ALL\Stata\04\_low\_analysis\_

> 26 Mar 2024.log log type: text

closed on: 26 Mar 2024, 20:36:39

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```
name: <unnamed>
       log: C:\Users\quachjl\OneDrive - The University of
Melbourne\Documents\1_Projects\1_InProgress\2022_RAISE Evaluation\10_Data
Analysis\ALL\Stata\05_improved_anal
> ysis_26 Mar 2024.log
  log type: text
 opened on: 26 Mar 2024, 20:36:39
 *************
. ** Part 3 - Predicting those who have improved scores
 *************
. foreach var of varlist epoc_engagement_change epoc_perser_change
epoc_optimism_change epoc_connect_change belonging_change SSE_total_change
brs_change wembs_change {
            gen `var'_improve = .
  2.
             replace `var'_improve=1 if z_`var'>0.1 & z_`var'<.</pre>
  3.
             replace `var'_improve=0 if z_`var'<=0.1</pre>
  4.
  5. }
(424 missing values generated)
(44 real changes made)
(68 real changes made)
(424 missing values generated)
(98 real changes made)
(99 real changes made)
(424 missing values generated)
(69 real changes made)
(109 real changes made)
(424 missing values generated)
(89 real changes made)
(89 real changes made)
(424 missing values generated)
(92 real changes made)
(91 real changes made)
(424 missing values generated)
(56 real changes made)
(62 real changes made)
(424 missing values generated)
(80 real changes made)
(97 real changes made)
(424 missing values generated)
(49 real changes made)
(54 real changes made)
. foreach var of varlist epoc_engagement_change epoc_perser_change
epoc_optimism_change epoc_connect_change belonging_change SSE_total_change
brs change wembs change {
            tab `var'_improve _treated, col
logistic `var'_improve _treated
  2.
  3.
```

```
4. }
+----+
Kev
| frequency
| column percentage |
+-----+
epoc_engag
ement chan | Treatment indicator
ge_improve | Non - Men Mentee | Total
-----+----
     0 | 59 9 | 68
         62.11 52.94 | 60.71
      -----
     1 | 36 8 | 44
| 37.89 47.06 | 39.29
-----
   Total | 95 17 | 112
| 100.00 100.00 | 100.00
                                     Number of obs = 112
Logistic regression
                                     LR chi2(1) = 0.50
                                     Prob > chi2 = 0.4795
                                     Pseudo R2 = 0.0033
Log likelihood = -74.790987
epoc_engagement_change_improve | Odds ratio Std. err. z P>|z| [95%
conf. interval]
  -----
              _treated | 1.45679 .7720136 0.71 0.478
.5155965 4.116082
                _cons | .6101695
                              .1290432 -2.34 0.019
.4031179 .9235681
Note: _cons estimates baseline odds.
| Key
.
| frequency
| column percentage |
+----+
epoc_perse |
```

r change i | Treatment indicator

mprove | Non - Men Mentee | Total

0 | 52 47 | 99 54.74 46.08 | 50.25

			+
1	43	55	!
	45.26	53.92	49.75
Total	95	102	197
	100.00	100.00	100.00

Logistic regression

Number of obs = 197

LR chi2(1) = 1.48

Prob > chi2 = 0.2243

Log likelihood = -135.80899

Pseudo R2 = 0.0054

1.23856

-----

Note: \_cons estimates baseline odds.

<u> </u>	. —
Key	İ
	۱.
frequency	ı
column percentage	
+	٠+

epoc_optim ism_change _improve	   Treatment   Non - Men		Total
0	53	56	109
	63.86	58.95	61.24
1	30	39	69
	36.14	41.05	38.76
Total	83	95	178
	100.00	100.00	100.00

Logistic regression

Number of obs = 178

LR chi2(1) = 0.45Prob > chi2 = 0.5022

Pseudo R2 = 0.0019

Log likelihood = -118.62208

-----

epoc\_optimism\_change\_improve | Odds ratio Std. err. z P>|z| [95%

```
conf. interval]
                treated | 1.230357 .380615 0.67 0.503
         2.256073
                  _cons | .5660377 .129326 -2.49 0.013
.3617148 .8857771
Note: cons estimates baseline odds.
Key
 frequency
| column percentage |
+----+
epoc_conne |
ct_change_ | Treatment indicator
 improve | Non - Men Mentee |
       0 | 52 37 | 89
| 62.65 38.95 | 50.00
       1 | 31 58 |
| 37.35 61.05 |
                     58 |
                               50.00
   Total | 83 95 | 178
| 100.00 100.00 | 100.00
                                             Number of obs = 178
Logistic regression
                                             LR chi2(1) = 10.05
                                             Prob > chi2 = 0.0015
Log likelihood = -118.35414
                                             Pseudo R2 = 0.0407
epoc_connect_change_improve | Odds ratio Std. err. z P>|z| [95%
conf. interval]
   -----
               _treated | 2.629468 .8136802 3.12 0.002
1.433727 4.822469
                 _cons | .5961538 .1352742 -2.28 0.023
.3821298 .9300489
Note: _cons estimates baseline odds.
    frequency
```

## | column percentage |

belonging_ change_imp rove	   Treatment   Non - Men		Total
0	51	40	91
	57.95	42.11	49.73
1	37	55	92
	42.05	57.89	50.27
Total	88	95	183
	100.00	100.00	100.00

Logistic regression

Number of obs = 183 LR chi2(1) = 4.61 Prob > chi2 = 0.0318 Pseudo R2 = 0.0182

Log likelihood = -124.53837

\_\_\_\_\_\_

-----

Note: \_cons estimates baseline odds.

+		· <del>-</del>
	Key	I
-		
	fre	equency
	column	percentage
+		+

SSE_total_ change_imp rove	   Treatment   Non - Men		Total
0	53	9	62
	52.48	52.94	52.54
1	48	8	56
	47.52	47.06	47.46
Total	101	17	118
	100.00	100.00	100.00

Logistic regression

Number of obs = 118 LR chi2(1) = 0.00Prob > chi2 = 0.9716

Pseudo R2 = 0.0000

Log likelihood = -81.638126

SSE\_total\_change\_improve | Odds ratio Std. err. z P>|z| [95% conf.

.-----

\_treated | .9814815 .5154529 -0.04 0.972 .3506303

2.747355

\_cons | .9056604 .1804544 -0.50 0.619 .612861

1.338347

\_\_\_\_\_\_

Note: \_cons estimates baseline odds.

++	
Key	
frequency	
column percentage	
++	

brs\_change | Treatment indicator \_improve | Non - Men Mentee | Total 50 | 0 | 47 56.63 53.19 54.80 -----+---+ 1 | 36 44 | 80 43.37 46.81 Total | 83 94 | 177 | 100.00 100.00 | 100.00

Logistic regression

Number of obs = 177

LR chi2(1) = 0.21Prob > chi2 = 0.6467

Pseudo R2 = 0.0009

Log likelihood = -121.76437

\_\_\_\_\_

brs\_change\_improve | Odds ratio Std. err. z P>|z| [95% conf. interval]

------

\_treated | 1.148889 .3480615 0.46 0.647 .6344573

\_cons | .7659575 .169646 -1.20 0.229 .496225

1.182308

Note: \_cons estimates baseline odds.

+			+
K	ey		
Í			Ì
į	fre	equency	į
c	olumn	percent	tage
+			+

wembs_chan ge_improve	Treatment   Non - Men		Total
0	45	9	54
	51.72	56.25	52.43
1	42	7	49
	48.28	43.75	47.57
Total	87	16	103
	100.00	100.00	100.00

Logistic regression Number of obs = LR chi2(1) = 0.11Prob > chi2 = 0.7387Pseudo R2 = 0.0008

Log likelihood = -71.217098

interval]	ge_improve					-
	'					
2 420044	_treated	.8333334	.4564355	-0.33	0.739	.2848371
2.438041	_cons	.9333333	.2002468	-0.32	0.748	.6129307
1.421223						

Note: \_cons estimates baseline odds.

foreach var of varlist epoc\_engagement\_change epoc\_perser\_change epoc\_optimism\_change epoc\_connect\_change belonging\_change SSE\_total\_change brs\_change wembs\_c

```
> hange {
```

- logistic `var'\_improve student\_difficultevent if \_treated==1 2.
- logistic `var'\_improve student\_australia if \_treated==1
  logistic `var'\_improve student\_disability if \_treated==1 3.
- 4.
- logistic `var'\_improve i.student\_year if \_treated==1 5.
- logistic `var'\_improve student\_gender if \_treated==1 6.
- 7.

Logistic regression	Number of obs = LR chi2(1) = Prob > chi2 =	0.71
Log likelihood = -11.398187	Pseudo R2 =	0.0303
<pre>epoc_engagement_change_improve   Odds ratio</pre>		_
student_difficultevent   2.4 2.536139	0.83 0.407	
.3025002 19.04131		
_cons   .5 .4330127 .0915815		
Note: _cons estimates baseline odds.		
Logistic regression	Number of obs = LR chi2(1) =	1.06
Log likelihood = -11.221754	Prob > chi2 = Pseudo R2 =	
epoc_engagement_change_improve   Odds ratio Std. err. conf. interval]		_
student_australia   3.5 4.48609 .283824 43.16055		
_cons   .3333333 .3849002 .0346734	-0.95 0.341	
<pre>Note: _cons estimates baseline odds. note: student_disability != 0 predicts success perfectl     student_disability omitted and 2 obs not used.</pre>	у;	
Logistic regression	Number of obs = LR chi2(0) =	0.00
Log likelihood = -10.095175	Prob > chi2 = Pseudo R2 =	0.0000
epoc_engagement_change_improve   Odds ratio Std. err. conf. interval]		_
student_disability   1 (omitted) _cons   .6666667 .3513642	-0.77 0.442	

.2372937 1.872972	
Note: _cons estimates baseline odds.	
Logistic regression	Number of obs = 17 LR chi2(1) = 0.57
Log likelihood = -11.470256	Prob > chi2 = 0.4512 Pseudo R2 = 0.0241
epoc_engagement_change_improve   Odds ratio Std. err. conf. interval]	· · · · · · · ·
student_year   Year 9   2.666667 3.569417	
.1934678	-0.53 0.594
.2602315 2.161537	
Note: _cons estimates baseline odds.	
Logistic regression	Number of obs = 17 LR chi2(1) = 2.20 Prob > chi2 = 0.1377
Log likelihood = -10.652331	Pseudo R2 = 0.0937
	Pseudo R2 = 0.0937
Log likelihood = -10.652331  epoc_engagement_change_improve   Odds ratio Std. err. conf. interval]	Pseudo R2 = 0.0937
epoc_engagement_change_improve   Odds ratio Std. err. conf. interval]	Pseudo R2 = 0.0937  z P> z  [95%
epoc_engagement_change_improve   Odds ratio	Pseudo R2 = 0.0937  z P> z  [95%  1.37 0.172
epoc_engagement_change_improve   Odds ratio Std. err. conf. interval]  student_gender   5.599999 7.067672	Pseudo R2 = 0.0937  z P> z  [95%  1.37 0.172
epoc_engagement_change_improve   Odds ratio	Pseudo R2 = 0.0937  z P> z  [95%  1.37 0.172
epoc_engagement_change_improve   Odds ratio	Pseudo R2 = 0.0937  z P> z  [95%  1.37 0.172  -1.24 0.215  Number of obs = 102 LR chi2(1) = 0.04
epoc_engagement_change_improve   Odds ratio	Pseudo R2 = $0.0937$ z P> z  [95%  1.37 $0.172$ -1.24 $0.215$ Number of obs = $102$
epoc_engagement_change_improve   Odds ratio   Std. err. conf. interval]  student_gender   5.599999   7.067672  .4719541   66.44712  cons   .25   .2795085  .0279427   2.236723	Pseudo R2 = 0.0937  z P> z  [95%  1.37 0.172  -1.24 0.215  Number of obs = 102 LR chi2(1) = 0.04 Prob > chi2 = 0.8425 Pseudo R2 = 0.0003

student_difficultevent   2.35783					
1.949621		.3156095			
Note: _cons estimates baseli					
Logistic regression			LR cl	ni2(1)	= 102 = 0.00 = 0.9929
Log likelihood = -70.386925					= 0.0000
<pre>epoc_perser_change_improve   conf. interval]</pre>					_
student_australia					
2.453119 _cons   2.522364	1.166667	.4589642	0.39	0.695	.5396173
Note: _cons estimates baseli	ine odus.		Necesia	C . l	102
Logistic regression	ne odus.		LR cl Prob	ni2(1) > chi2	= 102 = 0.30 = 0.5862
	ne odus.		LR cl Prob	ni2(1) > chi2	= 0.30
Logistic regression  Log likelihood = -70.238827  epoc_perser_change_improve   conf. interval]	Odds ratio		LR cl Prob Pseud z	ni2(1) > chi2 do R2 	= 0.30 = 0.5862 = 0.0021
Logistic regression  Log likelihood = -70.238827  epoc_perser_change_improve   conf. interval] student_disability	Odds ratio		LR cl Prob Pseud z	ni2(1) > chi2 do R2 P> z	= 0.30 = 0.5862 = 0.0021 [95%
Logistic regression  Log likelihood = -70.238827	Odds ratio 1.32466 1.136264	.6967542 .2338404	LR cl Prob Pseud z 0.53	ni2(1) > chi2 do R2 P> z  0.593 0.535	= 0.30 = 0.5862 = 0.0021  [95% 
Logistic regression  Log likelihood = -70.238827	Odds ratio  1.32466  1.136264	.6967542 .2338404	LR cl Prob Pseud z 0.53	ni2(1) > chi2 do R2 P> z  0.593 0.535	= 0.30 = 0.5862 = 0.0021  [95% 
Logistic regression  Log likelihood = -70.238827	Odds ratio  1.32466  1.136264	.6967542 .2338404	LR ch Prob Pseud z 0.53 0.62	ni2(1) > chi2 do R2  P> z  0.593 0.535	= 0.30 = 0.5862 = 0.0021 
Logistic regression  Log likelihood = -70.238827  epoc_perser_change_improve   conf. interval] student_disability   3.71386cons   1.700805	Odds ratio  1.32466  1.136264	.6967542 .2338404	LR chember Probes Pseud	ni2(1) > chi2 do R2 P> z  0.593 0.535	= 0.30 = 0.5862 = 0.0021 
Logistic regression  Log likelihood = -70.238827	Odds ratio  1.32466  1.136264	.6967542 .2338404	LR check Prob Pseud Pseu	ni2(1) > chi2 do R2  P> z  0.593 0.535 er of obs	= 0.30 = 0.5862 = 0.0021 

conf. interval	nange_improve   .]					_
	student_year	2.0625				
23.12593	Year 9	3.636364	3.43229	1.37	0.171	.5717884
2.729808	_cons	.5	.4330127	-0.80	0.423	.0915815
Note: _cons es	timates baselin	e odds.				
Logistic regre				LR ch Prob	er of obs ni2(1) > chi2 do R2	= 0.7433
conf. interval	nange_improve					
st 1.930341	udent_gender	.8764881	.3530742	-0.33	0.743	.397977
2.305923		1.263158				
	timates baselin					
Logistic regre				LR ch Prob	ni2(1) > chi2	= 95 = 0.51 = 0.4767 = 0.0039
conf. interval	_change_improve .]					_
student_ .5929835 3.	difficultevent	1.346154	.5630874	0.71	0.477	
.3344814 1.	_cons	.6				
	timates baselin					

	egression hood = -62.83				LR chi2 Prob > Pseudo	2(1) chi2 R2	= 95 = 2.96 = 0.0852 = 0.0230
epoc_optim	 ism_change_ir rval]	mprove	Odds ratio	Std. err.	z	P> z	[95%
	 student_aus 1.129192	tralia	.4347826		-1.71	0.087	
	2.964679 						
Logistic r	s estimates legression		oaas.		LR chi2 Prob >	2(1) chi2	= 95 = 0.57 = 0.4509 = 0.0044
conf. inte	ism_change_ir rval]						_
	 student_disa	•					
• 5500075	3.98/238						
	1.019844	_cons	.6655641	.1449224	-1.87	0.062	
.4343563				.1449224	-1.87	0.062	
.4343563  Note: _con Logistic r	1.019844  s estimates begression	baseline		.1449224	Number LR chi2 Prob >	of obs 2(2) chi2	= 95 = 0.16 = 0.9214
.4343563  Note: _con Logistic r	1.019844  s estimates l	baseline		.1449224	Number LR chi2 Prob >	of obs 2(2) chi2	= 0.16
.4343563 Note: _con Logistic r Log likeli	1.019844s estimates begression hood = -64.23 ism_change_ir	baseline 378 	odds.	Std. err.	Number LR chiz Prob > Pseudo	of obs 2(2) chi2 R2 	= 0.16 = 0.9214 = 0.0013
.4343563 Note: _con Logistic r  Log likeli epoc_optim conf. inte	1.019844s estimates begression  hood = -64.23 ism_change_ir rval] student	378  mprove   +	odds.		Number LR chií Prob > Pseudo	of obs 2(2) chi2 R2 P> z	= 0.16 = 0.9214 = 0.0013

cons   .5 .4330127 .0915815	-0.80 0.423						
Note: _cons estimates baseline odds.							
Logistic regression	Number of obs = 95 LR chi2(1) = 1.05 Prob > chi2 = 0.3049						
Log likelihood = -63.793374	Pseudo R2 = 0.0082						
epoc_optimism_change_improve   Odds ratio Std. err. conf. interval]							
student_gender   1.547619 .6624523 .6688221 3.581109							
_cons   .5384615 .1784983 .2811798							
Note: _cons estimates baseline odds.							
Logistic regression  Log likelihood = -63.465906	Number of obs = 95 LR chi2(1) = 0.09 Prob > chi2 = 0.7700 Pseudo R2 = 0.0007						
epoc_connect_change_improve   Odds ratio Std. err. conf. interval]	· · · · · ·						
student_difficultevent   .8842105 .3722461 .3874415 2.017926							
_cons   1.666667 .496904 .929115 2.989703							
Note: _cons estimates baseline odds.							
Logistic regression	Number of obs = $95$ LR chi2(1) = $0.95$						
Log likelihood = -63.035492	Prob > chi2 = 0.3307 Pseudo R2 = 0.0075						
epoc_connect_change_improve   Odds ratio Std. err. conf. interval]							

	+-								
	•								
stud .2243856 1.									
.9403477 5.			1.035802						
Note: _cons es	timates baseline	odds.							
Logistic regre	ssion			LR chi	2(1)	= 95 = 1.58			
Log likelihood	= -62.720333			Pseudo	R2	= 0.2092 = 0.0124			
conf. interval	hange_improve   ]					_			
	nt_disability								
.9520053 2.	_cons	1.458165	.3172041	1.73	0.083				
	timates baseline								
Logistic regre	ssion			LR chi	2(2)	= 95 = 0.60 = 0.7405			
Log likelihood				Pseudo	R2	= 0.0047			
conf. interval	_					_			
	+-								
	student_year								
.3259132 9.	524954		1.516992						
.2401033 7.	•	1.384615	1.237776	0.36	0.716				
.2018352 4.	_cons   _cons   954538	1	.8164966	0.00	1.000				
Note: _cons es	Note: _cons estimates baseline odds.								
Logistic regre	ssion					= 95 = 0.03			

Log likelihood = -63.492563	Prob > chi2 Pseudo R2	= 0.0003
<pre>epoc_connect_change_improve   Odds ratio</pre>		_
+		
student_gender   1.079365 .4594565 .4686352 2.486004		
_cons   1.5 .4841229 .7968327 2.823679		
Note: _cons estimates baseline odds.		
Logistic regression	Number of obs LR chi2(1)	= 0.55
Log likelihood = -64.382871	Prob > chi2 Pseudo R2	= 0.0043
belonging_change_improve   Odds ratio Std. err. interval]	z P> z	[95% conf.
	0.74 0.450	6010452
student_difficultevent   1.363248 .5685812 3.087398	0.74 0.438	.0019455
_cons   1.181818 .3423523 2.08511		
Note: _cons estimates baseline odds.		
Logistic regression	Number of obs	= 95
Logistic regression	LR chi2(1)	= 0.40
Log likelihood = -64.457416	Prob > chi2 Pseudo R2	
belonging_change_improve   Odds ratio Std. err. interval]		_
student_australia   1.359195 .6546484 3.493478		
_cons   1.090909 .4553712 2.472294		
	·	<b> </b>
Note: _cons estimates baseline odds.		

Logistic regression					Number of obs = 95 LR chi2(1) = 2.04 Prob > chi2 = 0.1534		
Log likelihood					eudo R2		
belonging_chang interval]						_	
 student_d 9.017308	isability	2.35742	1.613633	1.25	0.210	.6163069	
1.932264	_cons	1.267218	.2727598	1.10	0.271	.8310672	
Note: _cons est	imates basel	ine odds.					
Logistic regres	sion			LR	nber of obs chi2(2) ob > chi2	= 0.00	
Log likelihood	= -64.658808	1			eudo R2		
belonging_chang interval]			Std. err.				
	dent_year   Year 8	1.03125	.834805	0.04	0.970	.2110132	
5.039858 5.451992	Year 9	1.038462	.8785974	0.04	0.964	.1977997	
5.957371	_cons	1.333333	1.01835	0.38	0.706	.2984165	
Note: _cons est	imates basel	ine odds.					
Logistic regres	sion	LR	nber of obs chi2(1)	= 0.28			
Log likelihood = -64.519065					ob > chi2 eudo R2		
belonging_chang interval]						_	
	+-						

	nt_gender	.8	.337046	-0.53	0.596	.350326
1.82687 2.926426	_cons	1.5625	.5002441	1.39	0.163	.8342622
Note: _cons est	imates baselir	e odds.				
Logistic regres	sion			LR	chi2(1)	s = 17 = 0.03
Log likelihood				Pse	eudo R2	= 0.8576 = 0.0014
SSE_total_chang interval]						_
 student_diffi						
6.110696 4.954538	_cons	1	.8164966	-0.00	1.000	.2018352
Note: _cons est	imates baselir	e odds.				
Logistic regres	sion			LR	chi2(1)	s = 17 = 1.06 = 0.3022
Log likelihood				Pse	eudo R2	= 0.0453
SSE_total_chang interval]	e_improve   Oc					
43.16055	australia					
3.20451			.3849002			
Notes and other	:					
Note: _cons est	imates baselir	ie oaas.				
Logistic regres	sion			LR	chi2(1)	s = 17 = 0.01 = 0.9294
Log likelihood				Pse	eudo R2	= 0.0003
SSE_total_chang	e_improve   Od	lds ratio	Std. err.	Z	P> z	[95% conf.

interval]	+					
	·					
student_ 21.87011	_disability	1.142857	1.721075	0.09	0.929	.0597218
2 412041	_cons	.875	.4528555	-0.26	0.796	.3172995
2.412941						
Noto: cons os	timatos basolir	o odde				
Note: _cons es	stimates baselir	ie odus.				
Logistic regre	ession			LR	nber of obs	= 0.28
Log likelihood	d = -11.613603				ob > chi2 eudo R2	
interval]	nge_improve   Od					
	+					
st	tudent_year					
6.861766	Year 9	.5	.6681531	-0.52	0.604	.0364338
2.850929	_cons	1	.5345225	0.00	1.000	.3507629
Note: _cons es	stimates baselir	ne odds.				
Logistic regre	ession			Nun	nber of obs	= 17
8				LR	chi2(1)	= 2.20
Log likelihood	d = -10.652331				ob > chi2 eudo R2	
_						
interval]	nge_improve   Oc					-
stud 66.44712	dent_gender	5.599999	7.067672	1.37	0.172	.4719541
2.236723			.2795085			
Note: _cons es	stimates baselir	e odds.				
Logistic regre	ession				nber of obs chi2(1)	
Log likelihood	d = -62.811085			Pro	ob > chi2 eudo R2	= 0.0380

brs_change_improve   Odds ratio Std. err. interval]	· · · -
student_difficultevent   .4197531 .1776869 -29623179	.05 0.040 .1830919
_cons   1.35 .3982775 1.	.02 0.309 .7572023
2.406886	
Note: _cons estimates baseline odds.	
Logistic regression	Number of obs = 94 LR chi2(1) = 0.69
Log likelihood = -64.619437	Prob > chi2 = 0.4063 Pseudo R2 = 0.0053
<pre>brs_change_improve   Odds ratio Std. err. z interval]</pre>	
student_australia   .6666667 .326315 -0.83 1.739997	0.407 .2554283
_cons   1.2 .5138093 0.43 2.777435	
Note: _cons estimates baseline odds.	
Logistic regression	Number of obs = 94 LR chi2(1) = 0.01
Log likelihood = -64.961377	LR chi2(1) = 0.01 Prob > chi2 = 0.9399 Pseudo R2 = 0.0000
<pre>brs_change_improve   Odds ratio Std. err. z interval]</pre>	
student_disability   .962568 .4877651 -0.08 2.598744	
_cons   .8839296 .1900273 -0.57 1.347129	
Note: _cons estimates baseline odds.	
Logistic regression	Number of obs = 94

LR chi2(2) = 3.73Prob > chi2 = 0.1549Log likelihood = -63.09913Pseudo R2 = 0.0287\_\_\_\_\_\_ brs\_change\_improve | Odds ratio Std. err. z P>|z| [95% conf. interval] ----student\_year | Year 8 | 6.214286 6.911179 1.64 0.100 .7026398 54.96038 Year 9 | 5.25 5.987291 1.45 0.146 .5615883 49.07955 \_cons | .1666667 .1800206 -1.66 0.097 .0200653 1.384368 Note: cons estimates baseline odds. Number of obs = Logistic regression 94 LR chi2(1) = 0.01Prob > chi2 = 0.9364Pseudo R2 = 0.0000Log likelihood = -64.96103----brs\_change\_improve | Odds ratio Std. err. z P>|z| [95% conf. \_\_\_\_\_\_ student\_gender | 1.033835 .4309992 0.08 0.936 .4566573 2.340516 \_cons | .8636364 .2704797 -0.47 0.640 .4674613 1.595571 \_\_\_\_\_\_ Note: \_cons estimates baseline odds. Number of obs = 16 Logistic regression LR chi2(1) = 0.43Prob > chi2 = 0.5125Log likelihood = -10.750557Pseudo R2 = 0.0196wembs\_change\_improve | Odds ratio Std. err. z P>|z| [95% conf. ----student\_difficultevent | 2 2.144761 0.65 0.518 .2444658

16.36221

2.729808	_cons					.0915815
Note: _cons esti	mates baseline c	dds.				
Logistic regress	ion				LR chi2(1	obs = 16 ) = 0.80 i2 = 0.3723
Log likelihood =	-10.567107				Pseudo R2	= 0.0363
<pre>wembs_change_imp interval]</pre>		.o St	td. err.	z	P> z	[95% conf.
student_austr	alia	3 3	.872983	0.85	0.395	.2389048
3.20451	cons   .333333					
Note: _cons esti	mates baseline c					
Logistic regress	ion				LR chi2(1	obs = 16 ) = 0.04 i2 = 0.8494
Log likelihood =					Pseudo R2	= 0.0016
wembs_change_impinterval]	rove   Odds rati	.o St	td. err.	Z	P> z	
	lity   1.33333					.068607
2.161537	cons   .7					
Note: _cons esti	mates baseline c					
Logistic regress	ion					obs = 16 ) = 0.79
Log likelihood =					Pseudo R2	i2 = 0.3748 = 0.0359
<pre> wembs_change_imp interval]</pre>	rove   Odds rati	o St	td. err.	z	P> z	[95% conf.

student_year					
Year 9   45.19183	3.2	4.322962	0.86	0.389	.2265896
_cons	.625	.3563048	-0.82	0.410	.2044657
1.910467					
Note: _cons estimates base.	line odds	5.			
Logistic regression				LR chi2(1	obs = 16 ) = 0.04
Log likelihood = -10.94416				Pseudo R2	i2 = 0.8381 = 0.0019
<pre>wembs_change_improve   Odd: interval]</pre>					
student_gender	1.25	1.369306	0.20	0.839	.1460379
10.69928cons   .0	5666667	6085806	-0.44	0.657	.1113965
3.989752		.0003000	••••	0.037	1113303
Note: _cons estimates base	line odds	5 <b>.</b>			
		•			
<pre> log close     name: <unnamed></unnamed></pre>					
log: C:\Users\quac					\10 P-+-
<pre>Melbourne\Documents\1_Projo Analysis\ALL\Stata\05_impro</pre>			ZZ_KAISE	Evaluation	\10_Data
> ysis_26 Mar 2024.log					
log type: text	20.26.41				
closed on: 26 Mar 2024, 2	41 : ٥٥: ط2 				

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name: <unnamed>

log: C:\Users\quachjl\OneDrive - The University of

Melbourne\Documents\1\_Projects\1\_InProgress\2022\_RAISE Evaluation\10\_Data

Analysis\ALL\Stata\06\_outcomes\_anal

> ysis\_26 Mar 2024.log

log type: text

opened on: 26 Mar 2024, 20:36:41

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. \*\*\* Part 4 - Looking at differences at follow up

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. \*\*\* EPOCH \*\*\*\*\*

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. \*\*\* Pattern of missing

. mdesc epoc\_engagement\_post epoc\_perser\_post epoc\_optimism\_post epoc\_connect\_post epoc\_happy\_post SSE\_productivity\_post SSE\_belong\_post SSE\_aspiration\_post SSE\_total\_p

> ost

Variable	Missing	Total	Percent Missing
epoc_engag~t   epoc_perse~t   epoc_optim~t   epoc_conne~t   epoc_happy~t   SSE_produc~t   SSE_belong~t   SSE_aspira~t   SSE total ~t	302 227 246 246 302 296 296 296 296	424 424 424 424 424 424 424 424 424	71.23 53.54 58.02 58.02 71.23 69.81 69.81 69.81
SSE_total_~t	296 	424 	69.81

. \*\*\* EPOC Engagement

. mean epoc\_engagement\_pre, over(student\_mentee)

Mean estimation Number of obs = 312

Mean Std. err. [95% conf. interval]

-----+

c.epoc\_engagement\_pre@student\_mentee | 2.996075 .0579047 2.88214 No

```
3.110009
                       Yes | 2.736111 .1687208 2.404132
3.06809
. mean epoc_engagement_post, over(student_mentee)
Mean estimation
                                                  Number of
obs = 122
                           Mean Std. err. [95% conf.
interval]
c.epoc_engagement_post@student_mentee |
                        No | 3.005263 .0922515 2.822627
3.187899
                        Yes | 2.993827 .1617219 2.673656
3.313998
. foreach var of varlist epoc engagement post {
 2. regress `var' student mentee
 3. regress `var' student_mentee epoc_engagement_pre_low
 4. regress `var' student_mentee student_gender
 5. regress `var' student_mentee student_difficultevent
 6. regress `var' student_mentee##epoc_engagement_pre_low
student difficultevent student gender
 7. }
Source | SS df MS Number of obs = 122
------ F(1, 120) = 0.00
epoc_engagem~t | Coefficient Std. err. t P>|t| [95% conf. interval]
student_mentee | -.011436 .1933899 -0.06 0.953 -.3943345 .3714625
_cons | 3.005263 .0909779 33.03 0.000 2.825133 3.185393
    Source | SS
                       df MS Number of obs =
                                                       112
12.73
----- Adj R-squared = 0.1745
```

Total	85.4056919	111	.769420648	Root M	SE	=	.79699
epoc_engagem	ent_post   Co					[9	95% conf.
studen .4995436	t_mentee	.0765179	.2134373	0.36	0.721		3465078
epoc_engagement 464494		3.352937			0.000		.071739
3.565426							
Source	SS	df	MS 		of obs 19)		122 1.02
•	1.59570117			Prob >	F	=	0.3625
Residual	92.7644991				red squared		
Total				-	SE		.88291
epoc_engagem~t	Coefficient	t Std. er	r. t	P> t	[95%	conf.	interval]
student_mentee	.0946855	.206369	4 0.46	0.647	3139	9466	.5033175
student_gender							
_cons	3.09091	.108606.	2 28.46 		2.8/5	5859 	3.305961
Source	SS		MS		of obs		
Model	4.5862779				19) F		
Residual	89.7739224	119	.754402709	R-squa	red	=	0.0486
				Adj R-	squared	=	0.0326
TOTAL	94.3602003	121	.//9030300	KOOL M	3E	=	.00030
epoc_engageme interval]	<del>_</del>					_	
student .4671013	_mentee	.0842286	.1933603	0.44	0.664	29	986441
student_difficu0783461	ltevent	. 3983437	.1616068	-2.46	0.015	71	183413
3.376436	_cons   3	3.160408	.1090996	28.97	0.000	2.9	944379

-----

504. 66	, 33	•	5	- (-		
	+			F(5, 106)	=	5.39
Model	•			Prob > F	=	0.0002
Residual	68.1037502 10	6.6	542488209	R-squared		0.2026
	+			Adj R-squared	=	0.1650
Total	85.4056919 11	1 .	769420648	Root MSE	=	.80155
	epoc_engagement_p	ost	Coefficie	nt Std. err.	t	P> t
[95% conf.			•			
			+			
	student_men	tee	I			
		es		1 .3759693	0.49	0.628
5627347	.9280567	<b>C</b> 3	, 10200.	.5755055	0.45	0.020
.3027347	. 3200307		I			
	epoc_engagement_pre_	low	! 			
	Low at baseli		ı  705244	7 .1694147	-4.16	0.000
-1.041126		iie	/65244	/ .109414/	-4.10	0.000
-1.041126	3693635		ı			
-4	-#	1	 			
student_mente	e#epoc_engagement_pre_			161226	0.43	0.000
070075	Yes#Low at baseli	ne	057929	7 .461226	-0.13	0.900
972355	.8564956					
	student_difficultev	ent	1750384	4 .1615036	-1.08	0.281
495235	.1451583		_			
	student_gen	der	100656	2 .1636385	-0.62	0.540
4250854	.2237729					
	_c	ons	3.42761	7 .1288597	26.60	0.000
3.17214	3.683094					

df MS

Number of obs =

112

Number of obs =

Mean estimation

Source | SS

<sup>.</sup> mean epoc\_perser\_pre, over(student\_mentee)

424				abc. or obs
interval]	   		Std. err.	[95% conf.
c.epoc_perser_pre@student_mentee No 3.301853 Yes		3.192029 3.103604	.0558734	3.082205 2.95329
3.253918			.0704727	

<sup>. \*\*</sup> EPOC Perservence

. mean epoc\_perser\_post, over(student\_mentee) Mean estimation Number of obs = 197 Mean Std. err. [95% conf. interval] c.epoc\_perser\_post@student\_mentee | No 3.318421 .0907572 3.139435 3.497407 Yes | 3.310458 .1030719 3.107185 3.51373 \_\_\_\_\_\_ . foreach var of varlist epoc perser post { 2. regress `var' student\_mentee 3. regress `var' student\_mentee epoc\_perser\_pre\_low regress `var' student\_mentee student\_gender
 regress `var' student\_mentee student\_difficultevent 6. regress `var' student mentee##epoc perser pre low student difficultevent student\_gender 7. } ------ Adj R-squared = -0.0051 Root MSE = Total | 183.005005 196 .933699006 .96875 -----epoc\_perser\_~t | Coefficient Std. err. t P>|t| [95% conf. interval] student mentee | -.0079635 .138128 -0.06 0.954 -.2803801 \_cons | 3.318421 .0993914 33.39 0.000 3.122401 3.514441 Source SS df MS Number of obs = | Model | 45.8558274 | 2 22.9279137 | Prob > F | = 0.0000 | Residual | 137.149178 | 194 .706954525 | R-squared | = 0.2506 ------ Adj R-squared = 0.2428 .84081

epoc\_perser\_post | Coefficient Std. err. t > |t| [95% conf.

interval]							
		c22 110	20274	0.20	0.762	272	011
student_r .2001864	mentee  0363	623 .115	99374 -	0.30	0.762	272	911
epoc_perser_pi 7421182	re_low  9828	007 .122	20334 -	8.05	0.000	-1.223	483
3.930264	_cons   3.732	232 .100	94081 3	37.17	0.000	3.5	342
Source	SS	df	MS 		mber of obs 2, 194)		197 0.95
Model	1.77762218	2	.888811092	•	ob > F		0.3880
Residual	!		.934161768		squared		0.0097
	+			-	j R-squared		-0.0005
Total	183.005005	196	.933699006	S Roo	ot MSE	=	.96652
epoc_perser_~	t   Coefficient	Std. err	. t	P> 1	t  [95%	conf.	interval]
student_mente	.0372185	.1416559	0.26	0.79	93242	1649	.3166019
student_gender	- 1955399	.1418757	-1.38			3568	.084277
_cons	3.386345	.1107344	30.58	0.00	3.16	7948	3.604743
Source	SS	df	MS	Nur	mber of obs	; =	197
	+			•	2, 194)	=	1.65
Model	!				ob > F		0.1953
Residual	179.949844	194	.927576514		squared		0.0167
To+a1	t   102 aaeaae	106	022600006	-	j R-squared ot MSE		0.0066
IOCAL	183.005005	190	. 933099000	) KO	OL MSE	=	.96311
epoc_pei	rser_post   Coef	ficient S	Std. err.	-	t P> t	[9	5% conf.
interval]							
	nt_mentee   .0	198807	.1381792	0.3	14 0.886	2	526455
<del></del>	cultevent  2	519244	.1388832	-1.8	31 0.071	5	258393
.0219905	l 2	416520	1126402	20.	22 0 000	2	104265
3.638713	_cons   3.	416539	.1126492	30.	33 0.000	3.	194365
Source	!						
Model	+   45.8909824	5 (	9.17819649	· F(: ) Pr	op > E	=	0.0000
Residual	137.114023	191	717874465	R-:	squared	=	0.2508
	+			Ad	j R-squared	<b>=</b>	

Tot	al   183.005005	196	.933699006	Root MSE	=	.84727		
[95% conf.	epoc_perser_po interval]							
3478321	student_ment Ye .2866362	ee   s	030598	.1608316	-0.19	0.849		
-1.339642	epoc_perser_pre_l Low at baseling 6196429		9796427	.1825129	-5.37	0.000		
student_me	ntee#epoc_perser_pre_lontee#epoc_perser_lontee#epoc_perser_lontee#epoc_perser_pre_lonteepoc_perser_pre_lonteepoc_perser_		.0007853	.2475392	0.00	0.997		
2474424	student_difficulteve	nt	.0043254	.1276414	0.03	0.973		
		er	0279541	.1267148	-0.22	0.826		
3.496257	_co 3.981599	ns	3.738928	.1230296	30.39	0.000		
** EPOC Optimism . mean epoc_optimism_pre, over(student_mentee)  Mean estimation 424  Number of obs =								
interval]			Mean	Std. err.	[95%	conf.		
c.epoc_opt	imism_pre@student_ment N	ee	3.066425	.0608255	2.946	867		
3.096076	Ye	s	2.927365	.0858326	2.758	653		

. mean epoc\_optimism\_post, over(student\_mentee)

```
Mean Std. err. [95% conf.
interval1
 .-----
c.epoc_optimism_post@student_mentee |
                No | 3.205823 .1012877 3.005936
3.40571
                Yes | 3.194737 .1026782 2.992106
3.397368
  . foreach var of varlist epoc_optimism_post {
 2. regress `var' student_mentee
 3. regress `var' student_mentee epoc_optimism_pre_low
 4. regress `var' student_mentee student_gender
 5. regress `var' student mentee student difficultevent
 6. regress `var' student_mentee epoc_optimism_pre_low student_difficultevent
student_gender
 7. }
                           Number of obs = 178
F(1, 176) = 0.01
   Source SS df MS
   -----
Residual | 163.971499 176 .931656242
                           Adj R-squared = -0.0056
   epoc_optimis~t | Coefficient Std. err. t P>|t| [95% conf. interval]
3.414913
178
                                     1/8
42.81
                          Prob > F = 0.0000
R-squared = 0.3285
   Model | 53.8688497 2 26.9344249 Prob > F
 Residual | 110.108094 175 .629189106
-----
                           Adj R-squared = 0.3208
   epoc_optimism_post | Coefficient Std. err. t P>|t| [95% conf.
-----
student_mentee | .0169899 .1192176 0.14 0.887 -.2182995
.2522793
```

epoc_optimism_p	ore_low   -1.1	123768	.1214564	-9.25	0.000	-1.3	63476
3.834195	_cons   3.6	539083	.0988601	36.81	0.000	3.4	43972
Source	SS	df	MS	Numbe F(2,	r of obs 175)		178 0.59
Model   Residual	162.887724	175	.930786997	Prob R-squ	> F <sup>°</sup> ared	= =	0.5581 0.0066
Total	163.976943				-squared MSE		-0.0047 .96477
epoc_optimis~t	Coefficient	Std. er	·. t	P> t	[95%	conf.	interval]
student_mentee student_gender _cons	•	.1493806 .1493806 .1172766		0.852 0.282 0.000		094	.1336292
Source	SS	df	MS		r of obs		178
Model   Residual		175	.885335223	Prob R-squ	175) > F ared -squared	=	5.11 0.0070 0.0551 0.0444
Total	163.976943			•	MSE	=	.94092
epoc_optimi	ism_post   Coef	fficient	Std. err.	t	P> t		5% conf.
 student .3075935	_mentee   .0	275608	.1418884	0.19	0.846	2	524719
student_difficu 1736171	ultevent  4	1541483	.142141	-3.20	0.002	7	346796
3.625856	_cons   3						
Source	SS	df	MS	Numbe F(4.	r of obs 173)	=	178 23.78
Model   Residual	58.1796182 105.797325	4 173	14.5449046 .611545231	Prob R-squ	> F ared	=	0.0000 0.3548
Total	163.976943	177	.926423408	Adj R Root	-squared MSE	=	0.3399 .78201
	ism_post   Coe						

epoc_opt: .74085	imism_pre	.6452169	.0484539	13.32	0.000	.54	95838	
1.63961	_cons	1.286553	.1788813	7.19	0.000	.93	34968	
. regress epoc epoc_optimism	c_optimism_pos _pre	t student_r	mentee##stude	ent_diff	icultev	ent		
Source	•	df	MS		of obs		178 49.67	
Model Residual	87.6548886 76.3220547	173	.441167946	Prob >	→ F ared	= =	0.0000 0.5346	
Total	+   163.976943				squared ISE		0.5238 .6642	
[95% conf. :	interval]	<del>_</del> -	Coefficier					
1311943		ent_mentee Yes	1356707	7 .135	52056	1.00	0.317	
	student_diff	icultevent Yes	     .0351758	3 .153	30671	0.23	0.819	
2669436	.3372953		' 					
student_mente	_	icultevent Yes#Yes	  1955897	7 .20	32127	-0.97	0.335	
5945421			ļ					
.5532359		timism_pre			18641			
.8584996		_	1.229999		38218		0.000	
<pre> ** EPOC Connectiveness . mean epoc_connect_pre, over(student_mentee)</pre>								
Mean estimation	on					Number	of obs =	
interval]		 	Mean S	Std. err	·· [	95% con	if.	

```
c.epoc_connect_pre@student_mentee |
                    No |
                          3.838164 .0558608 3.728365
3.947964
                    Yes | 3.689189 .0755741 3.540642
3.837737
. mean epoc_connect_post, over(student_mentee)
Mean estimation
                                             Number of obs =
178
                           Mean Std. err. [95% conf.
interval]
 -----
c.epoc_connect_post@student_mentee |
                     No | 3.893574 .088029 3.719853
4.067296
                     Yes | 4.02807 .0839341 3.86243
4.193711
______
. foreach var of varlist epoc_connect_post {
 2. regress `var' student_mentee
 3. regress `var' student_mentee epoc_connect_pre_low
 4. regress `var' student_mentee student_gender5. regress `var' student_mentee student_difficultevent
 6. regress `var' student_mentee##epoc_connect_pre_low student_difficultevent
student gender
 7. }
    Source | SS df
                              MS
                                   Number of obs =
                                                   178
                                   F(1, 176)
                                              =
                                                   1.22
  0.2710
                                   Prob > F
                                   R-squared
                                                  0.0069
                                             =
                                   Adj R-squared = 0.0012
    Total | 116.453028 177 .65792671
                                   Root MSE
                                                  .81062
______
epoc_connect~t | Coefficient Std. err. t P>|t| [95% conf. interval]
student mentee .1344959
                    .1217948 1.10 0.271 -.1058704
                                                  .3748622
    _cons | 3.893574 .0889776 43.76 0.000 3.717974 4.069175
```

Source			MS		er of obs		178
Model   Residual	40.7444071 75.7086205	2 175	20.3722036	Prob R-squ	175) > F uared R-squared	= =	0.0000 0.3499
-	116.453028			Root	MSE	=	.65774
interval]	t_post   Coeffi					_	
.3758284	mentee   .180		0989403	1.82	0.070	014	7112
epoc_connect_p 7696092	ore_low  968	5458 .	1007982	-9.61	0.000	-1.16	7483
4.416017	_cons   4.2	5532 .	0814226	52.26	0.000	4.09	4624
Source	SS	df	MS	Numbe	er of obs	=	178
+				F(2,	175)	=	1.35
	1.76882434 114.684203				> F uared	=	0.2620 0.0152
				Adj F	R-squared	=	0.0039
Total					MSE		.80953
	Coefficient						interval]
student mentee	.0977007	.125343	5 <b>0.</b> 78	0.437	1496	5788	.3450801
	.1522993						
_cons	s   3.842196 	.098405	4 39.04		3.647		4.03641
Source	SS	df	MS		er of obs		
Hodel	5.62201552	2	2 81100776	F(2,	175)	=	4.44 a a132
Residual	110.831012	175	.633320069	R-sqi	uared	=	0.0483
+ 	446 452020	477		Adj F	R-squared	=	0.0374
	116.453028						
interval]	nect_post   Coef						
	nt_mentee   .1	627214	.1200065	1.36	0.177	0	741249
.3995676 student_diffic 0944133	cultevent  3	316812	.1202201	-2.76	0.006	5	689491

4.227353	_cons   4.0		.1002777			831534
Source	SS		MS	Number of		
Model Residual	43.1924839   73.2605437	5 172	8.63849678 .425933394		=	0.0000 0.3709
	116.453028					
[95% conf. in	epoc_connect terval]					
1417837	student_m	entee   Yes	.1102022	.127662	0.86	0.389
-1.290745 -			9905661	.1520779	-6.51	0.000
_	e#epoc_connect_pr Yes#Low at base .4724385		.0718034	.2029712	0.35	0.724
	student_difficult .0847471	event	1169887	.1022041	-1.14	0.254
.0097716 .4		ender	.2116169	.1022596	2.07	0.040
4.039852 4	.440306	_cons	4.240079	.1014395	41.80	0.000
· · ** EPOC Hap · mean epoc_ha	oiness appy_pre, over(st	udent_m	entee)			
Mean estimation 312	on				Number c	of obs =
interval]			Mean Sto		_	
-	ore@student_mente					

```
No | 3.38587 .0609724
                                  3.265899
3.50584
               Yes | 3.331019 .1683934 2.999684
3.662353
. mean epoc_happy_post, over(student_mentee)
                                    Number of obs =
Mean estimation
122
                   Mean Std. err. [95% conf.
interval]
-----
c.epoc_happy_post@student_mentee |
                No 3.45 .09169 3.268475
3.631525
                Yes | 3.067901 .1754588
                                   2.720534
3.415268
______
. foreach var of varlist epoc_happy_post {
 2. regress `var' student_mentee
 3. regress `var' student_mentee epoc_happy_pre_low
 4. regress `var' student_mentee student_gender
 5. regress `var' student_mentee student_difficultevent
 6. regress `var' student_mentee##epoc_happy_pre_low student_difficultevent
student_gender
 7. }
                             Number of obs = 122
= 3.81
  Source | SS df MS
------ F(1, 120) =
                                     = 0.0533
                 1 3.069579 Prob > F
   Model | 3.069579
  .
------ Adj R-squared = 0.0227
   epoc_happy_p~t | Coefficient Std. err. t P>|t| [95% conf. interval]
student_mentee | -.3820988 .1957622 -1.95 0.053 -.7696943 .0054967
   _cons | 3.45 .0920939 37.46 0.000
                                  3.267661
                                         3.632339
```

Total   91.468192		.824037765	Adj R Root I	-squared MSE	= =	0.4199 .69139
epoc_happy_post   Coefficient interval]		err.		t  [	95% co	nf.
 student_mentee  2962963 .070472	.185	50526 -1.	.60 0.	112	663064	6
epoc_happy_pre_low   -1.138007 8727239	.133	38484 -8.	.50 0.0	900 -	1.4032	9
_cons   3.905203 4.081345	.088	38724 43	.94 0.0	000 3 	3.72906	1
Source   SS	df	MS		r of obs		122
Model   3.0814363			Prob		=	1.90 0.1546
Residual   96.6747682	119	.81239301	R-squa Adi R	ared -squared	=	0.0309 0.0146
Total   99.7562045	121	.824431442	Root I	•	=	.90133
epoc_happy_p~t   Coefficient St	d. err	·. t	P> t	 [95%	conf.	interval]
student_mentee  3912545 .			0.066			.0259011
=== ;	108716	3 0.12 5 31.05	0.904 0.000		3784 3074	.3699229 3.662147
Source   SS	df	MS		r of obs		122 11 25
Model   15.8594879	2	7.92974396	Prob	> F	=	0.0000
Residual   83.8967165						
Total   99.7562045						
<pre>epoc_happy_post   Coeffic interval]</pre>	ient	Std. err.	t	P> t	[95	% conf.
+						
student_mentee  2222	957	.1869238	-1.19	0.237	59	24236
.1478322 student_difficultevent  6654	137	.1562273	-4.26	0.000	97	47594
356068 _cons   3.709 3.917998	161	.105468	35.17	0.000	3.5	00324

Sour	•	SS	df	MS	Number of		= 112 = 16.95	
Mode	el	40.642047 50.826145	5 106	8.12840939	Prob > F R-squared		= 0.0000 = 0.4443 = 0.4181	
	•	91.468192	111	.824037765	Root MSE		= .69245	
[95% conf.	inte	 epoc_happy_pos	t   (	Coefficient	Std. err.	t	P> t	_
-1.0723		 student_mente Yes	e		.2978704			
-1.414622		epoc_happy_pre_lo Low at baseline 769322	•	-1.095777	.1608217	-6.81	0.000	
	Ye	epoc_happy_pre_lo s#Low at baseline 082346	•	.3079845	.3905794	0.79	0.432	
5007112		ent_difficulteven 034072	t	198652	.1523553	-1.30	0.195	
2091792		student_gende	r	.0750305	.1433522	0.52	0.602	
3.724255		_con	s   	3.939617	.1086264	36.27	0.000	-

\*\*\*\*\*\*\*\*\*\*\*\*

. \*\*\* Help Seeking

\_\_\_\_\_\_

. mdesc helpseeking\_post helpseeking2\_1\_post helpseeking2\_2\_post helpseeking2\_3\_post

Variable	Missing	Total	Percent Missing
help~ng_post	311	424	73.35
helps~1_post	224	424	52.83
helps~2_post	225	424	53.07
helps~3_post	225	424	53.07

<sup>.</sup> foreach var of varlist helpseeking2\_1 helpseeking2\_2 helpseeking2\_3 {

```
2.
           tab `var' student_mentee, col
 3.
           tab `var'_post student_mentee, col
 4.
        regress `var'_post student_mentee
           regress `var'_post student_mentee student_difficultevent `var'
 5.
           regress `var'_post student_mentee##student_difficultevent `var'
 6.
 7.
.
| frequency
| column percentage |
   Do you
feel able |
 to trust
adults who
 can help | Is this student a
 when you | Raise mentee? need it? | No Yo
              No Yes | Total
-----
    Never | 19 0 | 19
| 6.88 0.00 | 4.48
   Rarely | 36 13 | 49
| 13.04 8.78 | 11.56
Sometimes | 130 71 | 201
| 47.10 47.97 | 47.41
   Always | 91 64 | 32.97 43.24 |
                                  36.56
    Total | 276 148 | 424
| 100.00 100.00 | 100.00
| frequency
| column percentage |
   Do you |
feel able |
 to trust
adults who
 can help \mid Is this student a
 when you | Raise mentee?
 need it? | No Yes | Total
```

Never | 2 1 | 3

I	2.02	0.99	1.50				
Rarely			+   21				
´		2.97					
Sometimes	39 39.39	11.88	51 25.50				
Always   	40 40.40	85 84.16	125   62.50				
Total   	99		200   100.00				
Source	l ss	C	df MS	Numb	er of obs 198)	=	200
Residual	18.61906   91.36093	61 19	98 .4614188	539 Prob 369 R-sq	> F uared	=	0.0000 0.1693
	109.				R-squared MSE		
helpsee~1_pos	st   Coeffic	ient Std.	err. t	P> t	[95%	conf.	interval]
student_mente cor	ee   .610 ns   3.181						
Source	SS	(	df MS	Numb	er of obs 196)	=	200
Residual	27.68882 82.29117	34 19	96 .4198529	387 Prob 925 R-sq	> F uared	=	0.0000 0.2518
	109.				R-squared MSE		
	ng2_1_post	Coefficie	nt Std.err	·. t	P> t	[9	5% conf.
	ent_mentee						
.7778062 student_diffi	cultevent	2539358	8 .0968518	3 -2.62	0.009	4	449413
•	seeking2_1	.199601	2 .0672301	L 2.97	0.003	.e	670139
.3321885 3.110993	_cons	2.653720	5 .2318631	l 11.45	0.000	2.	196459
Source	SS						

Model Residual	30.0820787 79.8979213	4 195	.40973293	Prob > F R-squared Adj R-squared	= = =	0.0000 0.2735 0.2586
Total	109.98	199		Root MSE	=	.6401
[95% conf.	interval]		•	Std. err.		
.1591541	.6418707	lent_mentee Yes	   .4005124	.12238	3.27	0.001
7651241	student_diff	icultevent Yes	4934494	.1377518	-3.58	0.000
student_mente	ee#student_diff .8122035	icultevent Yes#Yes	. 4472403	.1850534	2.42	0.017
.0520523	help .3153056	seeking2_1 _cons		.0667409	2.75 11.83	0.006 0.000
2.327136	3.258739			.2301033		

+-----+ | Key |------| | frequency | | column percentage |

+----+

Do you

feel able | to ask for help from | others | Is this student a when you | Raise mentee? No Yes | Total need it? Never | 20 0 | 20 7.25 0.00 | 4.72 46 16 | 62 Rarely | 16.67 10.81 14.62 138 78 | Sometimes | 50.00 52.70 50.94

Always   		72 26.09	54 36.49	 	126 29.72			
 Total   		276 100.00	148 100.00	+   	424 100.00			
+	_	+       tage						
Do you   feel able   to ask for   help from   others   when you   need it?	<u> </u>	Is this stu Raise men <sup>.</sup> No		I	Total			
+   Never 	·   	3 3.03	1 1.00	+   	4 2.01			
Rarely     Rarely		13 13.13	5 5.00	+     	18 9.05			
Sometimes		43 43.43	8 8.00	   	51 25.63			
Always   		40 40.40	86 86.00	+   	126 63.32			
Total   		99 100.00	100 100.00	+   	199 100.00			
Source	•   • -+	SS		df	MS	Number o F(1, 197	f obs = =	199 35.14
Residual	L	93.1354545	19	97	16.6132892 .472768805	Prob > F R-square	d =	0.0000 0.1514
		109.748744			.554286584			
helpsee~2_po	st	Coefficie	nt Std.	er	 r. t 		[95% conf	. interval]
student_ment	ee	.577878 3.21212	8 .093 1 .0693	7484 104		0.000 0.000		
Source	<u>.</u>	SS	(	df	MS	Number o	f obs =	199
Model Residual	-+- L   L	25.9637965 83.7849472	19	3 95	MS  8.65459884 .429666396	Prob > F R-square	) = = d =	20.14 0.0000 0.2366

Total	109.748	744 198	.554286584	Root N	-squared MSE	=	
helpseeking	g2_2_post	Coefficient	Std. err.	t	P> t		% conf.
		+					
	nt_mentee	.5451707	.0966244	5.64	0.000	.35	46076
.7357338 student_diffic 0595873	cultevent	2498812	.0964879	-2.59	0.010	44	01751
helpse	eeking2_2	.2082887	.0637128	3.27	0.001	.08	26342
.3339433 3.103556	_cons	2.691165	.2091018	12.87	0.000	2.2	78773
Source	SS .	df	MS		of obs		199 15 84
Model	27.021	535 4	6.75538375		> F		0.0000
		087 194		•	ared		0.2462
	_	744 198	.554286584	_	-squared MSE		0.2307 .65302
[95% conf. :	helps interval]	eeking2_2_post	•				
		student_mentee Yes	•	∕/ 12 <sup>1</sup>	53621	3 3/1	0 001
.1714389	.665934	163	1 .410000	<b>-</b> •12.	75021	3.34	0.001
	student_	difficultevent					
6852545	1340204	Yes	409637	4 .139	97463	-2.93	0.004
ctudont monto	o#studont	difficultovont	ļ				
student_mente	e#student_	difficultevent Yes#Yes	     .299516	1 .190	91754	1.57	0.117
student_mented	e#student_ .6745928		.299516	1 .190	ð1754	1.57	0.117
0755607	.6745928		1		91754 40953		0.117 0.003
_	.6745928	Yes#Yes helpseeking2_2	.194247	9 .064	10953	3.03	0.003
0755607	.6745928	Yes#Yes	.194247	9 .064			
0755607 .0678348	.6745928	Yes#Yes helpseeking2_2	.194247	9 .064	10953	3.03	0.003
0755607 .0678348	.6745928	Yes#Yes helpseeking2_2	.194247	9 .064	10953	3.03	0.003

   freque   column per			
Do you   feel like   you know   where to   get help   when you   need it?		student a mentee? Yes	Total
Never	13 4.71	0 0.00	13   3.07
Rarely	38	12	50
	13.77	8.11	11.79
Sometimes	112	68	180
	40.58	45.95	42.45
Always	113	68	181
	40.94	45.95	42.69
Total	276	148	424
	100.00	100.00	100.00
+			

Do you | feel like |

you know | where to | Is this student a get help | Raise mentee? when you No Yes | Total need it? | -----+----+ 2 1 | 3 2.04 0.99 | 1.51 Never | Rarely | 9 4 | 9.18 3.96 | 13 6.53 44 8 | 52 44.90 7.92 | 26.13 Sometimes | Always | 43 88 | 131 | 43.88 87.13 | 65.83

Total   	98 100.00		199 100.00				
Source	SS	df	MS		of obs		
Model Residual	12.722755   80.2420691		12.722755 .407320148	Prob :	197) > F ared	=	0.0000
Total	   92.9648241			•	-squared MSE		0.1325 .63822
helpsee~3_post	t   Coefficier	 nt Std. er	r. t	P> t	 [95%	conf.	interval]
student_mented	2   .5057587 5   3.306122						
Source	SS	df	MS	Numbe	of obs		199
	20.6767251 72.288099			R-squa	195) > F ared	=	0.2224
Total	92.9648241	198	.469519314	•	-squared MSE		0.2105 .60886
interval]	g2_3_post   Cc		Std. err.			_	
 studer	 nt_mentee				0.000		007222
.6493065 student_diffic .0281234	cultevent   -	.1487077	.0896617	-1.66	0.099	3	255388
	eeking2_3	.2333152	.0602013	3.88	0.000	.1	145859
3.037565	_cons	2.624229	.2095806	12.52	0.000	2.	210894
	SS +		MS		of obs		
Model Residual	22.0229331 70.941891	4 194	5.50573327 .365679851	Prob : R-squa	> F ared	= =	0.0000 0.2369
Total	92.9648241	198	.469519314	Adj R Root M	-squared MSE	=	0.2212 .60471
[95% conf. :	interval]	.ng2_3_post	:   Coefficie	ent Std	. err.	t	P> t
			+				

.1000115	stude	nt_mentee Yes	   .329077	4 .11614	134	2.83	0.005
5762996 ·	student_diffi 0724791	cultevent Yes	  324389	3 .12772	262	-2.54	0.012
		cultovent	ļ				
student_mente	e#student_diffi	Yes#Yes	.33440	7 .17428	389	1.92	0.056
0093374	.6781514		I				
1120242		eeking2_3	.230778	2 .05986	963	3.86	0.000
.1128243		_cons	2.69858	3 .21173	309	12.75	0.000
2.280993	3.116173						
helpseeking2_3 2. 3 4. 1	of varlist help 3_change { regress `var' s regress `var' s regress `var' s regress `var' s	tudent_men <sup>.</sup> tudent_men <sup>.</sup> tudent_men	tee tee SSE_pro tee student	ductivity _gender	_pre_lo		
	regress `var' s <sup>.</sup> cultevent stude   SS	tudent_men			/_pre_:		200
student_diffication 7. }  Source	cultevent stude	tudent_men nt_gender df	tee##SSE_pr MS	Oductivity  Number o	/_pre_: of obs 3)	= =	8.31
<pre>student_diffic 7. }</pre>	cultevent stude   SS +   5.96220072   141.992799	tudent_men nt_gender df 1 198	MS  5.96220072 .71713535	Number of F(1, 198 Prob > F	/_pre_: of obs 3) ed	= = =	8.31 0.0044 0.0403
student_diffic 7. }  Source  Model  Residual	SS +   5.96220072   141.992799	tudent_men nt_gender df  1 ! 198	MS  5.96220072 .71713535	Number of F(1, 198 Prob > FR-square Adj R-so	y_pre_: of obs 3) ed quared	= = = =	8.31 0.0044 0.0403 0.0355
student_diffic 7. }  Source  Model  Residual	cultevent stude   SS +   5.96220072   141.992799	tudent_men nt_gender df  1 ! 198	MS  5.96220072 .71713535	Number of F(1, 198 Prob > FR-square Adj R-so	y_pre_: of obs 3) ed quared	= = = =	8.31 0.0044 0.0403
student_diffication 7. }  Source  Model Residual  Total	SS +   5.96220072   141.992799 +	tudent_men nt_gender df 1 ! 198 199	MS  5.96220072 .71713535  .743492462	Number of F(1, 198 Prob > F R-square Adj R-so Root MSE	of obs 3) ed quared	= = = = = =	8.31 0.0044 0.0403 0.0355 .84684
Source  Source  Model Residual  Total  helps~1_change	SS +   5.96220072   141.992799 +   147.955	tudent_men nt_gender df 1 ! 198  199	MS  5.96220072 .71713535  .743492462	Number of F(1, 198) Prob > F R-square Adj R-so Root MSF	of obs  in obs	= = = = = = conf.	8.31 0.0044 0.0403 0.0355 .84684 interval]
Source  Source  Model Residual  Total  helps~1_change	SS +   5.96220072   141.992799 +   147.955  e   Coefficient e   .3453345 s   .0606061	tudent_mennnt_gender  df  1 198  199  Std. err  .119767	MS  5.96220072 .71713535  .743492462	Number of F(1, 198 Prob > F R-square Adj R-so Root MSF	of obs 3) ed quared [95% .1092	= = = = = conf.	8.31 0.0044 0.0403 0.0355 .84684 interval] .5815172 .2284454
Source  Source  Model Residual  Total  helps~1_change student_menteecons	SS +	tudent_men nt_gender  df  1 98 199 Std. err119767 .0851105	MS 5.96220072 .71713535743492462 t 2.88 0.71	Number of F(1, 198 Prob > F R-square Adj R-so Root MSF P> t	/_pre_: of obs 3) ed quared [ 95% .1093	= = = = = conf. 	8.31 0.0044 0.0403 0.0355 .84684 
Source  Source  Model Residual  Total  helps~1_change student_menteecons	SS +	tudent_men nt_gender  df  1 98 199 Std. err119767 .0851105	MS 5.96220072 .71713535743492462 t 2.88 0.71	Number of F(1, 198 Prob > F R-square Adj R-so Root MSF P> t	/_pre_: of obs 3) ed quared [ 95% .1093	= = = = = conf. 	8.31 0.0044 0.0403 0.0355 .84684 
Source  Source  Model Residual  Total  helps~1_change student_menteecons	SS +	tudent_men nt_gender  df  1 98 199 Std. err119767 .0851105	MS 5.96220072 .71713535743492462 t 2.88 0.71	Number of F(1, 198 Prob > F R-square Adj R-so Root MSF P> t	/_pre_: of obs 3) ed quared [ 95% .1093	= = = = = conf. 	8.31 0.0044 0.0403 0.0355 .84684 
Source  Model Residual Total  helps~1_change student_menteecons  Source  Model Residual	SS +	tudent_mennnt_gender  df  1 198  199  Std. err  .119767 .0851105	MS 5.96220072 .71713535743492462 t 2.88 0.71 MS125893359 .804508638	Number of F(1, 198 Prob > F R-square Adj R-so Number of F(2, 123 Prob > F R-square Adj R-so Adj R-so	of obs  and and and and and and and and and an	= = = = = conf.  1519 2333 = = = = =	8.31 0.0044 0.0403 0.0355 .84684 

interval]	2_1_change						_
.4706764	ent_mentee	.0848791	.1949023	0.44	0.664		3009181
.4700704 SSE_productivi	tv pre low	.0564924	.1696494	0.33	0.746	) -	2793181
.3923029	c)_p: c_=o		V=000.0.				V=/
	_cons	.0423459	.1055147	0.40	0.689	)	166514
.2512058							
Source 1	SS	df	MS	Number o	f ohs	=	200
•							
	5.96225338			Prob > F	,	=	4.14 0.0174
	141.992747			R-square	d	=	0.0403
							0.0306
	147.955						.84898
·							
1				n. I. I			1 33
nelps~1_change	Coetticier	it Std. err	'. t	P> t	[95%	cont.	. interval]
 student_mentee	- <del>1</del>	 7 12/10621	7 7Ω	a aak	100/	 L074	 52072Ω
student_mentee student_gender							
	.060251		0.64				
- I		1.0			<b>.</b> .		
	SS				t obs	=	200
	6 1540054			\ \	)	=	4.28
:	6.1549054						0.0152
	141.800095			•			0.0416
-				AU I N-SU	uareu	=	0.0319
TOTAL		านน	7/13/102/162	Poot MSE		_	Q/Q/1
	147.955	199	.743492462			=	.84841
	14/.955	199	.743492462	Root MSE		=	.84841
				Root MSE			
helpseeking2_				Root MSE			
interval]	 1_change   Co	oefficient	Std. err.	Root MSE	 > t	· <u>·</u>	95% conf.
interval]		oefficient	Std. err.	Root MSE	 > t	· <u>·</u>	95% conf.
interval] 		oefficient	Std. err.	Root MSE	 > t  	2]	95% conf.
interval]   studen	 1_change   Co	oefficient	Std. err.	Root MSE	 > t  	2]	95% conf.
interval]   studen .5757788		oefficient 	Std. err.	t P	 > t  	?] 	95% conf. 
interval]   studen		oefficient 	Std. err.	t P	 > t  	?] 	95% conf. 
interval]   studen .5757788 student_diffic		oefficient 	Std. err121051 .121926	t P 2.78 0 0.52 0	 > t   .006 .605	2] 	95% conf.  9983347 .177361
interval]   studen .5757788 student_diffic		oefficient  .3370567 .0630866	Std. err121051 .121926	t P 2.78 0 0.52 0	 > t   .006 .605	2] 	95% conf.  9983347 .177361
interval] 		oefficient  .3370567 .0630866	Std. err121051 .121926 .0966807	t P 2.78 0 0.52 0 0.38 0	 > t   .006 .605 .702		95% conf. 9983347 .177361
interval] 		.3370567 .0630866 .0370282	Std. err121051 .121926 .0966807	t P 2.78 0 0.52 0 0.38 0	 > t   .006 .605 .702		95% conf. 9983347 .177361
interval]	1_change   Cot t_mentee   ultevent   _cons	oefficient .3370567 .0630866 .0370282	Std. err121051 .121926 .0966807	t P 2.78 0 0.52 0 0.38 0	 > t   .006 .605 .702		95% conf. 9983347 .177361 1536337
interval] studen .5757788 student_diffic .3035342 .2276901 Source	1_change   Cot t_mentee   ultevent   _cons	.3370567 .0630866 .0370282	Std. err121051 .121926 .0966807	t P 2.78 0 0.52 0 0.38 0	 > t   .006 .605 .702 		95% conf. 9983347 .177361 1536337
interval] studen .5757788 student_diffic .3035342 .2276901 Source	1_change   Cot t_mentee   ultevent   _cons	oefficient .3370567 .0630866 .0370282	Std. err121051 .121926 .0966807	t P 2.78 0 0.52 0 0.38 0 Number of F(5, 120)	 > t   .006 .605 .702 		95% conf. 9983347 .177361 1536337
interval] studen .5757788 student_diffic .3035342 .2276901 Source   Model	1_change   Cot t_mentee   ultevent   _cons	oefficient .3370567 .0630866 .0370282df	Std. err121051 .121926 .0966807	Root MSE  t P  2.78 0  0.52 0  0.38 0  Number of F(5, 120 Prob > F	 > t   .006 .605 .702  f obs		95% conf. 9983347 .177361 1536337

Total	99.2063492			Adj R- Root M	squared SE		-0.0338 .90578
	helpseeking interval]						
5318914	.4946591	ent_mentee Yes		161 .	259239	-0.	07 0.943
3858353	SSE_productivi Low at .4168684	ty_pre_low baseline	:	166 .2	027101	0.	08 0.939
student_mentee	#SSE_productivi Yes#Low at 1.046273			647 .4	117865	0.	56 0.576
3891345	student_diff .3029902				747851		
2593198		ent_gender	·   .0834		731204	0.	48 0.631
2156012	.3033467	_cons	s   .0438	728 .	131052	0.	33 0.738
Source		df	MS		of obs		199
+   Model		1 5	5.06536775	F(1, 1 Prob >	97) F	=	7.32 0.0074
Residual	136.371818		.692242732		red squared		0.0358 0.0309
Total	141.437186	198	.714329222	Root M	SE	=	.83201
helps~2_change	Coefficient	Std. err	. t	P> t	[95%	conf.	interval]
	.3190909 .0909091						
_cons	6   .0909091 	.0836203	1.09 	0.278 	0739 	967 	.2558149
Source	SS	df	MS	Number F(2, 1	of obs 23)	=	126 0.04
Model	.071227817 99.9684547	2 .	.035613908	Prob >	F <sup>´</sup> .	=	0.9571
Residual	99.9684547	123 .	.812751664	R-squa	red sauared	=	0.0007 -0.0155
Total	100.039683	125	.80031746	Root M	SE	=	.90153
	2_2_change   Co						

interval]							
stud .3686568	lent_mentee	0191119	.1958983	-0.10	0.922	2 -	4068805
SSE_productivi	ty_pre_low	.0483016	.1705163	0.28	0.77	7 -	.2892249
.2852236	_cons	.0752964	.1060539	0.71	0.479	-	.1346307
Source	SS	df	MS	Number F(2, 1	of obs		199 4.49
Model	6.18979398	2	3.09489699	•	. F		0.0125
Residual	135.247392		.690037714	R-squa		=	
+				Adj R-	squared	=	0.0340
Total	141.437186	198	.714329222	Root M	ISE	=	.83069
helps~2_change	c   Coefficient	t Std. er	r. t	P> t	[95%	conf.	interval]
student_mentee	.2806756	.1215569	9 2.31	0.022	.0409	9482	.520403
student_gender	.1557377	.1220014	1.28	0.203	0848	3662	.3963416
_cons	.0389965	.0928649	9 0.42	0.675	1442	1462	.2221393
Source	SS	df	MS		of obs	=	199
+				F(2, 1	•		3.86
Model					F		0.0228
Residual	136.083502	196	.694303583	•	red squared		0.0379 0.0280
Total	141.437186	198	714329222	Root M	•	=	.83325
TOTAL	141.43/100	150	.714323222	NOOC 1	.JL	_	.03323
	_2_change   Coe	efficient	Std. err.	t	P> t	[9	95% conf.
interval]							
ctudon	nt_mentee	200540	1102620	2 50	0 010	c	7777170
.5437532	it_meritee	.300340	.1192039	2.59	0.010		0/33420
student_diffic	ultevent	0773719	120067	0 64	0 520	_ 1	1594172
.314161	arcevene	.0773713	.120007	0.04	0.520	• -	1334172
	cons   .	.0619923	.0950095	0.65	0.515	1	L253798
.2493644							
Source		df					126
•	2.4648986				.20)		0.61 0.6952
	97.5747839			R-cons	red red	=	0.0352 0.0316
+ vestanat	JI.JI4/033	120	7671610	n-3qua Δdi R-	sauared	=	-0.0160
				יאט ווי	Jquui Cu	-	0.0100

			. 60031746				.90173
[95% conf.		g2_2_change		ient Std.			
7731826		dent_mentee Yes		028 .258	0797	-1.6	0.31
4548302		ity_pre_low t baseline		733 .201	8035	-0.2	27 0.78
tudent_mentee		ity_pre_low baseline	.4121	724 .409	9449	1.6	0.31
3148304	student_diff .374199	ficultevent	.0296	843 .174	0034	0.1	17 0.86
104387		dent_gender			3462		
2395812	.2770458	_cons	.0187	323 .13	0466	0.1	14 0.88
Source	SS	df	MS	Number o F(1, 197			199 3.82
Model   Residual		197 .7	700555309	Prob > F R-square	d	= =	0.0522 0.0190
Total				Adj R-sq Root MSE			
elps~3_change	Coefficient	Std. err.	t	P> t	[95%	conf.	interval
tudent_mentee	.231865 .1938776	.118679	1.95 2.29	0.052 0.023	0021 .0271	794 404	.465909 .360614
Source	SS  .236622767	df	MS	Number o	f obs	=	125
Residual	100.531377	122 .8	324027682	R-square	d	=	0.0023
+	100.768			Adj R-sq	uared	=	-0.0140
	2_3_change   Co						

	+				. – – – – – –		
	dent_mentee	049373	.1974458	-0.25	0.803	-	.4402368
.3414908	•• •	0024447	4720204	0.40	0.630		2574402
SE_productiv .4236427	ity_pre_low	.0831117	.1720201	0.48	0.630	) –	.2574193
.4230427	_cons	.166739	.1075337	1.55	0.124	_	.0461346
.3796127	_cons	.100/39	.10/333/	1.55	0.124	_	.0401340
<b>C</b>	l 66	10	MC	No contra a co	- C - I		100
Source	SS .	df	MS	Number F(2, 19	of obs	=	199 1.93
Model	   2.72141147	2	1.36070573	Prob >	•	=	0.1474
Residual	137.962006		.703887784	R-squar		=	
	137.302000 +	150	.703007704	•	squared		
Total	140.683417	198	.710522309	Root MS	•	=	.83898
. 0 00.2	1 - 101000 1-1				-		10000
eтbs~з_cnang	e   Coefficient +	. sta. err	τ 	Y> T	[95% 	cont.	interval]
tudent_mente	e   .2239597	.1228004	1.82	0.070	0182	202	.4661395
tudent_gende	•	.1231368					.2747939
_con:	i	.0943495	1.94	0.054	0029	519	.3691892
Source	l ss	df	MS	Number	of ohs	=	199
	33 <del> </del>			F(2, 19		=	2.90
Model	4.03760882	2	2.01880441	Prob >	•	=	0.0576
Residual	•		.697172491	R-squar		=	
	+			•	quared		0.0188
Total	140.683417	198	.710522309	Root MS	•	=	.83497
helnseeking2	_3_change   Coe	fficient	Std. err.	t	P> +	[9	5% conf.
nterval]	_5_61141186   606	TTELETIC	Jea. C		. ,   c	L	370 - COTT :
. د ام ـ ام	nt montos I	2104542	1102770	1 70	0 070	^	· ) /[ () / L ()
	nt_mentee   .	2104542	.1193779	1.76	0.079	0	249759
4458843							
4458843 tudent_diffi	nt_mentee   . cultevent   .				0.079 0.164		689308
4458843	cultevent   .	1680607	.1201697	1.40	0.164	0	
4458843 tudent_diffi 4050523		1680607	.1201697	1.40	0.164	0	689308
4458843 tudent_diffi 4050523	cultevent   .	1680607	.1201697	1.40	0.164	0	689308
4458843 tudent_diffi	cultevent   .	1680607	.1201697	1.40	0.164	0	689308
4458843 tudent_diffi 4050523 3193038 	cultevent  cons	1680607 .130426	.1201697 .095773	1.40 1.36	0.164 0.175	0 0	689308 584517 
4458843 tudent_diffi 4050523 3193038  Source	cultevent   .	1680607 .130426 	.1201697 .095773 	1.40 1.36 	0.164 0.175 of obs	0 0	689308 584517 
4458843 tudent_diffi 4050523 3193038 Source	cultevent  cons	1680607 .130426 df	.1201697 .095773 	1.40 1.36 Number F(5, 11	0.164 0.175 of obs	0 0	689308 584517  125 0.28
4458843 tudent_diffi 4050523 3193038 Source Model	cultevent  cons     SS +	1680607 .130426 df 5	.1201697 .095773  MS 	1.40 1.36 Number F(5, 11 Prob >	0.164 0.175 of obs	0 0	689308 584517  125 0.28 0.9239
4458843 tudent_diffi 4050523 3193038 Source Model Residual	cultevent  cons     SS +   1.1672737   99.6007263	1680607 .130426 df 5 119	.1201697 .095773  MS  .23345474 .836980893	1.40 1.36 Number F(5, 11 Prob > R-squar	0.164 0.175 of obs	0 0	689308 584517  125 0.28 0.9239 0.0116
4458843 tudent_diffi 4050523 3193038 Source Model Residual	cultevent  cons    SS +	1680607 .130426 df 5 119	.1201697 .095773  MS  .23345474 .836980893	1.40 1.36  Number F(5, 11 Prob > R-squar Adj R-s	0.164 0.175 of obs 19) F red squared	0 0	125 0.28 0.9239 0.0116 -0.0299

	  helpseeking2_3_change	Coefficient	Std. err.	t	P> t
[95% conf.		•			
		•			
	student_mentee				
	Yes	1350635	.2619409	-0.52	0.607
6537326	.3836056				
	CCC mandustivity man lay	 			
	SSE_productivity_pre_low Low at baseline	l   .042139	.2050369	0.21	0.838
3638546	.4481326	.042133	.2030303	0.21	0.050
		1			
student_mentee	#SSE_productivity_pre_low	j			
	Yes#Low at baseline	.0270677	.4160839	0.07	0.948
7968199	.8509554	ı			
	student difficultovent	   .1242421	.1767273	0.70	0.483
2256957	<pre>student_difficultevent .4741799</pre>	1 .1242421	.1/0/2/3	0.70	0.465
.2230337	student_gender	.1257406	.1751547	0.72	0.474
2210832	.4725643				-
	_cons	.0908689	.1337737	0.68	0.498
1740163	.3557542				

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\*\*\*\*\*\*\*\*\*\*\*\*

. \*\*\*\* SSE

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. mdesc SSE\_productivity\_post SSE\_belong\_post SSE\_aspiration\_post SSE\_total\_post

Variable	Missing	Total	Percent Missing
SSE_produc~t SSE_belong~t SSE_aspira~t SSE_total_~t	296 296	424 424 424 424	69.81 69.81 69.81 69.81

<sup>. \*\*</sup> Productivity

. mean SSE\_productivity\_pre, over(student\_mentee)

Mean estimation	Number of
obs = 313	

------------

```
interval]
c.SSE_productivity_pre@student_mentee |
                    No | 6.53688 .1120223
6.757294
                   Yes | 6.055985 .2706841 5.523387
6.588582
. mean SSE_productivity_post, over(student_mentee)
Mean estimation
                                         Number of
obs = 128
_____
                       Mean Std. err. [95% conf.
interval]
-----+
c.SSE_productivity_post@student_mentee |
                          6.691757 .1783962
7.044771
                    Yes | 5.796857 .4193625 4.967015
6.6267
. foreach var of varlist SSE_productivity_post {
 2. regress `var' student_mentee
 3. regress `var' student_mentee SSE_productivity_pre_low
 4. regress `var' student_mentee student_gender5. regress `var' student_mentee student_difficultevent
 6. regress `var' student_mentee##SSE_productivity_pre_low
student difficultevent student gender
 7. }
   Source | SS df
                          MS
                              Number of obs =
                                            128
                              F(1, 126) =
                                            4.83
  0.0298
                                           0.0369
                              Adj R-squared = 0.0293
   Total | 461.953581 127 3.63742977
                              Root MSE =
                                           1.8791
______
SSE_producti~t | Coefficient Std. err. t P>|t| [95% conf. interval]
```

Source		df	MS		of obs		
Model   Residual	170.630426 243.537823	2 115	85.315213 2.1177202	R-squared		=	0.4120
	414.168249			Root M	Squared SE	=	0.4018 1.4552
interval]	:ivity_post   C						_
stud 8551608	dent_mentee	-1.610943	.3815525	-4.22	0.000		-2.366725
	ity_pre_low	-2.308287	.2887993	-7.99	0.000		-2.880343
7.762386	_cons	7.423095	.171289	43.34	0.000		7.083805
		1.6					
•	SS		MS 				128 3.10
	21.8318432			Prob >	F	=	0.0485
Residual	440.121738	125	3.5209739	R-squa Δdi R-	red squared	=	0.0473 0.0320
Total					SE		1.8764
	C   Coefficient						
student_mentee	+ e   -1.08489	.438077	2 2 -2 <b>.</b> 48	0.015	-1.9518	 899	2178809
student_gender	.4212238	.361893	3 1.16	0.247	29500	<b>0</b> 79	1.137455
_cons	5   6.554129 	.221003	1 29.66 	0.000	6.116	736 	6.991522
	SS						128
+	38.3081842		10 1540021	F(2, 1	25)	=	5.65
Residual	423.645397	125	3.38916318	R-squa	г red	=	0.0829
	423.645397			Adj R-	squared	=	0.0683
	461.953581						
interval]	vity_post   Coe						
	nt_mentee	6723641	.4086321	-1.65	0.102	-1	.481098
.1363696 student_diffic 1771088	cultevent	8451994	.3375688	-2.50	0.014	-	1.51329

7.438806	_cons   7.	001384	.2210177	31.68 0.000	0 6.563	963
Source			MS			
Model		5	35.4310249	F(5, 112) Prob > F	= 0	0.0000
-	237.013124			R-squared Adj R-square		
Total	414.168249	117	3.53989956	Root MSE	= 1	4547
[95% conf.	SSE_product interval]	ivity_pos	-	ient Std. er		
	 stud	ent_mente	ا م			
-2.827408	9351584	Yes		.4775099	-3.94	0.000
-2,027400		<b>4</b>				
		baseline		.388 .3261209	9 -7.09	0.000
-2.957555			I			
student_mentee	e#SSE_productivi Yes#Low at			256 .8364192	2 0.70	0.487
-1.073832	2.240683		1			
8343357	student_diff .3401444	iculteven	nt  2470	956 .2963805	-0.83	0.406
1334785		ent_gende	er   .4431	.999 .2910503	1.52	0.131
	7.780346	_con	ns   7.36	979 .2072079	35.57	0.000
	7.760540					
•						
<ul><li>. ** Belong</li><li>. mean SSE_bel</li></ul>	.ong_pre, over(s	tudent_me	entee)			
Mean estimation	on			1	Number of o	obs =
-			Mean St	d. err. [9	95% conf.	
interval]		+				
- c.SSE_belong_p	ore@student_ment	ee				

```
No | 5.936775 .1258086
                                    5.689235
6.184316
                Yes | 5.445946 .3350216
                                    4.786759
6.105133
. mean SSE_belong_post, over(student_mentee)
Mean estimation
                                      Number of obs =
128
                    Mean Std. err. [95% conf.
interval]
-----
c.SSE_belong_post@student_mentee |
                 No | 6.305281 .2020667 5.905427
6.705134
                 Yes | 5.322222 .4678286
                                     4.396474
6.247971
-----
. foreach var of varlist SSE_belong_post {
 2. regress `var' student_mentee
 3. regress `var' student_mentee SSE_belong_pre_low
 4. regress `var' student_mentee student_gender
 5. regress `var' student_mentee student_difficultevent
 6. regress `var' student_mentee##SSE_belong_pre_low student_difficultevent
student_gender
 7. }
                               Number of obs = 128
4.58
  Source SS df MS
------ F(1, 126) =
                                       = 0.0342
                   1 20.588925 Prob > F
   Model | 20.588925
  .
------ Adj R-squared = 0.0274
    SSE_belong_p~t | Coefficient Std. err. t P>|t| [95% conf. interval]
student_mentee | -.9830582 .4591967 -2.14 0.034 -1.891795 -.0743215
_cons | 6.305281 .2108996 29.90 0.000 5.887916 6.722645
  Source | SS df
MS Number of obs =
                                             118
```

Total	533.784857				-	•	= =	0.4422 1.5952
SSE_belong_p	ost   Coefficie		err.		-		[95% c	onf.
student_men	tee   -1.91091	6 .418	37757	-4.56	5 0.0	00	-2.740	43
SSE_belong_pre_ -2.103256	low   -2.7116	4 .307	71391	-8.83	0.0	90 -	3.3200	23
_c 7.684045	ons   7.29865	3 .194	15627	37.5	1 0.0	00	6.9132	62
Source	SS	df	MS			of obs		128 2.32
-+   [ahoM	20.9680088	2	10 18100	 11	Proh >	25) F	=	0.1028
•	565.655876							0.0357
+-					Adj R-			0.0203
Total	586.623885	127	4.619085		Root M	•		
 SSE_belong_p~t		 Std. err	·. t	 I	 P> t	 [95%	 conf.	interval]
student_mentee	-1.036618							0537084
student_gender _cons	i						7062	
Source	SS	df	MS			of obs		128
	50.9297192 535.694166							
	586.623885				-	-		
	ng_post   Coeff	icient	Std. err	•	t	P> t	[9	5% conf.
 student	_mentee  71							
.192289 student_difficu	ltevent   -1.0	10019	.3795942		-2.66	0.009	-1.	761283
258755 7.167165	_cons   6.6	75287	.2485331	2	26.86	0.000	6	.18341

Sourc			df		Number of		= 118		
Mode Residua	el   al	248.801516 284.98334	5 112	49.7603033 2.54449411	R-squared	=	= 19.56 = 0.0000 = 0.4661 = 0.4423		
		533.784857							
[95% conf.		SSE_belong_po cerval]	•						
-3.065582	 - 1			-2.039167	.5180323	-3.94	0.000		
-3.351021	-1	SSE_belong_pre_l Low at baselin 1.984706		-2.667863	.3447902	-7.74	0.000		
student_mer -1.571756	١	e#SSE_belong_pre_l /es#Low at baselin 2.074028		.2511359	.920015	0.27	0.785		
-1.017886		udent_difficulteve .2644531	•			-1.16	0.247		
2066037	1		ler	.4307053	.3216504	1.34	0.183		
6.819044	7.	<del>_</del>	ns	7.279896	.2325926	31.30	0.000		
<pre> ** Aspiration . mean SSE_aspiration_pre, over(student_mentee)  Mean estimation = 313</pre> Number of obs									
interval]					Std. err.	_			
		ion_pre@student_me		I	.1155984	7.9	976353		
8.05163									

```
. mean SSE aspiration post, over(student mentee)
Mean estimation
                                             Number of obs
= 128
                         Mean Std. err. [95% conf.
interval]
 c.SSE aspiration_post@student_mentee |
                     No | 7.079208 .2184535 6.646928
7.511488
                     Yes | 7.194444 .4462437 6.311409
8.07748
  . foreach var of varlist SSE_aspiration_post {
 2. regress `var' student_mentee3. regress `var' student_mentee SSE_aspiration_pre_low
 4. regress `var' student mentee student gender
 5. regress `var' student_mentee student_difficultevent
 6. regress `var' student_mentee##SSE_aspiration_pre_low student_difficultevent
student_gender
 7. }
    Source | SS df MS
                                  Number of obs = 128
F(1, 126) = 0.06
                                  F(1, 126) =
-----
    Model | .282914668 1 .282914668
                                  Prob > F = 0.8112
R-squared = 0.0005
  Residual | 621.783003 126 4.93478574
-----
                                  Adj R-squared = -0.0075
   Total | 622.065918 127 4.89815683
                                  Root MSE
SSE_aspirati~t | Coefficient Std. err. t P>|t| [95% conf. interval]
mentee | .1152365 .4812786 0.24 0.811 -.8371997 1.067673
_cons | 7.079208 .2210413 32.03 0.000 6.641774 7.516642
student mentee |
   Source | SS df MS Number of obs = 118
----- F(2, 115) =
    Model | 154.600152 2 77.3000762 Prob > F
                                           = 0.0000
  0.2722
-----+----+
                                  Adj R-squared = 0.2595
    Total | 568 117 4.85470085
                                  Root MSE =
```

interval]	ion_post   Coef					_	
studen	t_mentee  1	294387	501277	-0.26	0.797	-1.	122372
.8634945	c_merreee   • • •	224307	. 301277	0.20	0.757	-•	122372
SSE_aspiration	pre low   -2.	559411	.3959485	-6.46	0.000	-3.	343708
-1.775113							
	_cons   7.	712725	.2125965	36.28	0.000	7.	291613
8.133838							
				_			
	SS	df	MS		r of obs		
					125)		
	34.4386699				> F		
Residual	587.627248	125	4.70101798	R-squ	ared	=	0.0554
				Adj R	-squared	=	0.0402
Total	622.065918	127	4.89815683	Root	MSE	=	2.1682
	Coefficient			P> t	[95%	cont.	interval
	-+			0 420	1 20/	1075	6006505
	3931582						
student_gender	1.127148	.418162	5 2.70	0.008	. 2995	5528	1.954744
_cons	6.710932	. 255365	9 26.28	0.000	6.205	531	7.216333
Source	SS	٩ŧ	MC	Numbo	r of obs	_	120
		uı	CIT				
•	8.87552823	ີ	4 42776412	Γ(∠) Doob	125) > F	=	0.90
	613.19039			K-Squ	ared	=	0.0143
	 622.065918			Auj K	-squared MSE		
Total	622.065918	12/	4.89815683	KOOT	MSE	=	2.2148
SSE asnirat	ion_post   Coef	ficient	Std err	+	P> +	Г9	5% conf
interval]	1011_post   coc1	TETETIC	Sca. cii.	·	17   6	L	570 COIII :
_							
	·						
studen	t mentee   .2	2567566	.4916193	0.52	0.602		716219
1.229732			<del></del>		<b></b>	•	
student_diffic	ultevent   - 5	375003	.4061241	-1.32	0.188	- 1	.34127
.2662697				52	3.200	_	
,	_cons   7.	276114	.2659031	27.36	0.000	6.	749859
7.802369			0 0 0 0 0 2	_, , , , ,	2.000	٠.	
Source	SS	df	MS	Numbe	r of obs	=	118
+					112)		
Model	181.049048	5	36.2098096				
		-					

	386.950952			R-squared	=	0.3187
	-+   568			Adj R-squared Root MSE		0.2883 1.8587
10041	1 300		1.05170005	NOOC TISE		1.0307
	SSE_aspiration	noct	Coofficient	- C+d onn	+	D
[95% conf.	<del>_</del> -	_post	Coefficient	. Stu. em.	L	P> נ
=			+			
	student_m		•	.6373407	-0 92	0.357
-1.852288	.6733296	163	3894792	.03/340/	-0.92	0.557
			1			
	SSE_aspiration_pr	_	!			
-3.35448	Low at base	line	-2.4/9344	.4416816	-5.61	0.000
-3:33440	-1.004203		I			
student_mente	ee#SSE_aspiration_pr		į			
4 642404	Yes#Low at base	line	.389738	1.025513	0.38	0.705
-1.642184	2.42166		T.			
	student difficult	event	3144646	.3672627	-0.86	0.394
-1.042149	.4132194					
2202005		ender	.9882305	.3784903	2.61	0.010
.2383005		cons	7.48522	.2692129	27 80	0.000
6.951809		_00113	1 / 1-10322	.2002120	27.00	0.000

<sup>.</sup> mean SSE\_total\_pre, over(student\_mentee)

Mean	PSTIM	nation

Number of obs = 313

	Mean	Std. err.	[95% conf.	interval]
c.SSE_total_pre@student_mentee   No   Yes	6.892486 6.313707	.1053875 .2616026	6.685126 5.798978	7.099846 6.828435

<sup>.</sup> mean SSE\_total\_post, over(student\_mentee)

Mean estimation

Number of obs =

128

| Mean Std. err. [95% conf.

<sup>. \*\*</sup> Total

```
interval]
-----
c.SSE_total_post@student_mentee |
                  6.692082 .1809636 6.333987
7.050176
             Yes 6.104508 .4119097 5.289413
6.919603
. foreach var of varlist SSE_total_post {
 2. regress `var' student_mentee
 3. regress `var' student_mentee SSE_total_pre_low
 4. regress `var' student_mentee student_gender
 5. regress `var' student_mentee student_difficultevent
 6. regress `var' student_mentee##SSE_total_pre_low student_difficultevent
student_gender
7. }
SSE\_total\_post \mid Coefficient Std. err. t P>|t| [95% conf. interval]
______
Source | SS
   SSE total post | Coefficient Std. err. t P>|t| [95% conf.
interval]
------
 student_mentee | -1.070831 .3916133 -2.73 0.007 -1.846542
-.2951207
SSE_total_pre_low | -2.37849 .3063911 -7.76 0.000 -2.985392
-1.771588
      _cons | 7.327916 .1694223 43.25 0.000 6.992323
7.663509
```

Source | SS df MS Number of obs = 2.22 -----F(2, 125) = 2.22Prob > F = 0.1133 Model | 15.6574627 2 7.82873134 Prob > F = 0.1133 Residual | 441.559064 125 3.53247251 R-squared = 0.0342 0.0188 1.8795 SSE\_total\_post | Coefficient Std. err. t P>|t| [95% conf. interval] student\_mentee | -.8382221 .438792 -1.91 0.058 -1.706646 .0302017 student\_gender | .5557059 .3624837 1.53 0.128 -.1616944 1.273106 \_cons | 6.510514 .2213637 29.41 0.000 6.072408 6.948621 3.81 Model | 26.2747459 2 13.1373729 Prob > F = 0.0247 SSE\_total\_post | Coefficient Std. err. t P>|t| [95% conf. ------student\_mentee | -.3775783 .412136 -0.92 0.361 -1.193247 student\_difficultevent | -.7975729 .3404634 -2.34 0.021 -1.471392 -.1237537 \_cons | 6.984262 .2229128 31.33 0.000 6.54309 7.425434 \_\_\_\_\_\_ Source | SS df MS Number of obs = 118 ------ F(5, 112) = 15.61 Adj R-squared = 0.3844Root MSE = 1.4741SSE\_total\_post | Coefficient Std. err. t P>|t| [95% conf. interval] -----

student\_mentee |

-2.518575806846	Yes   -1	.549627	.4890262	-3.17	0.002				
SSE_total_ Low at ba -3.14914 -1.763549		.456344	.3496544	-7.03	0.000				
student_mentee#SSE_total_ Yes#Low at ba 6848674 2.572062		9435974	.8218875	1.15	0.253				
student_difficu 8167116 .3818149	ltevent	2174484	.3024486	-0.72	0.474				
student 0230738 1.149119	_gender   .		.2958033		0.060				
6.832257 7.656603	_cons	7.24443	.2080241	34.82	0.000				
<pre> *********** . ** Resilience . *********** mdesc brs_post  Variable   Missing Total Percent Missing</pre>									
	246	424	 58.0						
mean brs_pre, over(student_mentee)  Mean estimation  Number of obs = 420									
	Mean		r. [95%						
<pre>c.brs_pre@student_mentee     No     Yes</pre>		.0363403	1 3.05 9 2.79	0729 3 9637 3	3.193593 3.002177				
. mean brs_post, over(student_mentee)									
Mean estimation			Nu	mber of o	obs = 178				

c.brs\_post@student\_mentee | 3.079518 .0579237 2.965208 No 3.193828 3.001754 .0712713 2.861104 3.142405 Yes . foreach var of varlist brs\_post { regress `var' student\_mentee
 regress `var' student\_mentee brs\_pre\_low 4. regress `var' student\_mentee student\_gender 5. regress `var' student\_mentee student\_difficultevent6. regress `var' student\_mentee##brs\_pre\_low student\_difficultevent student\_gender 7. } Source | SS df MS Number of obs = 178 0.69 178 ------ F(1, 176) = = 0.4068 Prob > F = R-squared = 0.0039 ------ Adj R-squared = -0.0017 Total | 68.463874 177 .386801548 Root MSE = .62248 brs\_post | Coefficient Std. err. t > |t| [95% conf. interval] 2.944675 \_cons | 3.079518 .0683257 45.07 0.000 3.214361 Source | SS df MS Number of obs = 177 . ------ Adj R-squared = 0.0190 brs\_post | Coefficient Std. err. t P>|t| [95% conf. interval] ------\_cons | 3.13151 .0719093 43.55 0.000 2.989583 3.273437 Source | SS df MS Number of obs = 178 -----Adj R-squared = -0.0036Total | 68.463874 177 .386801548 Root MSE = .62305 brs\_post | Coefficient Std. err. t P>|t| [95% conf. interval]

student_mente student_genden _cons	;	.0962992	-0.82	0.538 0.411 0.000	26	94 .1107147
Residual	1.92421678 66.5396572	175 	.962108388 .380226613	F(2, 17 Prob > R-squai	F red squared	= 2.53 = 0.0825 = 0.0281 = 0.0170
interval]	brs_post   Co					-
 studer .1222981	nt_mentee   - cultevent   -	.0612189	.0929853		0.511 0.038	
3.312507	_cons	3.15916	.0776987	40.66	0.000	3.005813
Source	SS	df 	MS		of obs 71)	
Model Residual	62.9851502		.368334212	Prob > R-squai	f <sup>°</sup> red squared	= 0.0134 = 0.0800
Total	68.4624207	176	.388991027	Root MS	SE 	= .60691
conf. interval	<del></del> -	•	nt Std. er		·	[95%
.0829031	tudent_mentee Yes					53828204
Low 1645811	<pre>brs_pre_low at baseline</pre>	    476642 	5 .158091	1 -3.6	0.00	37887039
student_mented Yes#Low .8228369	e#brs_pre_low at baseline	.427043	5 .200509	9 2.1	13 0.03	5 .0312502
student_d: 0049898	ifficultevent	  187627	8 .092524	9 -2.6	0.04	43702658

student\_gender | -.0085724 .0965123 -0.09 0.929 -.1990812 \_cons | 3.274123 .0876786 37.34 0.000 3.101051 3.447195 \*\*\*\*\*\*\*\*\*\* \*\*\* Mental Health \*\*\*\*\*\*\*\*\*\* . mdesc wemwbs Variable | Missing Total Percent Missing wemwbs 127 424 . \*\* Raw . mean wemwbs, over(student\_mentee) Mean estimation Number of obs = 297\_\_\_\_\_\_ Mean Std. err. [95% conf. interval] -----c.wemwbs@student\_mentee No | 21.76145 .3251802 21.12149 22.40141 Yes | 20.55629 .9677113 18.65182 22.46075 . mean wemwbs\_post, over(student\_mentee) Mean estimation Number of obs = 119Mean Std. err. [95% conf. interval] c.wemwbs\_post@student\_mentee | 

 No
 22.27333
 .462928
 21.35661
 23.19006

 Yes
 20.05962
 .8252993
 18.4253
 21.69393

 . foreach var of varlist wemwbs\_post { 2. regress `var' student\_mentee

- 3. regress `var' student\_mentee wemwbs\_low
- 4. regress `var' student\_mentee student\_gender
- 5. regress `var' student\_mentee student\_difficultevent
- 6. regress `var' student\_mentee##wemwbs\_low student\_difficultevent
  student\_gender

Source	SS	df	MS		of obs	=	119
+	00 5750272	1	00 5750272	F(1, 1:	•	=	5.12
Model	99.5758272	1		Prob >		=	0.0255
Residual	2276.29805	117	19.4555389	R-squar		=	0.0419
+ 	2275 07207	440	20 4245244	•	squared	=	0.0337
Total	2375.87387	118	20.1345244	Root M	SE	=	4.4108
wemwbs_post	Coefficient	Std. er	r. t	P> t	[95%	conf.	interval]
student mentee	-2.213718	.978513	8 -2.26	0.026	-4.151	613	2758226
cons		.457383		0.000	21.36		23.17916
Source	SS	df	MS	Number	of obs	=	103
·+				F(2, 10	<b>00</b> )	=	17.39
Model	518.872919	2	259.43646	Prob >	F	=	0.0000
Residual	1492.08849	100	14.9208849	R-squai	red	=	0.2580
·+				•	squared	=	0.2432
Total	2010.96141	102	19.7153079	Root M	SÉ	=	3.8628
·							
wemwbs_post	Coefficient	Std. er	r. t	P> t	[95%	conf.	interval]
	-+						
student_mentee	•	1.059		0.078	-3.987		.2157674
wemwbs_low	•	.7780143		0.000	-5.707		-2.620748
_cons	23.63433	.513762	4 46.00	0.000	22.61	L504	24.65362
C 1	cc	٦.	MC	Nila a .a	- C - h -		110
Source	SS	df	MS		of obs	=	119
Model	117.914094	2	E0 0E70472	F(2, 1: Prob >	•	=	3.03
Residual	2257.95978	2 116	58.9570472 19.4651705	R-squai		=	0.0522 0.0496
Kesiduai	2237.93976	110	19.4051705	•	squared	=	0.0332
Total	2375.87387	112	20 13/152//				
TOCAL	2373.87367	110	20.1343244	NOOC M.	JL	_	4,4117
wemwbs post	Coefficient	Std. er	r. t	P> t	Γ95%	conf.	intervall
student mentee	-1.853922	1.046	6 -1.77	0.079	-3.926	5845	.2190016
student gender	8462913	.871906	3 -0.97	0.334	-2.573	3212	.8806291
cons	22.56453	.547091	7 41.24	0.000	21.48	3095	23.64811
	· 						
Source	SS	df	MS	Number	of obs	=	119
+				F(2, 1	16)	=	6.92
Model	253.400521	2	126.70026	Prob >	F	=	0.0014
Residual	2122.47335	116	18.2971841	R-squa	red	=	0.1067
+	253.400521 2122.47335			Adj R-	squared	=	0.0913
Total	2375.87387	118	20.1345244	Root M	SE	=	4.2775

wemwbs_post   Co	efficient	Std. err.	t	P> t	[95% conf.
interval]					
student_mentee   -	1.680715	.9665785	-1.74	0.085	-3.595145
<pre>.2337159 student_difficultevent  7399037</pre>	2.334786	.8052416	-2.90	0.004	-3.929669
_cons   24.25088		.5421299			22.10337
Source   SS					= 103
Model   523.329061 Residual   1487.63235	5 97	104.665812 15.336416	Prob > R-squar	F ed	= 0.0000 = 0.2602
Total   2010.96141		19.7153079			= 0.2221 = 3.9162
wemwbs_post   conf. interval]		nt Std. err			_
student_mentee   Yes   1.123277		2 1.598197	-1.28	0.203	-5.220681
wemwbs_low   Low at baseline   -2.129732		5 .9697589	-4.18	0.000	-5.979139
student_mentee#wemwbs_low   Yes#Low at baseline 4.793004	.5035229	9 2.161249	0.23	0.816	-3.785959
student_difficultevent   1.418688	3717283	.9020987	-0.41	0.681	-2.162145
student_gender   1.470526	1641913	.8236502	-0.20	0.842	-1.798909
_cons   25.02758	23.7975	5 .6197455	38.40	0.000	22.56753

•

name: <unnamed>

log: C:\Users\quachjl\OneDrive - The University of
Melbourne\Documents\1\_Projects\1\_InProgress\2022\_RAISE Evaluation\10\_Data

<sup>.</sup> log close

Analysis\ALL\Stata\06\_outcomes\_anal
> ysis\_26 Mar 2024.log
log type: text
closed on: 26 Mar 2024, 20:36:43

```
name: <unnamed>
       log: C:\Users\quachjl\OneDrive - The University of
Melbourne\Documents\1_Projects\1_InProgress\2022_RAISE Evaluation\10_Data
Analysis\ALL\Stata\07_Restandardise
> d_analysis_26 Mar 2024.log
 log type: text
opened on: 26 Mar 2024, 20:36:43
 **************
. *** Repeating analyses based on standardisation of just matched cohort
 *************
. *** Create z-scores for change scores
. foreach var of varlist epoc_engagement_change epoc_perser_change
epoc_optimism_change epoc_connect_change SSE_aspiration_change SSE_belong_change
SSE productivity cha
> nge helpseeking_change helpseeking2_1_change helpseeking2_2_change
helpseeking2_3_change belonging_change sei_change brs_change wembs_change {
            qui sum `var'
 3.
            gen mz_`var'=((`var'-`r(mean)')/`r(sd)')
 4.
. }
(312 missing values generated)
(227 missing values generated)
(246 missing values generated)
(246 missing values generated)
(306 missing values generated)
(306 missing values generated)
(306 missing values generated)
(334 missing values generated)
(224 missing values generated)
(225 missing values generated)
(225 missing values generated)
(241 missing values generated)
(245 missing values generated)
(247 missing values generated)
(321 missing values generated)
. sum mz_epoc_engagement_change mz_epoc_perser_change mz_epoc_optimism_change
mz_epoc_connect_change mz_helpseeking2_1_change mz_helpseeking2_2_change
mz_helpseeking2_3
> _change mz_SSE_aspiration_change mz_belonging_change mz_brs_change
```

Variable		Mean	Std. dev.	Min	Max
mz_epoc_en~e		-6.85e-09	1	-2.634972	3.692891
mz_epoc_pe~e	197	2.27e-09	1	-3.45269	2.455259
mz epoc op~e	178	4.86e-09	1	-2.275458	3.339276

```
------
           199 -4.79e-09
199 6.29e-09
118 6.95e-10
183 7.28e-09
177 -2.46e-09
                           1 -2.663641 2.069078
1 -2.742306 2.003076
mz ~2 change |
mz_~3_change
                           1 -3.809928 2.471795
1 -3.724411 3.105356
1 -3.058645 2.941562
mz SSE asp~e |
mz_belongi~e |
mz_brs_cha~e |
. foreach var of varlist mz epoc engagement change mz epoc perser change
mz_epoc_optimism_change mz_epoc_connect_change mz_helpseeking2_1_change
mz helpseeking2 2 chang
> e mz_helpseeking2_3_change mz_belonging_change mz_SSE_aspiration_change
mz_brs_change {
       regress `var' _treated
regress `var' _treated##student_difficultevent
regress `var' _treated##student_gender
 2.
 3.
 4.
 5. }
                              Number of obs = 112
= 0.96
  Source | SS df MS
------ Adj R-squared = -0.0004
    Total | 110.999999 111 .999999989
                               Root MSE =
-----
mz_{epoc_{en\sim e}} | Coefficient Std. err. t P>|t| [95% conf. interval]
-----
------ Adj R-squared = 0.0041
    Total | 110.999999 111 .999999989 Root MSE =
   mz_epoc_engagement_change | Coefficient Std. err. t P>|t|
[95% conf. interval]
_treated
              Mentee | -.2450822 .4279657 -0.57 0.568
-1.093384 .6032201
     student difficultevent |
                Yes | .0312816 .2099692 0.15 0.882
-.3849137 .447477
```

_treated#stude 321868 1.		tevent   e#Yes	.764912	. 5482769	1.40	0.	166
3110702 .	. 2084067	_cons	0513317	.1310372	-0.39	0.	696
Source	SS	df	MS	Numbe	r of obs	=	112
<del>-</del>	+			- F(3, :	108)	=	0.93
		.9 3		3 Prob	> F	=	0.4283
		6 108					
	110 0000			- Adj R	-squared	=	-0.0019
lotal	110.99999	9 111	.999999989	e Root i	MSE	=	1.0009
			<b>.</b>	_	_ 1.1		F 0 = 0/
mz_epoc_engage interval]		Coefficient					_
	_treated						
	_ Mentee	2002788	.4653295	-0.43	0.668	_	1.122643
.722085							
1.stude	ent_gender	.0848502	.2156816	0.39	0.695	-	.3426681
5123685							
_treated#stude		606040	F7.470.6F	4 05	0 204		<b>533300</b> 6
	Mentee#1	.606018	.5/4/865	1.05	0.294	-	.5333086
L.745345	ı						
. 1833479	_cons	0686226	.1271183	-0.54	0.590	-	.3205932
Source	SS	df	MS	Numbe	r of obs	=	197
	+			- F(1, :	195)	=	0.13
Model	.12822036	9 1	.128220369	Prob :	> F	=	0.7213
		1 195					
Total	196.00000	1 196	1.0000000	L Root I	MSE	=	1.0022
mz_epoc_pe~e		t Std. err.			[95% coi	nf.	interval]
•	•	.1429025			332889	<b></b> 3	.2307766
_		.1028269					
Source	SS	df	MS	Numbei	r of obs	=	197
,				_			0.75

Residual	•	88	193		165 R-squared		red	=	0.0116
Total	+   196.0000						squarea SE		1.0019
[95% conf. in	epoc_perser_ terval] 			fficient					t
3698233		reated lentee		0095056	.19	23251	0.05	0.9	61
stude	ent_difficul	tevent. Yes	 	.287689	. 21	07991	1.36	0.1	74
1280767	.7034548		! 	. 207 003	•	0,332	2.50	0.1	
_treated#stude	<b>—</b>	tevent e#Yes		1847182	.28	94834	-0.64	0.5	24
7556755	.386239								
3450824	.1738583	_cons	-	.085612	.13	15551	-0.65	0.5	16
Source	SS		df	MS		Number	of obs	=	197
Model	+   .88406054			2046969		F(3, 19	93) F	=	0.29
Residual	•					R-squa	red	=	0.0045
Total	+   196.0000	)1	196	1.000000	001	Adj R-squared Root MSE			-0.0110 1.0055
mz_epoc_persinterval]	ser_change			Std. er				[	95% conf.
	·								
	_treated   Mentee		3226	.199541	.1	-0.17	0.868		4268838
.3602386 1.stud .5980103	ent_gender	.170	6869	.21665	59	0.79	0.432		2566365
_treated#stude	 ent_gender     Mentee#1	0988	8418	. 295950	18	-0.33	0.739	_	6825549
.4848714	. iciiccca	.000	0	. 20000		0.00	0.,55	•	~~~~
.2189997	_cons	032	2856	.127694	14	-0.26	0.797		2847117

Source	SS	df	MS	Number of ob		178
Model	   .384537293	1	.384537293	F(1, 176) Prob > F	=	0.38 0.5367
Residual	!			R-squared	=	0.0022
T-4-1	+		1 0000000	Adj R-square		
Total	177.000004	1//	1.00000002	Root MSE	=	1.0017
mz_epoc_op~e	Coefficient +	Std. err. 	t I	P> t  [95% 	conf.	interval]
_treated	I			0.5372038		
_cons	0497258	.109956	-0.45	0.6522667	7278	.1672762
Source	SS	df	MS	Number of ob		
M a d a 1	t		1 25062760	F(3, 174) Prob > F	=	
Model Residual	3.75188306 173.248121		1.25062769			0.2911 0.0212
	+			Adj R-square		
Total	177.000004	177	1.00000002	Root MSE	=	.99784
mz_epo	oc_optimism_cha	ange   Coe	efficient S	td. err. t	. P>	t
[95% conf. in	_					
		+				
	trea	ated				
	_ Men	ated   tee   .	2676491 .:	2026408 1.3	32 0.	188
1323013	_ Men	•	2676491 .:	2026408 1.3	32 0.	188
	 Men <sup>-</sup> .6675995	tee       . 	2676491	2026408 1.3	32 0.	188
	Men <sup>-</sup> .6675995 ent_difficulte	tee   .   vent				188 068
	Men- .6675995 ent_difficulte	tee   . vent				
stude 0302699	Mendendendendendendendendendendendendende	tee   . vent   Yes   .				
stude 0302699	Men- .6675995 ent_difficulter .8488952 ent_difficulter	tee   . vent   Yes   . vent	4093127		34 0.	068
stude 0302699	Men- .6675995 ent_difficulter .8488952 ent_difficulter Mentee#	tee   . vent   Yes   . vent	4093127	.222721 1.8	34 0.	068
stude 0302699 _treated#stude	Men- .6675995 ent_difficultev .8488952 ent_difficultev Mentee# .1740534	vent   vent   ves   vent   vent   ves	4093127 4230743 .:	.222721 1.8 3025436 -1.4	34 0. 10 0.	068 164
stude 0302699 _treated#stude -1.020202	Men6675995  ent_difficulte .8488952  ent_difficulte Mentee# .1740534	vent   vent   ves   vent   vent   ves	4093127 4230743 .:	.222721 1.8	34 0. 10 0.	068 164
stude 0302699 _treated#stude	Men6675995  ent_difficulte .8488952  ent_difficulte Mentee# .1740534	vent   vent   ves   vent   vent   ves	4093127 4230743 .:	.222721 1.8 3025436 -1.4	34 0. 10 0.	068 164
stude 0302699 _treated#stude -1.020202	Men6675995  ent_difficulte .8488952  ent_difficulte Mentee# .1740534	vent   vent   ves   vent   vent   ves	4093127 4230743 .:	.222721 1.8 3025436 -1.4	34 0. 10 0.	068 164
stude0302699 _treated#stude -1.0202024987421	Mennone Mennone Mennone Mennone Mentoulte Ment	vent   vent   vent   vent   vent   vent   cons	4093127 4230743 .3 2173961 .3	.222721 1.8 3025436 -1.4 1425482 -1.5	34 0. 40 0. 53 0.	<ul><li>068</li><li>164</li><li>129</li><li></li></ul>
stude0302699 _treated#stude -1.0202024987421	Men6675995  ent_difficultev .8488952  ent_difficultev	vent   vent   vent   vent   vent   vent   df	4093127 4230743 .: 2173961 .: MS	.222721 1.8 3025436 -1.4 1425482 -1.5 Number of ob F(3, 174)	34 0. 40 0. 53 0. 	068 164 129  178 2.16
stude0302699 _treated#stude -1.0202024987421	Men6675995  ent_difficultev .8488952  ent_difficultev	vent   vent   vent   vent   vent   vent   des   df	4093127  4230743 .3  2173961 .3  MS  2.12274751	.222721 1.8 3025436 -1.4 1425482 -1.5 Number of obe F(3, 174) Prob > F	34 0. 40 0. 53 0. 	068  164  129   178  2.16  0.0939
stude0302699 _treated#stude -1.0202024987421	Men6675995  ent_difficultev .8488952  ent_difficultev	vent   vent   vent   vent   vent   vent   des   df	4093127  4230743 .3  2173961 .3  MS  2.12274751 .980642307	.222721 1.8 3025436 -1.4 1425482 -1.5 Number of ob F(3, 174) Prob > F R-squared	34 0. 40 0. 53 0. 	068  164  129   178  2.16  0.0939  0.0360
stude0302699 _treated#stude -1.0202024987421	Mennoments	vent   vent   vent   vent   vent   vent   des   df	4093127  4230743 .3  2173961 .3  MS  2.12274751 .980642307	.222721 1.8 3025436 -1.4 1425482 -1.5 Number of ob F(3, 174) Prob > F R-squared Adj R-square	34 0. 40 0. 53 0. 	068  164  129   178  2.16  0.0939  0.0360  0.0194
stude0302699 _treated#stude -1.0202024987421	Men6675995  ent_difficultev .8488952  ent_difficultev	vent   vent   vent   vent   vent   vent   des   df	4093127  4230743 .3  2173961 .3  MS  2.12274751 .980642307	.222721 1.8 3025436 -1.4 1425482 -1.5 Number of ob F(3, 174) Prob > F R-squared Adj R-square	34 0. 40 0. 53 0. 	068  164  129   178  2.16  0.0939  0.0360  0.0194

mz_epoc_optim: interval]			nt Std.err.			_
		+				
	_treated	I				
	_ Mentee	-	4 .2057812	1.09	0.276	1814592
.6308381						
1.stud 1.011073	ent_gender	.55/32	7 .2298974	2.42	0.016	.1035806
1.0110/3		I				
_treated#stude	ent_gender	İ				
	Mentee#1	459744	7 .3085429	-1.49	0.138	-1.068713
.1492237		ı				
	cons	   _ 227730	8 .1335285	_1 70	0 077	_ 5012939
.0258043	_cons	-,23//33	.1333203	-1.70	0.077	5012050
C			C MC	Ni. anda a a		170
Source	•	d 			of obs	= 178 = 4.32
Model			1 4.24430402			= 0.0390
Residual			6 .981566474		ared	
				,	-squared	= 0.0184
Total	177.0000	<b>2</b> 3 17	7 1.00000002	Root M	1SE	= .99074
mz_epoc_co~e	Coefficie	nt Std. er	r. t	P> t	[95% con	f. interval]
	+					
	.309536	9 .148856	9 2.08	0.039	.0157626	.6033111
_cons	165202	3 .108747	9 -1.52	0.131	3798199	.0494154
Source	l ss	ď	f MS	Number	of obs	= 178
	•			F(3, 1	L74)	= 2.53
Model	7.387080	<b>39</b>	3 2.46236003	Prob :	> F	= 0.0591
Residual	169.6129	23 17	4 .974786915	R-squa	ared	= 0.0417
Total	+   177 0000	 23 17	7 1.00000002	Root N	-squarea NSF	= 0.0252 = .98731
TOCAL		J 1/			1JL	50/51
	177.0000			11000		
mz_e	  ooc_connect					
mz_e <sub> </sub> [95% conf. in	  ooc_connect <sub>.</sub> terval]		oefficient S	 td. err.	t	P> t
mz_e <sub> </sub> [95% conf. in	 ooc_connect terval]		oefficient S	 td. err.	t	
mz_e <sub>l</sub> [95% conf. in	 ooc_connect terval] 	 _change	oefficient S	 td. err.	t	P> t
mz_e <sub> </sub> [95% conf. in	 ooc_connect terval] 	 _change	oefficient S	 td. err.	t	P> t
mz_e <sub>l</sub> [95% conf. in	 ooc_connect terval] 	 _change	oefficient S	 td. err.	t	P> t
mz_e <sub> </sub> [95% conf. in	ooc_connect terval]   	_change   C _change   C + treated   Mentee	oefficient S	 td. err.	t	P> t
mz_e <sub> </sub> [95% conf. in	 ooc_connect terval] 	_change   C _change   C + treated   Mentee	oefficient S	 td. err.  2005035	t 2.57	P> t

cc 59904630422888	· 	3206676	1410447	-2.27	0 024
	<i>م</i> د				0.024
	طد				
Source   SS	uт 	MS		of obs 74)	= 178 = 1.87
Model   5.5398823 Residual   171.460121	174	.985402995	Prob > R-squa	F red	= 0.1357 = 0.0313
Total   177.000003					= .99267
mz_epoc_connect_change   Coeinterval]		Std. err.			_
_treated   Mentee   . .6377395	. 2306062	.2062801	1.12	0.265	1765272
1.student_gender   .	.1242641	.2304547	0.54	0.590	3305824
.5791106					
 _treated#student_gender					
Mentee#1	.0844788	.3092909	0.27	0.785	5259661
.6949236 					
_cons   .0570603	. 2071227	.1338522	-1.55	0.124	4713057
Source   SS		MS			
Model   8.01918115	1	8.01918115	Prob >	F	
Residual   190.980819	198	.964549588	R-squa	red	= 0.0403
Total   199	199	.999999998	Аај К- Root M	squared SE	= 0.0355 = .98211
mz_~1_change   Coefficient S		t	 P> t  	 [95% con	ıf. interval]
_treated   .4004993 _cons  2022521 .	.138899	2.88	0.004	.1265881	.6744104
Source   SS	df	MS			= 200 = 3.13

Residual	9.0924322 189.90756	7	196		516	R-squai	red	=	0.0269 0.0457
Total							squared SE		.98434
[95% conf. int	oseeking2_1_ terval]		-	fficient					:
0913954	М	reated entee		2775901	.18	70989	1.48	0.14	10
stude	ent_difficul		•	2624022	20	44062	0.20	0.74	- 4
4654589	. 3410924	Yes	6	9621832	.20	44862	-0.30	0.76	01
_treated#stude		tevent e#Yes	•	2595669	.28	31807	0.92	0.36	50
2989053	.8180392		1						
4255507	.0675269	_cons	:	1790119	.12	50107	-1.43	0.1	54
Source	SS		df	MS			of obs		
Model Residual					316	Prob > R-squai	96) F red	= =	0.0356 0.0427
Total	   19	9	199	.9999999	998		squared SE		
mz_helpseeking interval]								[9	95% conf.
	+								
	_treated   Mentee	.308	7627	.194599	94	1.59	0.114	6	750148
.6925403 1.stude .309096	ent_gender	105	4312	.210191	L <b>4</b>	-0.50	0.617	5	5199583
_treated#stude	 ent_gender   Mentee#1	. 202	3104	.289476	54	0.70	0.485		. 368578
.7731989	1								
.0722189	_cons	167	1084	.121354	11	-1.38	0.170	4	1064358

Source	SS .	df	MS		er of obs	=	199
Model	   7.0910831	 1	7.0910831	- F(1, 1 Prob	•	=	7.32 0.0074
Residual	!						0.0358
	+			- Adj R	k-squared		0.0309
Total	198.000001	198	:	1 Root	MSE	=	.98442
mz_~2_change	Coefficient	Std. err.	t	P> t	[95% cor	ıf.	interval]
_treated	.377542	.139569	2.71	0.007	.1023009	)	.652783
_cons	•						
Source	l cc	df	MS	Numbo	er of obs	=	199
	SS +	uı 	כויו		195)		2.61
Model	7.65042355	3	2.55014118	8 Prob	> F	=	0.0526
Residual	190.349577	195	.976151678	-			0.0386
	+			•	-squared		
Total	198.000001	198	:	1 Root	MSE	=	.988
	oseeking2_2_cha	ange   Coe	fficient S	Std. err.	t	P>	t
[95% conf. in	=						
		+					
	trea	ated					
	_trea Ment		3150733	.188854	1.67	0.	097
0573854	 Ment		3150733	.188854	1.67	0.	097
	 Ment .6875319	tee   .	3150733	.188854	1.67	0.	097
	Ment .6875319 ent_difficultev	tee   . /ent					
stude	Ment .6875319 ent_difficulte	tee   . /ent		.188854	1.67 0.16		
stude	Ment .6875319 ent_difficultev	tee   . /ent					
stude 3722979	Ment .6875319 ent_difficulte .4372851 ent_difficulte	/ent   . /es   . /ent	0324936	. 2052483	0.16	0.	874
stude 3722979 _treated#stude	Ment .6875319 ent_difficulte .4372851 ent_difficulte Mentee#	/ent   . /es   . /ent		. 2052483	0.16	0.	874
stude 3722979	Ment .6875319 ent_difficulte .4372851 ent_difficulte Mentee#	/ent   . /es   . /ent	0324936	. 2052483	0.16	0.	874
stude 3722979 _treated#stude	Ment .6875319  ent_difficultev .4372851  ent_difficultev	/ent   . /es   . /ent   /es   . /ent	.113806	.2052483	0.16 0.40	0.	874 690
stude 3722979 _treated#stude	Ment .6875319  ent_difficulte .4372851  ent_difficulte Mentee# .675758	/ent   . /es   . /ent   /es   . /ent	0324936	.2052483	0.16 0.40	0.	874 690
stude 3722979 _treated#stude 4481461	Ment .6875319  ent_difficulte .4372851  ent_difficulte Mentee# .675758	/ent   . /es   . /ent   /es   . /ent	.113806	.2052483	0.16 0.40	0.	874 690
stude 3722979 _treated#stude 4481461	Ment .6875319  ent_difficulte .4372851  ent_difficulte Mentee# .675758	/ent   . /es   . /ent   /es   . /ent	.113806	.2052483	0.16 0.40	0.	874 690
stude3722979 _treated#stude44814614493292	Ment .6875319  ent_difficulte .4372851  ent_difficulte Mentee# .675758	/ent   . /ent   . /es   . /ent   . /es   .	0324936 .113806 2018637	.2052483 .284936 .1254766	0.16 0.40 -1.61	0.: 0.:	874 690 109
stude 3722979 _treated#stude 4481461	Ment .6875319  ent_difficulte .4372851  ent_difficulte Mentee# .675758	/ent   . /ent   . /es   . /ent   . /es   .	.113806	.2052483 .284936 .1254766 	0.16 0.40 -1.61	0.9	874 690 109 
stude3722979 _treated#stude44814614493292	Ment .6875319  ent_difficultev .4372851  ent_difficultev	/ent	0324936 .113806 2018637	.2052483 .284936 .1254766  Numbe - F(3, 2 Prob	0.16 0.40 -1.61 	<ul><li>0.</li><li>0.</li><li>1.</li><li>2.</li><li>3.</li><li>4.</li><li>4.</li><li>5.</li><li>6.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.</li><li>7.&lt;</li></ul>	874 690 109  199 3.23 0.0235
stude3722979 _treated#stude44814614493292	Ment .6875319  ent_difficulte .4372851  ent_difficulte Mentee# .675758  .0456018   SS +   9.37805136   188.621949	/ent   . /ent   . /es   . /ent   . /es   . /ent   . /es   . /es   . /es   .	0324936 .113806 2018637	.2052483 .284936 .1254766  Numbee - F(3, 2 Prob 8 R-squ	0.16  0.40  -1.61  er of obs 195)  > F Hered	<ul><li>0.3</li><li>0.3</li><li>= = = = = = = = = = = = = = = = = = =</li></ul>	874 690 109  199 3.23 0.0235 0.0474
stude3722979 _treated#stude44814614493292	Ment .6875319  ent_difficultev .4372851  ent_difficultev	/ent	0324936 .113806 .2018637	.2052483 .284936 .1254766 Numbe - F(3, 2 Prob 8 R-squ - Adj R	0.16  0.40  -1.61  er of obs 195) > F Jaredsquared	0.0 0.0 = = = = = = = = = = = = = = = = = = =	874 690 109  199 3.23 0.0235 0.0474 0.0327
stude3722979 _treated#stude44814614493292	Ment .6875319  ent_difficulte .4372851  ent_difficulte Mentee# .675758  .0456018   SS +   9.37805136   188.621949	/ent	0324936 .113806 .2018637	.2052483 .284936 .1254766 Numbe - F(3, 2 Prob 8 R-squ - Adj R	0.16  0.40  -1.61  er of obs 195)  > F Hered	0.0 0.0 = = = = = = = = = = = = = = = = = = =	874 690 109  199 3.23 0.0235 0.0474 0.0327

mz_helpseeking interval]	g2_2_cha~e		Std. err.			
	+					
	_treated   Mentee		.1941306	1.13	0.258	1626196
.6031108						
1.stude .4673226	ent_gender	.0537809	.2096851	0.26	0.798	3597608
_treated#stude	ent_gender   Mentee#1	.2483283	.2892681	0.86	0.392	3221675
.818824						
.0311119	_cons	2076466	.1210617	-1.72	0.088	446405
Source	•	df			of obs	
Model		7 1	3.76345867		97) F	= 0.0522
Residual	194.23654	9 197	.985972332	R-squar		
Total	198.00000	8 198	1.00000004	_	squared SE	
mz_~3_change	Coefficien +	it Std.err.		?> t  	95% con[ 	f. interval]
_treated _cons	!		1.95 6 -1.39 6			.5527296 .0581985
Source	•			Number		
Model	5.8099602	.8 3	1.93665343	Prob >	95) F	= 0.1206
Residual	192.19004	8 195	.985589988	R-squar	red	= 0.0293 = 0.0144
Total	198.00000	8 198	1.00000004	Root MS	squareu SE	= 0.0144
mz heli	 nseeking2 3	change   Coe	efficient St	d. err.	t	P> +
[95% conf. in	terval]	-				
		reated	204506	1002012	1 00	0. 201
1689221		iencee	.204596 .1	1033317	1.08	0.201
c+d.	ont difficul	toyont				
Stude	ent_difficul		. 1455993 .	. 206869	0.70	0.482
2623885	.5535872					

_treated#stud	_	  tevent    e#Yes		1028354	. 2	2860627	0.36	0.7	20
4613387	.6670094								
4452695	.0561083	_cons		1945806		.127111	-1.53	0.1	27
Source	•		df 	MS 			of obs		
Residual	•	7 :	195	.989757	369	R-squai	red	=	1.68 0.1720 0.0252
Total	+   198.00000						squared SE		
mz_helpseekin	g2_3_cha~e					t		[	95% conf.
	+								
510110	_treated   Mentee	.1216	765	.19695	86	0.62	0.537		2667661
.510119 1.stud .2894151	ent_gender	1299	726	.21264	92	-0.61	0.542		5493602
_treated#stud	 ent_gender     Mentee#1	.3170	643	.29250	79	1.09	0.279		2592423
.8945283	ı								
.1475227	_cons	09584	431	.12339	79	-0.78	0.438		3392089
Source	SS		df	MS		Number	of obs	=	183
Model Residual	2.1812660   179.81873	9	1 181	2.18126 .99347	609 367	Prob > R-squa	F red	= = =	0.1401 0.0120
Total	+	2 :	 182		1	Adj R-S Root MS	squared SE	= =	0.0065 .99673
mz_belongi~e	   Coefficien						 [95% co	 nf. i	nterval]
_treated	.2185127 1134356	.1474	689	1.48	(	0.140	072466 323087	7 3	.5094921 .0962162
Source	   SS +					Number F(3, 1			183 3.77

Residual	10.824134 171.17586	48 3 3.6080449 65 179 .95628979		51	R-squared		=	0.0595	
Total							Squareu SE		
[95% conf. int	 z_belonging_ terval] 	_		fficient				-	:
	 t	reated	I						
.075495 .84	M	entee		4582948	.19	39891	2.36	0.01	19
	ent_difficul	tevent	 						
	_ .065816	Yes	:	6433477	.21	40917	3.01	0.00	93
_treated#stude	ent difficul	tevent	 						
-1.204631	_ Mente	e#Yes	•	6255926	.29	34357	-2.13	0.03	34
6246002	.0994033	_cons	  :	3620017	.13	30754	-2.72	0.06	97
Source	SS		df	MS			of obs		
Model     Residual	   4.7671819   177.23281				66	Prob >	79) F red	=	0.1899
Total	- 	182 182				Adj R-squared		=	0.0099
mz_belongi interval]	ing_change			Std. er				[9	95% conf.
	+								
6052545	_treated   Mentee	.29	7392	.201622	4	1.47	0.142	1	1004705
.6952545 1.stude .8106892	ent_gender	.36	1296	.227736	3	1.59	0.114	6	880973
_treated#stude	ent_gender   Mentee#1	297	8244	.307162	8	-0.97	0.334	9	9039504
.0250986	_cons	228	3934	.128460	6	-1.78	0.077	4	1818854

Source	ss	df	MS	Number of obs	= 118
	.030542393				= 0.8621
Residual	116.969456	116	1.00835738	it squarea	- 0.0005
Total	116.999998	117	.999999985	Adj R-squared Root MSE	
mz_SSE_asp~e	Coefficient	Std. err.	t P	?> t  [95% co	nf. interval]
+noatod	t   015015	 2622469	0 17 0	 0.862567208	 Л ИЗБЕЗОБ
_cons	•			0.94719130	
Source	l ss	df	MS	Number of obs	= 118
	<del> </del>			F(3, 114)	= 0.12
	.372833003   116.627165			Prob > F R-squared	= 0.9473 = 0.0032
	+			Adj R-squared	= -0.0230
Total	116.999998	117	.999999985	Root MSE	= 1.0115
mz_SSE_ [95% conf. int				d. err. t	
		+			
	 trea	ated			
	_ Men	:	0624603 .4	318479 -0.14	0.885
9179477	.7930271	1			
stude	ent_difficulte	vent			
	_ ,		1150331 .2	0.55	0.583
298776 .5	5288422	1			
_treated#stude	ent_difficulte	vent			
		Yes	<b>0241817</b> .5	5542081 -0.04	0.965
-1.122064	1.0/3/	1			
2860014	_	cons	0355404 .1	.264321 -0.28	0.779
Source			MS		
Model	+   2 0020021	 2	1 20/20/27	F(3, 114) Prob > F	= 1.30
Residual	113.117115	3 114	.992255396	R-squared	= 0.0332
	+			Adj R-squared	= 0.0077
Total	116.999998	117	.999999985	Root MSE	= .99612

mz_SSE_aspiratinterval]						_
.3815038	_treated   Mentee		.4615659 .2113299	-1.15	0.251	-1.447213
_treated#stude	ent_gender   Mentee#1		.5707884	0.96	0.338	5816492
.1601934	_cons	0791051	.1207973	-0.65	0.514	3184037
Source			MS		of obs	
Model	1.3603874 174.63961	12 1 11 175	1.36038742 .997940636	Prob > R-squa	red	= 1.36 = 0.2446 = 0.0077 = 0.0021
Total	175.99999	99 176	.999999993	Root M	SE	= .99897
mz_brs_cha~e	Coefficier	nt Std. err.	t P>	> t	[95% con	nf. interval]
	.175677	nt Std. err. 7 .1504652 4 .1096512	1.17 0.	. 245	 1212829	.472637
_treated	.175677  0932974 	.1504652 1 .1096512 	1.17 0. -0.85 0. -MS	.245 .396 	 1212829 3097063 	.472637 3 .1231115 
_treated   _cons   Source   Model   Residual	.175677  0932974 	df 2 3 26 173	1.17 00.85 0MS 1.57632417 .990005932	.245 .396 	 1212829 3097063  of obs 73) F	.472637 3 .1231115 = 177 = 1.59 = 0.1930 = 0.0269
_treated   _cons   Source   Model   Residual	.175677  0932974   SS   SS   4.7289725   171.27102	df 2 3 26 173	1.17 00.85 0.  MS 1.57632417 .990005932	.245 .396  Number F(3, 1 Prob > R-squa		.472637 3 .1231115 = 177 = 1.59 = 0.1930 = 0.0269 = 0.0100
_treated   _cons   Source   Model   Residual   Total	.175677  0932974   -SS   4.7289725   171.27102	df 2 3 26 173 29 176	1.17 00.85 00.8	.245 .396  Number F(3, 1 Prob > R-squal Adj R-s		.472637 3 .1231115 = 177 = 1.59 = 0.1930 = 0.0269 = 0.0100 = .99499
treated  cons   Source   Model   Residual   Total	.175677  0932974  0932974   SS   SS   4.7289725   171.27102   175.99999	df	1.17 00.85 0.  MS  1.57632417 .990005932999999993	.245 .396 		.472637 3 .1231115 = 177 = 1.59 = 0.1930 = 0.0269 = 0.0100 = .99499
treated  cons   Source   Model   Residual   Total	.175677  0932974  0932974   SS   4.7289725   171.27102   175.99999   mz_brs_   terval]	df	1.17 00.85 0. MS -1.57632417 .990005932999999993	.245 .396 		
	.175677  0932974  0932974   SS   4.7289725   171.27102   175.99999   mz_brs_   terval]	df .1504652 df .1096512 df	1.17 00.85 0. MS -1.57632417 .990005932999999993	.245 .396 		

_treated#stude	_ Mente	tevent   e#Yes	0781439	.302	4069	-0.26	0.796
278341 .2	2827686	_cons   .	0022138	.142	1415	0.02	0.988
Source	•						= 177
Model	4.3966822	1 3 6 173	1.465560	74 85	Prob > R-squa	F <sup>°</sup> red	= 1.48 = 0.2224 = 0.0250 = 0.0081
		9 176					= .99596
mz_b	ors_change	Coefficient					[95% conf.
	+						
	_treated   Mentee	.2635261	.205495	4	1.28	0.201	1420748
.6691271 1.stude .8508623	ent_gender	.3944942	.231216	3	1.71	0.090	061874
_treated#stude	ent_gender   Mentee#1	3142691	.31043	5	-1.01	0.313	9269969
.038687	_cons	2263798	.134294	6	-1.69	0.094	4914465
log close     name: <unnamed>         log: C:\Users\quachjl\OneDrive - The University of  Melbourne\Documents\1_Projects\1_InProgress\2022_RAISE Evaluation\10_Data Analysis\ALL\Stata\07_Restandardise &gt; d_analysis_26 Mar 2024.log     log type: text closed on: 26 Mar 2024, 20:36:44</unnamed>							



## **Learning Intervention**

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