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Abstract

Adolescent mental health is key for later well-being. Yet, causal evidence on environmental drivers of adolescent mental health is scant. We study how an important classroom feature—the gender composition in compulsory-school—affects mental health. We use Swedish administrative data ($N = 576,285$) to link variation in gender composition across classrooms within cohorts to mental health. We find that a higher share of female peers in a classroom increases the incidence of mental health diagnoses, particularly among boys. The effect persists into adulthood. Peer composition is thus an important and persistent driver of mental health.

JEL Classifications: I12, I19, I21, I31, J16, J24

Keywords: school, gender, peer effects, mental health

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1 Introduction

Mental health problems among adolescents are common, particularly among boys, and have large long-run negative impacts.¹ Individuals with bad mental health have lower income, non-cognitive skills, educational attainment, and invest less in their children (Currie and Stabile, 2006; Lundborg, Nilsson and Rooth, 2014; Baranov et al., 2020).² To better understand human capital accumulation and to devise policy interventions, it is crucial to know whether and to what extent the environment in adolescence affects mental health (Cunha, Heckman and Schennach, 2010; Heckman, Pinto and Savelyev, 2013).

Children spend much of their time with their peers at school. Psychological research suggests an intimate link between peer relationships and mental health: Roughly half of the children with conduct disorders have problems with peers (Parker et al., 2015). Gender composition is a key feature of peer environment and an important correlate of school well-being and motivation (Belfi et al., 2012; Pahlke, Hyde and Allison, 2014). Gender composition creates constraints for socialization and friendship formation, affects self-image, and shapes conflict.³ Because of the dearth of large-scale diagnostic data and the difficulties of conducting randomized trials, however, findings examining the link between the school environment and mental health remain largely correlational.

In this paper, we study the causal impact of the gender composition in the compulsory school classroom on mental health. We use a unique and barely exploited link between classroom peer composition and mental health diagnoses based on administrative data of 576,285 students. A key advantage of the data is that, in addition to knowing which cohort a student was in, we know the classroom for each student within their cohort. Our identification

¹According to the Center for Disease Control, suicide is currently the second most prominent cause of death for men under the age of 20 (Heron, 2016).

²Papers examining the negative impacts of mental health problems early in life include Currie and Stabile (2006); Currie et al. (2010); Salm and Schunk (2012); De Neve and Oswald (2012); Lundborg, Nilsson and Rooth (2014). For papers discussing the effects of bad mental health across different stages in life on productivity, see, e.g., Currie and Stabile (2006); Lundborg, Nilsson and Rooth (2014); Borowiecki (2017); Prinz et al. (2018); Bütikofer, Cronin and Skira (2020); Ridley et al. (2020). Individuals with mental health issues are also more likely to commit crimes (Blattman, Jamison and Sheridan, 2017). One reason for the strong impacts of mental health on economic outcomes may be the intimate link between mental health, economic preferences, and beliefs (Haushofer and Fehr, 2014; De Quidt and Haushofer, 2019; Meier, 2019).

³Cillessen and Mayeux (2004); Sullivan, Farrell and Kliewer (2006); Belfi et al. (2012) detail links between peer relationships and mental health diagnoses (see also, Viner et al., 2012). A large literature examines how gender composition affects classroom dynamics and identity for formation, see, e.g., Simmons and Rosenberg (1975); Benenson, Apostoleris and Parnass (1997); Rose and Rudolph (2006); Dijkstra, Lindenberg and Veenstra (2007); Faris and Felmlee (2011); Kornienko et al. (2016).

strategy thus relies on both idiosyncratic variation of the classroom share of female peers across classrooms within cohorts and cohort variation.

We find that students in classrooms with more girls have more mental health diagnoses. A 10 percentage point ($sd=9pp$) increase in the share of female peers increases the likelihood of a mental health diagnosis by 2.2%. Boys drive this effect: for them, a 10 percentage point increase in the share of girls in the classroom leads to a 4.5% increase in the likelihood of receiving a mental health diagnosis. A higher share of female peers also substantially increases the risk of self-harm among boys. The mental health effect is not only substantial, it also persists into early adulthood.

The extensive administrative data allows us to conduct a series of balancing and robustness checks. We observe balance for parental characteristics, including mental health, education, and income. Importantly, student mental health before class formation also does not predict the share of females in the classroom. Moreover, student gender does not predict the share of female peers, classroom fixed effects do not predict student gender (Chetty et al., 2011; Balestra, Eugster and Liebert, 2020), and the distribution of the share of female peers looks as if randomly assigned. We also do placebo checks which show i) no impact of previous or following cohorts on mental health and ii) no impact on mental health diagnoses related to biological origins. Finally, the coefficient estimates are equivalent when we drop students who move during the sample period.

An advantage of the data is that we can exploit cohort and classroom variation. We also document sizable and statistically significant effects when we rely only on cohort variation. Theory and evidence highlights the classroom as the relevant peer environment (Lazear, 2001; Chetty et al., 2011), but data limitations have traditionally hindered the use of classroom-level variation. Exceptions include Ammermüller and Pischke (2009); Burke and Sass (2013); Balestra, Eugster and Liebert (2020); Bietenbeck (2020), who use naturally occurring classroom assignment to estimate the effects of peers like we do. Given the theoretical arguments and the empirical evidence highlighting classrooms as the relevant peer environment, our focus lies on this variation.

We explore mechanism which could be responsible for the impact of peer composition on mental health. In the first step, we document an increase of diagnoses related to peer problems. A marked increase in conduct disorders and learning and socialization disorders drives the impact on mental health. Both diagnostic types are intimately linked to peer

relationships (Parker et al., 2015; Ackermann et al., 2019), suggesting personal relationships as a key factor for explaining the effects of peer gender composition. The impact on specific mental health diagnoses and other results, such as the differential impact between boys and girls, indicate that the overall effect does not simply result from a general increase in diagnoses.

Second, we highlight an obvious explanation for the link between peer relationships and mental health: Boys have smaller potential friend networks than girls in classes with more girls. Same-gender friendships in adolescence are much more common than opposite-gender friendships (see, e.g., Hill, 2015), and having more friends benefits mental health (see, e.g., Parker and Asher, 1987; Bond et al., 2007; Ho, 2016). Finding many friends is particularly relevant for boys, as they tend to have larger same-gender networks than girls (Benenson, Apostoleris and Parnass, 1997; Rose and Rudolph, 2006). Moreover, boys have closer relationships to boys than to girls (Hill, 2015). Accordingly, boys have a tougher time finding close friends in classes with more female peers, with detrimental consequences for mental health.

Third, we examine educational attainment and average peer ability as well as mental health. The average grade in the rest of the class or average mental health cannot explain the results. However, consistent with the impacts on mental health, we observe lower educational attainment of boys in classes with more females. The difference in own educational attainment accounts for roughly a third to half of the relationship between class composition and mental health.

Fourth, we study bullying, which is a key factor in shaping peer relationships (Ammermüller, 2012; Eriksen, Nielsen and Simonsen, 2014; Klomek, Sourander and Elonheimo, 2015; Rees, Sabia and Kumpas, 2020). Economic research has documented that boys and girls report less fighting with more girls in the cohort, according to Lavy and Schlosser (2011).⁴ Our evidence is consistent with these findings: More girls in the classroom reduce the incidence of physical injury and have no effect on assaults. Accordingly, physical bullying can not explain the higher incidence of mental health diagnoses when there are more females in the classroom. While physical bullying seems lower, psychological bullying may matter as females engage more in relational aggression (Lagerspetz, Björkqvist and Peltonen, 1988; Björkqvist, Lagerspetz and Kaukiainen, 1992; Crick and Grotpeter, 1995; Card et al., 2008; Smith, Rose

⁴Giardili (2020) reports a lower incidence of bullying in single-sex classes consisting only of boys and Lee et al. (2014) estimate a non-statistically significant reduction in bullying in single-sex classes.

and Schwartz-Mette, 2010). Psychological bullying also has a disproportionate effect on boys (Crick and Grotpeter, 1996; Goldstein, Young and Boyd, 2008; Williams et al., 2009). Taken together, while physical bullying does not seem to explain the peer effects, psychological bullying might offer a potential explanation.

Taken together, the exploration of mechanisms indicates that i) diagnostic types with intimate links to peer relationships drive the main results, suggesting peer relationships as the key driver of the results, ii) boys prefer large friendship networks than girls which will be difficult to attain in classrooms with more girls, providing a straightforward explanation for lower mental health, iii) educational attainment in classes with more females is lower for boys, explaining part of the co-variation between classroom composition and mental health, and iv) psychological bullying might play a role.

We add to three strands of literature. First, we add to a growing economic literature on early determinants of mental health (see, e.g, Black, Devereux and Salvanes, 2016; Black, Grönqvist and Öckert, 2018). Persson and Rossin-Slater (2018) document the role of the family environment for mental health: In-utero exposure to negative family events has lasting negative effects on children’s mental health. Studying the impact of economic shocks, Adhvaryu, Fenske and Nyshadham (2019) document lower adult mental health after negative agricultural shocks in early life, and Baird, de Hoop and Özler (2013) document better adolescent mental health after cash transfers. Moreover, Fruehwirth, Iyer and Zhang (2019) find that more religious students tend to have fewer mental health problems and Singhal (2019) shows that children exposed to war have more. In sum, the literature indicates that early economic and family circumstances can have lasting effects on mental health.⁵ We add evidence that school environment is also a persistent driver of mental health.

⁵Two concurrent working papers use survey data from AddHealth to estimate the impact of cohort composition in middle and high school on answers to survey items from the Center for Epidemiological Studies Depression Scale (CES-D). Giulietti, Vlassopoulos and Zenou (2020) find that going to school with a higher share of peers with low well-being is linked to lower reported well-being. Kiessling and Norris (2020) show that students’ ability rank within a cohort is linked to well-being: a higher ability rank relates to higher reported well-being. In related studies, Eisenberg et al. (2013) finds small, if any, impact of peer mental health on own mental health in college students, and Bütikofer et al. (2020) show that students admitted to more selective high schools have less mental health problems. We provide complementary evidence on the impact of the compulsory school environment in early adolescence and focus on the peer gender composition. Further differences to Giulietti, Vlassopoulos and Zenou (2020); Kiessling and Norris (2020) include the use of diagnostic data from comprehensive administrative registers, including data *before* school start, classroom-level composition (in addition to cohort-level), and the population-wide coverage of students (AddHealth covers roughly 20,000 students).

Second, we complement the literature on the impact of peers on educational and labor-market outcomes (see, e.g., Chetty et al., 2011; Eisenkopf et al., 2015; Feld and Zölitz, 2017; Carrell, Hoekstra and Kuka, 2018; Balestra, Eugster and Liebert, 2020; Balestra, Sallin and Wolter, 2020; Bietenbeck, 2020; Golsteyn, Non and Zölitz, 2020; Thiemann, 2020). Several papers examine the impact of the gender composition of school peers on educational attainment (Lavy and Schlosser, 2011; Park, Behrman and Choi, 2012; Black, Devereux and Salvanes, 2013; Eisenkopf et al., 2015; Giardili, 2020) and educational choices (Black, Devereux and Salvanes, 2013; Anelli and Peri, 2019; Brenøe and Zölitz, 2019).⁶ Focusing on long-run labor-market outcomes, Getik and Meier (2020) study the impact of more female peers on gender wage inequality at age 30. They document that more females in the classroom reduce wage inequality because females select into less gender-stereotypical occupations. Taken together, previous literature has examined the impact of the school environment on educational and labor-market outcomes, but evidence on the impact on mental health is scant.

Third, our paper complements the literature studying peer effects on risky behaviors (see, e.g., Kremer and Levy, 2008; Cawley and Ruhm, 2011; Card and Giuliano, 2013). Several papers demonstrate important peer effects on alcohol consumption and smoking (Lundborg, 2006; Elsnor and Isphording, 2018). Among other effects, peers who take drugs may induce others to consume drugs as well or may act as gatekeepers (Clark and Lohéac, 2007; Meier, Odermatt and Stutzer, 2020).

The paper is structured as follows: Section 2 describes the Swedish schooling system, mental health care provision, and the data. Section 3 discusses the empirical strategy and the main identifying assumption. It also provides evidence on balancing and placebo checks. Section 4 and Section 5 show the main results on early and later mental health. Section 6 discusses mechanisms, and Section 7 concludes.

⁶Black, Devereux and Salvanes (2013) document lower educational attainment for boys with more girls in the cohort, while Lavy and Schlosser (2011) document higher attainment for boys and girls with more girls in the cohort. The latter finding is consistent with some findings exploiting random assignment to single-sex schooling (Park, Behrman and Choi, 2012; Giardili, 2020). Regarding educational choices, Brenøe and Zölitz (2019) find that female high-school students who joined the math track are more likely to make gender-congruent choices when exposed to more boys. Anelli and Peri (2019) find no impact, and Giardili (2020) finds that female students in single-sex schools make less gender-congruent choices.

2 Institutional Background and Data

2.1 Compulsory Education in Sweden

All children in Sweden have to complete nine school years with a standardized curriculum across schools and classes. Students are generally assigned to the nearest compulsory school (grundskolan). After students enter compulsory school in the year when they turn seven, they usually pass through three stages: grades 1-3 (low), grades 4-6 (middle), and grades 7-9 (high).⁷ At the beginning of each of these stages, students are assigned to classes, in which they remain for the duration of the stage.⁸ Importantly, students are not reshuffled into different classes after grade 7 (around age 13), in which they remain until the end of compulsory schooling.

Based on the institutional context, it is likely that the share of female peers is arbitrary: Parents are not allowed to influence class assignment of their children by legal means (Öquist and Wikström, 2006). It is also consistent with the mandate for educational facilities, and by implication for school principals, to provide equal access and uniform standards for students.⁹ A battery of balance and robustness checks detailed below corroborate these legal prescriptions, suggesting that gender composition in classrooms is indeed largely arbitrary.

Note that for the late stages of compulsory schooling we examine, teachers specialise in specific subjects in their educational training. They are normally assigned their teaching load based on the subjects they are qualified to teach rather than based on specific classrooms or cohorts.¹⁰ Accordingly, teachers generally do not select classrooms, but subjects which are then taught across classrooms.

In the last year of compulsory school, students can apply to proceed to a high school within their municipality. To do so, students need to attain a passing grade in a sufficient number of courses in compulsory school. Almost all students take this hurdle: Around 90%

⁷Around 6% to 12% of students in our sample attend charter schools (fristående skolor), which apply additional criteria for admission. Moreover, instead of the absolute distance, some municipalities apply a measure known as relative distance (relativ närhet). This metric involves comparing the relative distance between a school and the next best alternative across students. See, for example, the explanation by the schooling authority for [Stockholm](#).

⁸Note that schools have some discretion over which stages to offer and when to reshuffle. However, there is no reshuffling after grade 7 and students remain in the same school.

⁹Swedish Primary School Regulation (Grundskoleförfordning), SFS:1994:1194, 4 kap 4§.

¹⁰For more information, see the [report](#) by the Higher Education Authority (Högskolverket).

of students in our sample proceed to high-school. When entering high-school, students are naturally assigned to new classes.

2.2 Swedish Mental Health Care

Children and adolescents with mental health issues first have to go to the so-called first-line psychiatric care (första linjens psykiatrivård) and can then be referred to specialists. In most regions, local primary care (vårdcentral) is responsible for first-line care and thus the treatment of more common mental issues, such as mood disorders (Heurgren, 2019). Psychiatrists treat the vast share of these cases in outpatient consultation. Children aged 17 and below who need more attention are referred to a special branch known as child and adolescent psychiatry (barn- och ungdomspsykiatri, BUP), which provides inpatient and specialist outpatient treatment. Some of the wards also operate emergency rooms where both children and adults can be admitted in acute cases. Regular inpatient treatment occurs in open clinics, where patients usually pre-book a stay.

Our analysis relies on the nationwide register data covering all inpatient and outpatient mental health diagnoses of all Swedes from 2001 to 2012. The outpatient data covers all mental health diagnoses from the first-line psychiatric care and the child and adolescent psychiatry. Due to the standardized admission to care through outpatient referrals, the outpatient data covers over 92% of all mental health diagnoses.

2.3 Data and Descriptive Statistics

We use administrative data on classroom composition linked to all outpatient and inpatient mental health diagnoses for the years 2004 and 2012. We link the data using an anonymized version of the unique personal ID (personnummer) of each Swedish resident. The reason for this time horizon is that classroom composition is only available from 2004 onward, and diagnostic data only from 2001 to 2012. We further link the data on student classroom and mental health to information about students' family background. The final dataset covers 576,285 students from 26,278 classes in 855 schools over the course of 12 years.

We use data from administrative registers on schooling (National Exams Register, Nationella prov årskurs 9) to construct the main independent variable, the leave-one-out share of female peers in the classroom. The data comprises the population of Swedish compulsory-school students who took their national exams, and hence graduated, between 2004 and

2012. We know the class in which each student was. The classroom indicator comes from the group in which a student took the national exams. To ensure that it corresponds to their class, we cross-reference the indicator for each of the three exams in the register. For roughly 10% of students the class indicator is missing and we drop the corresponding observations.¹¹ We then use this information to compute the leave-one-out share of female peers for each class of all schools that can be observed throughout.¹² Note that, according to Skolverket (the Swedish school authority), approximately 25% of students do not change the classroom after entering school. This means students may well be exposed to the same peers from the beginning of compulsory school.

We use data from administrative registers on mental health diagnoses (Inpatient and Outpatient Register) to construct the main dependent variable, an indicator capturing whether a student was diagnosed with a mental health issue in the last three years of compulsory school.¹³ The last reshuffling of classrooms takes place in grade 7 which corresponds to ages 13 to 14 for most students. We identify mental health diagnoses using ICD-10 codes for each patient. The indicator is 100 if any of the ICD-10 codes between F00 and F99 is present and it is 0 otherwise.¹⁴ We also present results for specific diagnoses (see Table 3) and for the inpatient and outpatient data separately (see Table D.1).

Table A.1 presents the summary statistics of the data. The mean class size is around 23 students while there are, on average, around 5 classes per cohort. The mean share of female peers is 49%, with most of the observations between 40% to 60% share of females

¹¹We treat groups of students larger than 35 as having a missing class indicator. This class size is implausibly large in the Swedish context and suggest inaccurate class indicators. When including schools we do not observe over the whole sample period and all classes with more than 35 students, the estimates from the specification including school-by-cohort fixed effects and controls (the specification in column 7 of Table 1) are $\beta = 1.08$, $se = 0.29$, for all genders, $\beta = 1.78$, $se = 0.38$, for only boys, and $\beta = 0.37$, $se = 0.41$, for only females. The results are therefore equivalent to the results with the sample restrictions.

¹²Following Brenøe and Zölitz (2019), we exclude students whose class contained less than ten students (roughly 2% of the sample). These seemingly very small classes are likely coming from misreporting of classroom information and/or are because the corresponding students attend non-regular education, such as special education classes.

¹³This roughly corresponds to ages 13 to 16, depending on the birth month of the student.

¹⁴The 0 and 100 binary designation allows us to interpret the estimates as a percentage change. The ICD-10 codes indicating mental health issues contain 10 classifications: organic disorders (F00 - F09); mental and behavioural disorders due to psychoactive use (F10 - F19); schizophrenia, schizotypal and delusional disorders (F20 - F29); mood disorders (F30 - F39); neurotic, stress-related and somatoform disorders (F40 - F49); behavioural syndromes associated with physiological disturbances and physical factors (F50 - F59); disorders of adult personality and behaviour (F60 - F69); mental retardation (F70 - F79); disorders of psychological development (F80 - F89); behavioural and emotional disorders with onset in adolescence (F90 - F98).

(see Figure A.1). Regarding mental health, we observe that roughly 4.5% of students have a diagnosed mental health issue. The most prominent cluster of diagnoses with a prevalence of 1.5% are hyperactivity and conduct disorders including disorders of social functioning. For parents, mood-related diagnoses, covering such illnesses as depression, are most prevalent with 4.2% of parents having such a diagnosis.¹⁵

3 Empirical Strategy and Plausibility Checks

3.1 Specification and Identifying Assumptions

We estimate the following main specification separately for boys and for girls:

$$Y_{ic} = \beta_1 \times ShareFemPeers_{ic} + \alpha_{school \times cohort} + X_i \gamma' + \epsilon_{ic}$$

In the equation above, Y_{ic} is an indicator for a mental health diagnosis for student i in classroom c . The key explanatory variable is $ShareFemPeers_{ic}$, the proportion of female students in a given classroom. It is defined as $\frac{ngirls_c - gender_i}{size_c - 1}$, where $ngirls_c$ is the number of girls in a given classroom, $size_c$ is the classroom size, and $gender_i$ is the student's gender (1 = female). The estimate for β_1 gives the effect of gender peer composition. $\alpha_{school \times cohort}$ represent one fixed effect for each school and cohort interaction. The specification therefore exploits within school-cohort variation coming from differences in gender composition across classrooms. X_i denotes individual and school-level controls. The individual controls are log family income, dummies for parental education (indicators for high-school degree, vocational degree, college degree, stem degree), and an indicator for any parental mental health diagnosis before school start.¹⁶ The included school-level controls are classroom size, cohort size, and the number of schools in a given municipality. When we show results for the pooled sample of boys and girls, we interact all fixed effects and controls with gender. We cluster the standard errors on the school level, allowing for correlation in classroom makeup within school.

¹⁵This is in line with depression being the most frequently reported mental health diagnosis, please see the following WHO [factsheet](#).

¹⁶We only observe parental mental health before school start for cohorts finishing school after 2009. Parents of students of earlier cohorts receive an indicator for missing mental health data. When using parental mental health as a control variable, we include a dummy for each value of the categorical variable capturing no mental health diagnosis, any mental health diagnosis, or a missing mental health diagnosis.

The main threats for a causal interpretation of the estimates from the above specification come from potential sorting of students. Here, the institutional context ensures that there is little selection apart from geographic location and age into schools and cohorts. The empirical strategy addresses this selection and even school-cohort specific selection since it exploits idiosyncratic variation in classroom composition conditional on *school*×*cohort* fixed effects. Based on theory (Lazear, 2001) and recent evidence (Ammermüller and Pischke, 2009; Chetty et al., 2011; Burke and Sass, 2013; Balestra, Eugster and Liebert, 2020; Bietenbeck, 2020), we expect the classroom environment to be the relevant peer environment. Previous literature has pointed out that peer composition across cohorts, conditional on including one set of school and a separate set of cohort fixed effects, is already as good as randomly assigned (Hoxby, 2000; Black, Devereux and Salvanes, 2013; Helene et al., 2015). Many peer effects papers therefore exploit the within-school, across-cohort variation in peer composition.¹⁷ We show that the results are qualitatively equivalent when we use these less restrictive specifications (see Table D.2). In the main specification, we choose the most restrictive specification: We take dynamic selection into account by controlling for *school*×*cohort* fixed effects, which fully accounts for potential changes in sorting to schools over time.

Given the above specification, the main identifying assumption is that no omitted variable satisfies the following conditions: (i) time-variant and class-specific, (ii) not captured by cohort×school fixed effects, (iii) associated with mental health diagnoses and gender peer composition, (iv) not captured by a large set of individual controls based on administrative data. The existence of such a factor seems highly unlikely given that parents are not allowed to influence the choice of class by legal means (Öquist and Wikström, 2006). Still, to assess the likelihood of such a factor existing, we examine the relationship between high-quality and detailed observable characteristics from administrative registers and gender peer composition in the classroom. We also examine whether student mental health before school predicts the share of female peers. Taken together, the results from these and further checks which we detail below support the main identifying assumption.

¹⁷To account for potential dynamic selection across cohorts into schools, some recent contributions have included school-specific time trends to analyze deviations from peer composition conditional on dynamic trends (Brenøe and Zölitz, 2019).

3.2 Balance and Placebo Checks

The school \times cohort fixed effects address static and potential dynamic selection at the school-cohort level. We now check whether remaining variation in the classroom gender peer share is likely arbitrary in our sample. Here is a summary of what we find:

1. Across 154 bivariate regressions, we find no indication that family characteristics, including parental mental health, income, and education systematically predict gender peer share.
2. Student mental health before school start does not predict the share of female peers.
3. The residualized share of females is normally distributed.
4. The distribution of the residualized share of females is similar to the simulated residualized share of females where we randomly assign students to classes.
5. Student gender does not predict the share of female peers.
6. Classroom fixed effects do not jointly predict student gender.
7. Including classroom fixed effects does not markedly affect the coefficient size of estimated gender differences.
8. Gender peer share in previous and following classes does not affect mental health.
9. There is no increase in mental illnesses which are almost exclusively biologically driven.

Balance Checks:

Family Characteristics. We provide a series of balance checks for high-quality background variables on parental and family characteristics from administrative registers. The 22 variables include, among others, family income, parental education, and parental mental health diagnoses. For each variable, we examine whether there is a correlation with the share of female peers in the classroom across 7 specifications. If there were many such correlations, this would indicate that parents may be able to select to specific classrooms within school-cohorts. We show the results of the resulting 154 bivariate regressions in Table B.1. In the absence of systematic sorting, one would expect to have approximately 10%, 5%, and 1% of the coefficients to be significant at each corresponding level. Out of 22 variables, there are only 2 statistically significant coefficient estimates at the 10% level in the most restric-

tive specification. We conclude that the fraction of statistically significant coefficients is not higher than could be expected by chance.

Prior Mental Health. We further check whether students with mental health diagnoses *before* starting school are more likely to end up in classes with a higher share of females. The data offers a unique possibility for such a check. Usually, prior measurement of later outcome variables are not observable in peer effects studies before students are exposed to peers. To examine whether there is selection to classrooms based on pre-existing mental health issues, we regress the indicator capturing a diagnosis before school starts on the share of later female peers.¹⁸ Note that this check also directly tests for the potential role of pre-existing differences in mental health of boys and girls.¹⁹

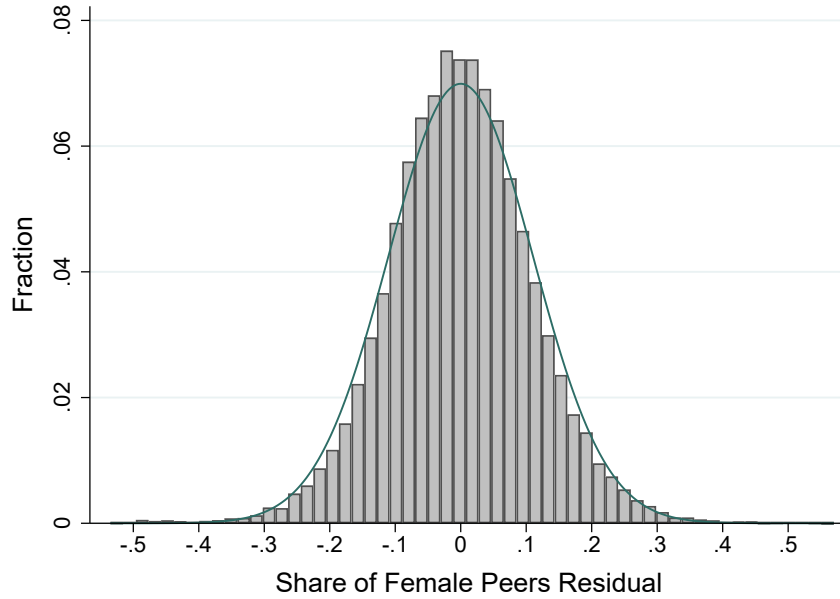
We do not find any indication that students with worse mental health before starting school are more likely to end up in classes with a higher share of female peers, see Table B.2. None of the coefficients are statistically significant, and they are a fraction of the magnitude of the effect estimates of the impact of the share of female peers on later mental health. The coefficient signs, if anything, indicate better mental health among students ending up in classes with a high share of female peers.

Distribution and Simulated Distribution of the Share of Females. In a next plausibility check of as-good-as-random assignment to classrooms, we examine the variation of the gender peer share that we exploit. If gender peer share was as good as randomly assigned, we would expect that the corresponding distribution of peer shares would look normally distributed conditional on school \times cohort fixed effects. Figure 1 suggests that the corresponding distribution is well-behaved and follows a normal distribution, suggesting that gender peer share may indeed be arbitrary.

¹⁸We examine diagnoses received prior to entering school rather than diagnoses before the last stage of compulsory school because we do not have information the share of female peers in earlier stages. Therefore, we do not know with certainty the classroom composition before the last stage of compulsory school. According to the Skolverket (the Swedish school authority), approximately 25% of students do not change the classroom after entering school. A substantial share of students is thus treated from when they enter school, which is why we examine diagnoses before school start.

¹⁹If pre-existing differences played a role, it would likely bias the results in the opposite direction that we observe. Females have more mental health issues, which indicates that we should see a higher incidence of mental health issues among females in classes with more females. However, we observe the opposite.

Figure 1: Variation of the Share of Female Peers Across Classrooms Within School-Cohorts



Note: The figure shows the residuals from OLS regressions of the share of female peers on school-by-cohort fixed effects. The overlaid curve shows a normal distribution with the same mean and variance.

We further test whether peer-gender variation within schools is consistent with random assignment by comparing the distribution of the share of female peers to a simulated distribution of the share of female peers based on randomly assigning students to classrooms. To this end, we do Monte Carlo simulations in which we assign students randomly to classes within their school-cohorts. We take the number and size of classes from the actual data. In the spirit of Bietenbeck (2020), we then regress the share of female peers on school \times cohort fixed effects in the simulated data and collect the residuals. We plot the residuals from the simulations alongside the residuals from the actual data in Figure B.1. The distributions look very similar, a result consistent with as-good-as-random assignment of the share of female peers.

Gender and the Share of Female Peers. In addition, we also examine whether a student's own gender correlates with the proportion of females in their class following the methodology proposed by Guryan, Kroft and Notowidigdo (2009). To examine the correlations we use seven specifications analogous to the main results table (Table 1). Across specifications, we control for the school \times cohort-level leave-one-out class mean of the share of females to account for the mechanical relationship between peer and own gender following Guryan, Kroft and Notowidigdo (2009). That is, we control for the share of female peers in the other classes of a

student’s school–cohort. There is no statistically significant correlation between own gender and the share of female peers, see Table B.3.

Gender and Classrooms. Following Chetty et al. (2011); Balestra, Eugster and Liebert (2020), we regress student gender on class fixed effects. The class fixed effects should be jointly insignificant if assignment to classroom is random regarding student gender (Chetty et al., 2011), which is what we find. We proceed as follows: In the first step, we regress student gender on the school×cohort fixed effects and controls as indicated and we then retrieve the residuals from this regression. In the second step, we regress the residuals obtained in the prior regression on classroom fixed effects. We then do a joint F -test to determine whether those fixed effects are jointly significant. Across three different specifications the F -statistics suggest no predictive power of classroom fixed effects for student gender: $F = 0.727$ without controls, $F = 0.724$ with school-level controls, and $F = 0.722$ with school-level and individual-level controls. All F -statistics are well below critical values required for statistical significance.

Stability of Gender Differences to Classroom Fixed Effects. We conduct a final check, where we include classroom fixed effects to better understand the potential role of selection into classrooms. Note that this check does not directly assess the robustness of the main effects, which show how the share of female peers affects the level of mental health in males and females. It is not possible to estimate these levels when we include classroom fixed effects. However, it is possible to estimate how the share of females shapes the difference between males and females conditional on classroom fixed effects.

Examining the stability of the gender difference to the inclusion of classroom fixed effects is informative: If including classroom fixed effects changes the estimated effect of the gender peer share on the gender difference in mental health, this may indicate potential selection. However, if the estimated impact on gender difference remains similar when we include classroom fixed effects, it is unlikely that selection into classrooms has large effects on levels of mental health in males and females. The argument is that if we observe stable gender differences across specifications, it is unlikely that selection into classrooms is a key driver of the main results. Table B.4 in the Appendix shows that gender differences are stable

even when including classroom fixed effects, which provides another piece of evidence that classroom selection does likely not drive the results.

Placebo Checks:

Previous and Subsequent Cohorts. We also do the following placebo checks to further assess the plausibility of the identification strategy. First, we examine whether the share of female peers in the previous or the subsequent cohort affects mental health diagnoses. Table B.5 shows that the corresponding regression coefficients are not significantly different from zero. The check indicates that any effect of current class or cohort composition picks up idiosyncratic variation coming from the current cohort rather than from the previous or past cohort.

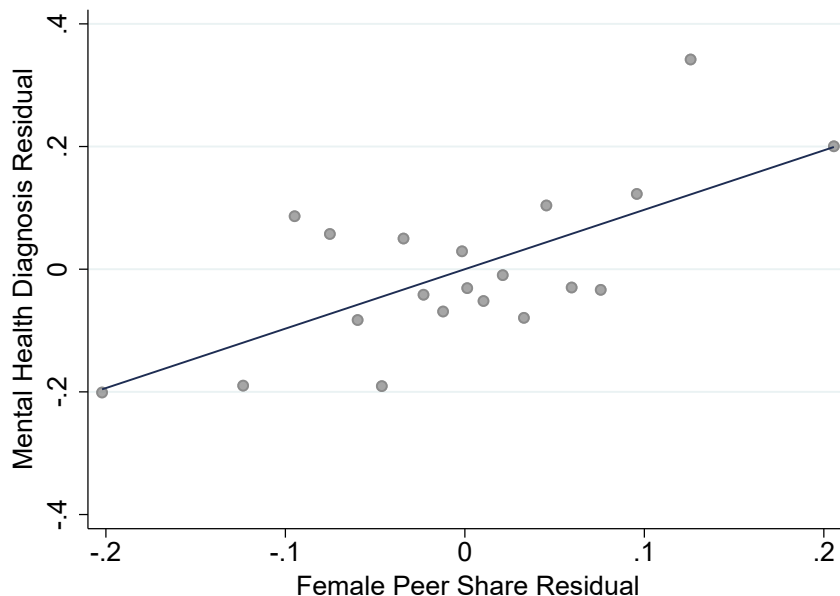
Diagnoses with Biological Origins. We next examine mental health issues that are very unlikely to be influenced by peer relationships: organic disorders and mental retardation. Both have biological origins often determined at birth (Wittchen, 2001; Costeff, Cohen and Weller, 2008). We present the estimates for these subcategories in Table B.6. There are no statistically significant effects of peer composition on mental health diagnoses mainly determined by biology.

4 Peer Gender and Mental Health

Figure 2 and Table 1 show the main estimates for the impact of the gender peer composition on mental health. Table 1, row 1, shows the effect estimates for the pooled sample, row 2 shows the estimates for only males, and row 3 the estimates for only females. Across different specifications, the results indicate that a higher share of female peers leads to a higher prevalence of mental health diagnoses. As the results in row 2 suggest, boys drive this effect (see Figure C.1 for the scatter plots showing the results for only males and only females).

The effects are precisely estimated overall, and for boys, and are substantial in size. The coefficient estimates for all genders indicates a 1.06 ($se = 0.35$) percentage point increase in mental health diagnoses when changing from an all male to an all female classroom. Accordingly, the likelihood of mental health diagnosis increases by 2.2% with a 10 percentage point

Figure 2: Gender Composition and Mental Health



Note: The figure shows the relationships between the residual of mental health diagnoses and residual of the share of female peers in the classroom ($N = 576,285$). The residuals stem from regressions of the respective variables on cohort-by-school fixed effects. The dots show the binned averages across 20 quantiles. The line shows the linear fit from OLS regressions using all data with $\beta = 0.97$, $se = 0.35$. The figure shows the relationship for the entire population, whereas it is also estimated separately for boys and girls in Table 1.

increase in the share of female peers ($sd = 0.09$) when compared to the average prevalence of mental health issues of 4.7%.²⁰ For adding just one female student to an averaged sized classroom, this corresponds to a 1.1% increase in mental health diagnoses. When adding 5 female students, the effect size translates into a 5.5% increase. For boys, a 10 percentage point increase in the share of female peers corresponds to a 4.5% increase in the likelihood of receiving a mental health diagnosis. Note also that the differences between boys and girls are statistically significant (see Table B.4).

These are large increases when compared to the transmission of mental health between parent and children: Having one parent with a mental health diagnosis increases the likelihood of receiving a mental health diagnosis at this age by roughly 4 percentage points ($se = 0.001$). Therefore, the effect of a 10 percentage point increase in the share of female peers on mental health diagnosis is 2.7% as large as the relationship between parental and child mental health. The effect is also large when compared to the link between parental education and child mental health: The effect corresponds to roughly 15% of the link between parental education

²⁰When not interacting fixed effects with gender, the magnitude of the effect for both genders in specification (7) of Table 1 is around 1.8% ($\beta = 0.79$, $se = 0.34$).

Table 1: Gender Composition and Mental Health

	Mental Health Diagnosis [0,100]						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
All Genders (Avg: 4.67)	1.04*** (0.30)	0.96*** (0.30)	1.05*** (0.29)	1.02*** (0.29)	1.10*** (0.29)	0.97*** (0.35)	1.06*** (0.35)
Males (Avg: 4.21)	1.47*** (0.38)	1.59*** (0.38)	1.69*** (0.37)	1.70*** (0.38)	1.77*** (0.37)	1.82*** (0.46)	1.92*** (0.45)
Females (Avg: 5.16)	0.41 (0.40)	0.31 (0.41)	0.38 (0.40)	0.34 (0.41)	0.40 (0.41)	0.08 (0.48)	0.17 (0.48)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	576,285	576,285	576,285	576,285	576,285	576,285	576,285
Schools	855	855	855	855	855	855	855
<i>R</i> -squared	0.00	0.01	0.01	0.02	0.02	0.04	0.04

Note: The table shows the estimated relationship between the incidence of mental health diagnoses and the share of female peers in the classroom. The first row shows the results for the sample with both genders (All Genders). In this specification, we interact all fixed effects and controls with gender. The next two rows show coefficient estimates from separate estimations based on samples including either only males or only females. The number of observations and the corresponding *R*-squared come from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

and child mental health. Taken together, the coefficient estimates suggest an important role of peer gender composition in shaping mental health.

The coefficient estimates are stable across different specifications. Column (1) shows the results of OLS regressions without any adjustments or controls. Adding school and cohort fixed effects in columns (2) and (3) yields similar results. To partly address potential selection into school-cohorts across time, column (4) includes school specific linear time trends. Column (5) then adds individual- and school-level controls. Both additions do not decrease the size of the coefficient estimates. Columns (6) and (7) present our main specifications including school \times cohort fixed effects in column (6) as well as controls in column (7). The last two specifications address potential selection across time to school-cohorts. Again, the coefficient estimates are similar to and statistically indistinguishable from the coefficient estimates in less restrictive specifications. The coefficient stability to the inclusion of fixed effects and controls suggests any potential bias because of remaining unobservables is likely small (Oster, 2019).

4.1 Robustness Checks and Further Results

Inpatient and Outpatient Diagnoses. The outpatient registry covers the vast majority of diagnoses (approximately 92%) . This makes sense as most patients with mental health issues enter the system through outpatient referral or get outpatient care after an inpatient stay. In the main specification, we examine the effects on mental health diagnoses using data from the outpatient and the inpatient registers. However, we also separately examine the effect of the share of female peers on inpatient and outpatient diagnoses.

Table D.1 in the Appendix shows the results for the combined inpatient and outpatient data in row 1 (corresponding to the main results), only outpatient data in row 2, as well as only inpatient data in row 3. The results show that outpatient diagnoses are the key driver of the results, including inpatient diagnoses does not strongly affect the results. The coefficient estimates are also positive and large across the board for inpatient diagnoses but less precisely estimated.

Exploiting Within-School Across-Cohort Variation. In additional specifications, we exploit only cohort variation rather than classroom variation. We show the results exploiting within-school, across-cohort variation in Table D.2. The overall estimates go in the same direction as the main estimates and are statistically significant. Because of more imprecision in measuring the share of female peers, however, the results are less precisely estimated. Again, as before, the effects for girls are relatively smaller than the effects for boys. Taken together, the additional results confirm that peers affect mental health.

Movers. Individuals who stay in the same neighbourhood do not have an institutional reason to change schools. However, could people who move from one municipality to another drive the result? To examine this concern, we split the sample into those who move across municipalities and those who do not. We define non-movers as students who lived in the same municipality throughout the entire nine-year period of compulsory schooling as well as the year before the start of compulsory school. Just under 80% of students do not move across municipalities and therefore likely remain in the same school. The coefficient estimates across the groups do not differ substantially (see Table D.3). This suggests that the results are likely not driven by students changing schools.

Definition of Mental Health Diagnosis. The main outcome variable captures any mental health diagnosis of a student during the last three years of compulsory schooling. In this check, we examine whether the results are similar when we only examine diagnostic codes observed in the last year of school (see Table D.3). We do not observe significant changes in the direction and the magnitude of the coefficients relative to the the alternative definition of a diagnosis. This check suggests that the results are not sensitive to how we define the cut-off for a diagnosis during compulsory school.²¹

Outliers. The results might be driven primarily by the tails of the distribution. To examine the possibility of outliers driving the results, we estimate the main specification without the top and bottom 5% of classrooms with the highest share of female peers. In our data, these cut-offs roughly correspond classrooms having a two-to-one gender ratio. Dropping extreme observations results in qualitatively equivalent coefficient estimates (see Table D.3).

5 Persistent Effects on Mental Health

Do the effects on mental health persist? We examine whether peer environment affects student mental health also in early adulthood. We follow students up to 13 years after entering primary school or age 19.²²

Table 2 shows the effects of compulsory-school gender composition on the likelihood of a mental health diagnosis within three years after completing compulsory schooling. The specifications therefore test the persistence of effects even after students switched to a different class in high-school, or to a job. Similarly to the main results, we document robust and large effects of gender composition on longer run mental health. Boys drive the effect, and, as before, we observe positive but less precisely estimated coefficients for girls.

The estimate in column (7) indicates that a 10 percentage point increase in the share of girls in compulsory school results in an approximately 2.6% increase in the likelihood of a

²¹It also suggests that intermittent changes in classroom composition do not play a large role if they occur.

²²This corresponds to the last year of high-school. We follow both students who transition to a different high-school class and students who leave school after compulsory school for up to three years after graduation. In the sample, over 90% of students transition to high-school. Because we examine how students fare up to 3 years after they left compulsory school, we only have observations for students completing compulsory school before 2010. The reason for the sample reduction is that students who completed compulsory-school after 2010 are not observable for the whole duration of high school. We show the compulsory-school or more short-run effects for this sample in Table D.4.

Table 2: Gender Composition in Compulsory School and Later Mental Health

	Mental Health Diagnosis [0,100]						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
All Genders (Avg: 7.72)	1.45*** (0.46)	1.36*** (0.45)	1.34*** (0.44)	1.43*** (0.46)	1.55*** (0.45)	1.78*** (0.52)	1.99*** (0.51)
Males (Avg: 6.4)	1.68*** (0.54)	2.03*** (0.53)	2.02*** (0.52)	2.15*** (0.54)	2.21*** (0.54)	2.41*** (0.62)	2.58*** (0.61)
Females (Avg: 9.1)	0.68 (0.65)	0.68 (0.66)	0.64 (0.65)	0.78 (0.67)	0.93 (0.67)	1.14 (0.76)	1.36* (0.75)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	414,716	414,716	414,716	414,716	414,716	414,716	414,716
Schools	890	890	890	890	890	890	890
<i>R</i> -squared	0.01	0.02	0.02	0.02	0.02	0.04	0.04

Note: The table shows the estimated relationship between the incidence of mental health diagnoses up to 3 years after the end of compulsory schooling (or approximately age 19, depending on the birth month) and the share of female peers in classroom in compulsory schooling. The number of observations is lower here since we need to observe students for three more years. Therefore, we only have observations for students completing compulsory school before 2010. The first row shows the results for the sample with both genders (All Genders). In this specification, we interact all fixed effects and controls with gender. The next two rows show coefficient estimates from separate estimations based on samples including either only males or only females. The number of observations and the corresponding *R*-squared come from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

later mental health issue when compared to the incidence of mental health issues at this age of 7.7%. For boys, a 10 percentage point increase in the share of female peers corresponds to a 4% increase in the likelihood of a diagnosis. These relative increases are in the ballpark of the relative increases of mental health issues during compulsory school. The coefficient estimates are larger in absolute size, which could be because of the higher baseline prevalence or because peer influences take time to manifest. For instance, problems may emerge during compulsory school but may only be diagnosed later on. Taken together, the results show persistent and sizable impacts of gender peer composition on mental health.

6 Mechanisms

6.1 Links to Peer Relationships Based on Diagnosis Type and Self-harm

Diagnosis Type. We examine whether peer relationships are likely responsible for our findings by examining specific diagnoses. Table 3 shows which of 8 diagnostic types drive the main results. We use the main specification including school \times cohort fixed effects as well as controls (specification 7 from Table 1) for the estimations.

The overall pattern of results for boys matches priors based on psychological research: Peer environment mainly affects diagnostic categories intimately linked to peer relationships (Parker et al., 2015; Ackermann et al., 2019). Learning and socialization disorders, as well as hyperactivity and conduct disorders, are most affected.²³ The hyperactivity and conduct disorders category includes diagnoses such as disturbances of attention or conduct disorders such as disorders of social functioning, and emotional disorders with onset specific to adolescence. Learning and socialization disorders include problems such as dyslexia. Boys experience a statistically significant increase in both categories, suggesting that the effects on these two categories drive the impact of classroom composition on mental health.

A 10 percentage point increase in the share of female peers increases the incidence of hyperactivity and conduct disorders in boys by 0.12 percentage points. Compared to the average prevalence of 2%, this corresponds to a 6% relative increase of hyperactivity and conduct disorders for boys.²⁴ The relative increase for conduct disorders excluding hyperactivity related disorders is 6.5% ($p < 0.01$). With regards to learning and socialization disorders, the corresponding relative increase is 6%. Again, these are large effects on mental health.

In addition, we see some indication that depression and mood disorders are more common with a higher share of female peers as well as that diagnoses related to substance abuse are more common among boys. Thus, it seems that substance abuse related diagnoses also drive the effect for boys. For girls, the positive coefficients on depression and mood disorders,

²³We exclude two categories of diagnoses that should not be affected by the environment in adolescence. First, we exclude mental retardation since mental retardation is largely biologically driven. Second, we exclude organic mental disorders (F00 - F09) which are also mostly driven by biology. We show that these categories are not affected in Table B.6.

²⁴Table A.1 shows the average incidence across categories.

neurotic and stress disorders, learning and socialization disorders, as well as schizotypal disorders, seem to provide an explanation for an imprecisely estimated positive impact on overall mental health diagnoses.

Taken together, an increase in diagnoses of hyperactivity and conduct disorders as well as learning and socialization disorders among boys drives the impact on overall mental health. For girls, there is no glaring heterogeneity across subcategories, which is consistent with the main results and suggests that gender environment has little impact on girls.

Table 3: Gender Composition and Diagnosis Type

Diagnosis:	All	Males	Females
	(1)	(2)	(3)
Hyperactivity and Conduct Disorder	0.59*** (0.19)	1.21*** (0.30)	-0.05 (0.22)
Substance-Related Disorder	0.21 (0.16)	0.53*** (0.21)	-0.14 (0.24)
Depression and Mood Disorder	0.26** (0.13)	0.21 (0.15)	0.33 (0.23)
Eating and Sleeping Disorder	-0.02 (0.11)	0.06 (0.07)	-0.10 (0.21)
Neurotic and Stress Disorder	0.13 (0.15)	0.07 (0.17)	0.20 (0.25)
Learning and Socialization Disorder	0.17* (0.10)	0.33** (0.16)	0.01 (0.11)
Schizotypal Disorders	0.05* (0.03)	0.01 (0.03)	0.09** (0.04)
Adult Personality Disorder	0.03 (0.03)	-0.01 (0.03)	0.07 (0.05)

Note: The table shows the estimated relationships between specific diagnosis types and the share of female peers in the classroom conditional on school-by-cohort fixed effects and controls. Column (1) shows the results for the sample with both genders (All). In this specification, we interact all fixed effects with gender. The next two columns show coefficient estimates from separate estimations based on samples including either only males or only females. The dependent variable in each row refers to a specific diagnostic category of ICD-10 codes F00 - F99. The names of some of the categories in the table have been adjusted for easier reference of non-psychological/psychiatric readership. Hyperactivity and Conduct Disorders corresponds to the WHO category Behavioural and emotional disorders with onset usually occurring in childhood and adolescence (F90 - F98); Substance-Related disorders to Mental and behavioural disorders due to psychoactive substance use (F10 - F19); Eating and Sleeping Disorders to Behavioural syndromes associated with physiological disturbances and physical factors (F50 - F59); Learning and Socialization Disorders to Disorders of psychological development (F80 - F89). * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Self-harm. In addition to typical diagnostic codes capturing mental health problems, we estimate the effect of the share of female peers on the incidence of self-harm. The incidence of intentional self-harm is much lower than the incidence of mental health issues, so the results have to be interpreted with caution.²⁵ We consider this check useful to examine whether the higher incidence of mental health diagnoses also reflects actions related to self-harm. The results in Table E.1 suggest that not only the diagnostic incidence of mental health problems increases among boys, but also actions related to self-harm.

The impact of the gender composition on self-harm in boys is positive, relatively large, and statistically significant across the board. A 10 percentage point increase in the share of female girls leads to a more than 8% increase in the incidence of intentional self-harm. The magnitude of the effects is stable across specifications. The point estimates for girls are negative, but not statistically significant and decrease with more restrictive specifications. The results are broadly consistent with the mental health findings that show statistically significant effects of female peers on boys and imprecisely estimated effects on girls.

6.2 Friendship Constraints and Educational Attainment

Friendship Constraints. A straightforward explanation for the impact on diagnoses linked to peer relationships is friendship formation. Having fewer same-gender peers in the classroom limits the number of potential friends given that same-gender friendships are considerably more common than opposite-gender friendships in adolescence (see, e.g., Hill, 2015). Naturally, having friends is critical for well-being (see, e.g., Parker and Asher, 1987; Bond et al., 2007; Ho, 2016). Large friendship networks are particularly important for boys (Benenson, Apostoleris and Parnass, 1997; Rose and Rudolph, 2006). So if a class has fewer boys, it might be harder for boys to have the number of social contacts that they desire. The lack of a large social network for boys could be one reason for lower mental health.

Educational Attainment, Classroom Ability, and Classroom Mental Health. One further possible mechanism could be that peer gender composition affects compulsory-school grades, which, in turn, affect mental health. We see that boys perform worse in classes with

²⁵To examine the effects on intentional self-harm we use the ICD-10 codes X60 to X84. Naturally, there is a statistically significant relationship between mental health diagnoses and intentional self-harm ($\beta = 0.45$, $se = 0.02$).

more girls and girls perform better (see Table E.2). For boys, the coefficient is -0.25 and the standard error 0.03. Also, boys are much more affected than girls. The results suggest that educational attainment impacts may be partly responsible for the mental health effects.

Doing naive accounting of the variation by simply controlling for compulsory grades reduces the effects of peer composition on mental health by up to 50% (see Table E.3). Thus, there seems to be an important link between the two, but the link between mental health and educational attainment does not fully explain the link between gender peer composition and mental health.

We also test the possibility that average ability approximated by grades in the classroom explains the effects in Table E.4. Controlling for average grades in the rest of the classroom does not explain the effects. Neither does controlling for average mental health in the classroom. Taken together, key peer characteristics are not responsible for the results.

6.3 Bullying

A key reason for why mental health diagnoses related to peer relationships could increase is bullying. Bullying has large negative consequences for mental health (Klomek, Sourander and Elonheimo, 2015; Rees, Sabia and Kumpas, 2020), and it reduces educational attainment (Ammermüller, 2012; Eriksen, Nielsen and Simonsen, 2014). Differential levels of physical and psychological bullying in classes with more females could therefore provide an explanation for the results.

Physical Bullying. Using survey data, Lavy and Schlosser (2011) find that a higher proportion of girls in the classroom relates to less fighting. Lee et al. (2014) and Giardili (2020) also observe that single-sex schools are associated with fewer self-reported bullying incidents for both sexes. The previously documented impact cannot explain our results. The reason is that if bullying decreases in classrooms with more girls, we would expect a *decrease* rather than an *increase* in mental health problems.

We examine the role of bullying in our setting. To this end, we study the impact on physical assaults and injuries. Table E.5 shows the corresponding results.²⁶ We do not find a statistically significant impact on assault. This may be because of a very low incidence

²⁶Roughly 0.22% of students in our sample experience physical assault and roughly 22% any kind of physical injury.

of assault in the data. In contrast, we estimate a statistically significant and substantial reduction of physical injuries for boys and girls. The latter result is in line with previous evidence: More girls in the class seem to, if anything, reduce physical harm. This reduction in physical harm is at odds with the documented mental health effects.

Psychological Bullying. A large literature suggest that boys and girls engage in different types of aggression (Lagerspetz, Björkqvist and Peltonen, 1988; Björkqvist, Lagerspetz and Kaukiainen, 1992; Crick and Grotpeter, 1995; Card et al., 2008; Smith, Rose and Schwartz-Mette, 2010). A key finding of the literature is that conventional measurements of aggression tend to overlook indirect forms of aggression usually aimed at undermining reputation.

The indirect form of aggression, also called relational aggression, seems to be more commonly used by girls (Lagerspetz, Björkqvist and Peltonen, 1988; Björkqvist, Lagerspetz and Kaukiainen, 1992; Crick and Grotpeter, 1995; Card et al., 2008; Smith, Rose and Schwartz-Mette, 2010). Like direct aggression, relational aggression relates to a worse school environment (Goldstein, Young and Boyd, 2008) and peer difficulties (Cillessen and Mayeux, 2004). It also relates to higher anxiety levels (Loukas, Paulos and Robinson, 2005), loneliness and depression (Crick and Grotpeter, 1995), increased drug use (Sullivan, Farrell and Klierer, 2006), and has intimate links to mental health (Ackermann et al., 2019). Intriguingly, boys seem to be more affected by psychological bullying (Crick and Grotpeter, 1996; Goldstein, Young and Boyd, 2008; Williams et al., 2009).

Previous literature thus suggests that more girls in the class increases the incidence of psychological bullying with likely negative consequences for mental health, particularly among boys. Psychological bullying could therefore explain part of the documented relationship between peers and mental health.

6.4 Increased Discovery Rates

One concern could be that gender composition affects the discovery of more mental health problems, but does not affect their actual incidence. For instance, it may be that girls are more observant and therefore mental health problems surface quicker. However, an overall look at the results suggests that this is an unlikely channel. First, only certain diagnosis types—particularly the ones with an intimate link to peer relationships—are affected. Second, the impact on self-harm among boys suggests an actual increase in mental health problems.

Third, girls and boys are differentially affected. If discovery rates are higher with more girls, the incidence in diagnoses among girls should also increase.

Fourth, if discovery rates were higher, the coefficient estimates should be negative for inpatient diagnoses as problems are discovered earlier in classes with more females. In turn, the overall estimated impact using inpatient and outpatient diagnoses jointly should be much lower than when just using outpatient diagnoses. However, the point estimates for inpatient diagnoses are positive and the overall estimated impact is similar comparing the results for only outpatient data and for inpatient and outpatient data combined. Fifth, we do not observe heterogeneity depending on the socioeconomic status of the parents (see Section 6.5). If reporting was an issue, it seems likely that the impact on mental health diagnoses would vary by parental characteristics. For instance, if there is under-reporting among children from low socioeconomic status backgrounds, there should be a relative increase in mental health diagnosis when there are more girls in the class. In sum, the results indicate an actual increase of mental health issues rather than just a higher discovery rate.

6.5 Further Mechanisms

The Role of Parents. We discuss further mechanisms which could contribute to the observed effects. Table E.6 shows heterogeneity in effects for students by parental characteristics. We estimate a large level difference relative to the baseline effect for students with a parent with a mental health diagnosis when compared to those without. Having at least one parent with a mental health diagnosis substantially amplifies the effect of having a lot of female peers in the classroom, but the coefficient estimate is not statistically significant. The coefficient estimates for parental education background and income level are small and statistically imprecisely estimated. Accordingly, parents do not seem to play an important role in how children’s mental health reacts to classroom composition.

Teacher–Student Relationships. There are other environment-related mechanisms that could be responsible for the results. For instance, more females in the cohort relate to better teacher-student relationships (Lavy and Schlosser, 2011; Giardili, 2020).²⁷ However, previous

²⁷Note that selection of teachers on school-cohort level should not affect our estimates as we consider within-school variation.

results do not correspond to the mental health effects we observe. Based on these results, we would expect that more girls improve mental health among girls and potentially boys.

7 Conclusion

Previous research has demonstrated the importance of early mental health for adult mental health and economic well-being (Currie and Stabile, 2006; Currie et al., 2010; Salm and Schunk, 2012; De Neve and Oswald, 2012; Lundborg, Nilsson and Rooth, 2014). But what drives adolescent mental health? We show that the classroom composition—as captured by the gender composition—has a large and persistent effect on mental health.

To this end, we use a unique data set which combines classroom gender composition and mental health diagnoses. We find that a higher share of female peers increases the likelihood of mental health problems, particularly among boys. These effects are large and persist into adulthood. The results highlight that peer relationships are crucial in shaping well-being.

We find that the mental health diagnoses which are affected are the ones with close ties to peer relationships according to psychological research. The likelihood of diagnoses related to conduct disorders as well as learning and socialization disorders increase. We further examine potential reasons for peer relationship by discussing friendship constraints, educational attainment, and bullying. We conclude that boys having stronger constraints on friendship networks, lower educational attainment of boys, and a higher incidence of psychological bullying in classrooms with more females might offer explanations for the decrease in mental health.

The results indicate that compulsory-school environment has large causal impacts on mental health and therefore reinforce calls for early life policy interventions. It is likely that interventions geared at improving mental health early on during compulsory school could have disproportionate positive impacts on later life well-being. Based on our results, one intervention could be focusing more strongly on achieving classroom balance in gender composition.

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Peer Gender and Mental Health

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Appendix

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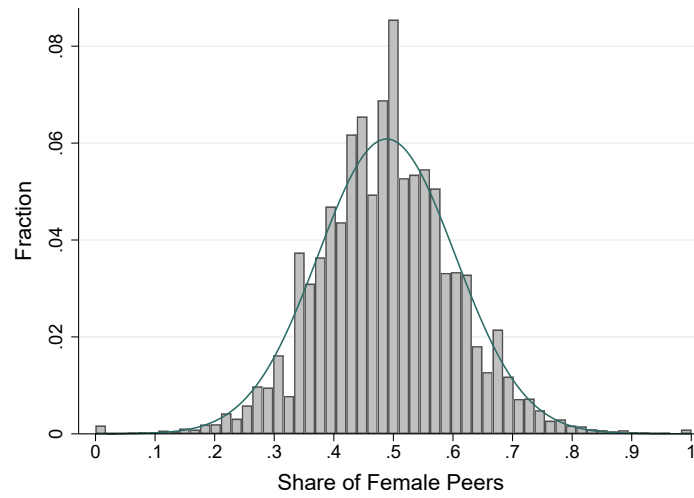
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A Summary Statistics

Figure A.1: Variation of the Share of Female Peers Across Classrooms



Note: The figure shows the distribution of the share of female peers across classrooms. The overlaid curve shows a normal distribution with the same mean and variance.

Table A.1: Summary Statistics

Variable	N	Mean	SD	N	Mean	SD
Background Variables						
		Men		Women		
Immigrant	294,602	0.07	0.26	281,684	0.07	0.26
2nd-Generation Immigrant	294,602	0.30	0.46	281,684	0.31	0.46
Number Siblings	294,602	1.45	1.15	281,684	1.45	1.17
Adopted	294,602	0.01	0.08	281,684	0.01	0.09
Parent Unknown	294,602	0.08	0.27	281,684	0.08	0.27
Teenage Parent	294,602	0.02	0.13	281,684	0.02	0.14
Single Parent	294,602	0.07	0.25	281,684	0.07	0.25
Log Family Income	262,857	12.44	2.78	250,912	12.44	2.77
Parents High School	294,602	0.70	0.46	281,684	0.70	0.46
Parent Vocational Degree	294,602	0.27	0.44	281,684	0.27	0.44
Parent College Degree	294,602	0.34	0.47	281,684	0.34	0.47
Parent STEM Degree	294,602	0.05	0.23	281,684	0.05	0.23
School Variables						
Class Size	294,602	23.17	4.71	281,684	23.26	4.70
Cohort Size	294,602	107.65	48.26	281,684	107.32	48.71
Classes in Cohort	294,602	4.73	2.13	281,684	4.71	2.15
Schools in Municipality	294,602	34.67	45.95	281,684	34.63	45.76
Share Females	294,601	0.49	0.11	281,684	0.49	0.12
Share Females Cohort	294,602	0.49	0.07	281,684	0.49	0.07
Average Grade	294,602	-0.16	0.96	281,684	0.17	1.01
Diagnostic Variables, %						
Mental Health Disorder	294,602	4.21	20.09	281,684	5.16	22.11
Hyperactivity and Conduct Disorder	294,602	1.97	13.91	281,684	1.09	10.40
Substance-Related Disorder	294,602	0.93	9.60	281,684	1.13	10.59
Depression and Mood Disorder	294,602	0.49	6.98	281,684	1.29	11.28
Eating and Sleeping Disorder	294,602	0.14	3.76	281,684	0.90	9.46
Neurotic and Stress Disorder	294,602	0.66	8.08	281,684	1.56	12.38
Learning and Socialisation Disorder	294,602	0.53	7.24	281,684	0.29	5.41
Schizotypal Disorder	294,602	0.03	1.58	281,684	0.04	1.92
Adult Personality Disorder	294,602	0.02	1.40	281,684	0.05	2.33
Organic Disorder	294,602	0.02	1.29	281,684	0.02	1.43
Mental Retardation	292,505	0.04	2.09	279,686	0.04	1.96
Parental Diagnostic Variables, %						
Mental Health Disorder	294,602	5.35	19.70	281,684	5.30	19.57
Hyperactivity and Conduct Disorder	294,602	0.43	6.56	281,684	0.41	6.39
Substance-Related Disorder	294,602	2.17	14.57	281,684	2.26	14.87
Depression and Mood Disorder	294,602	4.28	20.23	281,684	4.22	20.11
Eating and Sleeping Disorder	294,602	0.35	5.93	281,684	0.37	6.03
Neurotic and Stress Disorder	294,602	5.04	21.88	281,684	5.03	21.86
Learning and Socialisation Disorder	294,602	0.05	2.21	281,684	0.04	1.95
Schizotypal Disorder	294,602	0.57	7.53	281,684	0.57	7.54
Adult Personality Disorder	294,602	0.47	6.87	281,684	0.47	6.87
Organic Disorder	294,602	0.19	4.35	281,684	0.20	4.47
Mental Retardation	294,602	0.02	1.34	281,684	0.01	1.21

Note: This table presents summary statistics for family- and school-level background variables as well as diagnostic variables. To avoid missing variables because of 0 income, we use $\ln(\text{income}+1)$. Income is missing for some parents, which explains the lower number of observations. For those observations, we replace missing income by 0 income in the estimations. Thus, we use data from all parents, also from those without income information, in the estimations. Parental mental health issues in this table include all parental mental health issues in the sample period. When using parental mental health before school start as a control variable, we have missing values for students in cohorts before 2009. When we use it as a control variable, we therefore include a dummy for each value of the categorical variable capturing no mental health diagnosis, any mental health diagnosis, or a missing mental health diagnosis. Correspondence between diagnoses names and formal ICD definitions is the same as in Table 3. Average grades refer to the grade at the end of compulsory schooling and are standardized.

B Balance and Placebo Checks

B.1 Balance Checks

Table B.1: Balance Checks: Family Characteristics

Dependent variable	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Immigrant	0.003 (0.006)	-0.005 (0.005)	-0.004 (0.005)	-0.004 (0.005)	-0.002 (0.005)	-0.008 (0.006)	-0.004 (0.006)
2nd-Generation Immigrant	0.017 (0.016)	-0.016 (0.013)	0.002 (0.008)	0.003 (0.008)	0.007 (0.008)	0.000 (0.010)	0.007 (0.009)
Number Siblings	0.032 (0.024)	-0.017 (0.016)	-0.014 (0.016)	-0.010 (0.016)	-0.008 (0.016)	-0.030 (0.020)	-0.025 (0.020)
Adopted	-0.001 (0.001)	-0.001 (0.001)	-0.001 (0.001)	-0.001 (0.001)	-0.001 (0.001)	-0.000 (0.001)	-0.000 (0.001)
Parent Unknown	-0.019 (0.012)	-0.023* (0.013)	-0.001 (0.002)	-0.002 (0.002)	-0.001 (0.002)	-0.004 (0.003)	-0.003 (0.003)
Teenage Parent	-0.002 (0.002)	-0.002 (0.002)	-0.002 (0.002)	-0.002 (0.002)	-0.001 (0.002)	-0.000 (0.002)	0.001 (0.002)
Log Family Income	-0.103 (0.072)	-0.028 (0.052)	-0.025 (0.050)	-0.039 (0.052)	-0.064 (0.052)	-0.057 (0.065)	-0.107* (0.065)
Parents High School	0.007 (0.013)	-0.001 (0.009)	0.005 (0.008)	0.003 (0.008)	-0.001 (0.008)	-0.000 (0.010)	-0.007 (0.010)
Parent Vocational Degree	0.019** (0.009)	0.009 (0.007)	0.002 (0.006)	0.001 (0.006)	-0.001 (0.006)	0.009 (0.008)	0.005 (0.008)
Parent College Degree	0.039** (0.018)	0.003 (0.012)	0.019** (0.008)	0.018** (0.008)	0.016** (0.008)	0.021** (0.010)	0.017 (0.010)
Parent STEM Degree	0.010* (0.006)	0.006* (0.003)	0.005 (0.003)	0.004 (0.003)	0.003 (0.003)	0.003 (0.004)	0.001 (0.004)
Mental Health Disorder	0.078 (0.181)	0.112 (0.184)	0.085 (0.175)	0.084 (0.178)	0.122 (0.179)	0.192 (0.212)	0.243 (0.213)
Hyperactivity and Conduct Disorder	-0.027 (0.020)	-0.040* (0.022)	-0.041* (0.022)	-0.038* (0.022)	-0.038* (0.022)	-0.048* (0.028)	-0.049* (0.028)
Substance-Related Disorder	0.074 (0.102)	0.112 (0.104)	0.100 (0.102)	0.098 (0.104)	0.122 (0.104)	0.078 (0.125)	0.104 (0.125)
Depression and Mood Disorder	0.004 (0.091)	0.024 (0.096)	0.017 (0.094)	0.015 (0.096)	0.022 (0.096)	0.086 (0.109)	0.097 (0.109)
Eating and Sleeping Disorder	0.026 (0.024)	0.022 (0.026)	0.021 (0.026)	0.022 (0.026)	0.022 (0.026)	0.023 (0.030)	0.023 (0.030)
Neurotic and Stress Disorder	0.103 (0.100)	0.090 (0.102)	0.086 (0.098)	0.105 (0.100)	0.111 (0.101)	0.141 (0.120)	0.156 (0.120)
Learning and Socialisation Disorder	0.006 (0.005)	0.007 (0.006)	0.007 (0.006)	0.007 (0.006)	0.007 (0.006)	0.005 (0.005)	0.004 (0.005)
Schizotypal Disorder	-0.060 (0.053)	-0.050 (0.054)	-0.055 (0.055)	-0.050 (0.056)	-0.047 (0.056)	-0.019 (0.069)	-0.018 (0.069)
Adult Personality Disorder	0.037 (0.041)	0.049 (0.042)	0.047 (0.042)	0.041 (0.043)	0.046 (0.043)	-0.004 (0.051)	0.001 (0.051)
Organic Disorder	0.016 (0.015)	0.021 (0.016)	0.021 (0.016)	0.017 (0.017)	0.019 (0.017)	0.019 (0.023)	0.020 (0.023)
Mental Retardation	0.003 (0.004)	0.004 (0.004)	0.004 (0.004)	0.005 (0.005)	0.005 (0.005)	0.003 (0.007)	0.003 (0.007)

Note: The table shows the estimated relationship between student family characteristics and the share of female peers in their classroom. The diagnostic variables refer to parental diagnoses observed before the school start of the child. The layout of specifications corresponds to that in Table 1. All mental health diagnoses in the table refer to parental diagnoses prior to starting school. All fixed effects and controls are fully interacted with student gender as in the main specifications. Columns (5) and (7) include school-level controls. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

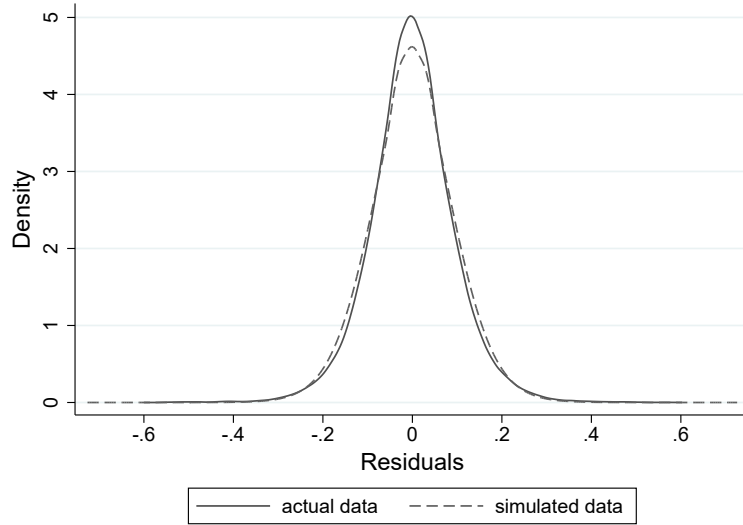
Table B.2: Mental Health Before School Start and Gender Composition

	Prior Mental Health Diagnoses [0,100]						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
All Genders (Avg: .72)	-0.17 (0.18)	-0.11 (0.18)	-0.01 (0.18)	-0.02 (0.18)	-0.02 (0.18)	-0.16 (0.19)	-0.17 (0.19)
Males (Avg: .95)	-0.22 (0.27)	-0.27 (0.27)	-0.13 (0.27)	-0.12 (0.28)	-0.13 (0.27)	-0.23 (0.30)	-0.26 (0.30)
Females (Avg: .47)	-0.01 (0.20)	0.06 (0.23)	0.13 (0.22)	0.04 (0.23)	0.04 (0.23)	-0.09 (0.25)	-0.08 (0.25)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	227,773	227,773	227,773	227,773	227,773	227,773	227,773
Schools	855	855	855	855	855	855	855

Note: The table shows the estimated relationship between the incidence of mental health diagnoses prior to an individual starting school and the share of female peers in their subsequent classroom. The first row shows the results for the sample with both genders (All Genders). In this specification, we interact all fixed effects and controls with gender. The next two rows show coefficient estimates from separate estimations based on samples including either only males or only females. The number of observations comes from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. Regressions only include students who finish compulsory schooling after 2008 as we need to observe their mental health at the start of the schooling period (the mental health data starts in 2001). * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

B.2 Peer Share Residual Distribution

Figure B.1: Actual and Simulated Variation in the Share of Female Peers



Note: The figure shows kernel density plots of residuals from regressions of the share of female peers on school-by-cohort fixed effects. The solid line corresponds to residuals from a single regression using the actual data, whereas the dashed line corresponds to residuals from 1,000 regressions in simulated data in which students are randomly assigned to classes within their school.

B.3 Student Gender

Table B.3: Student Gender and Gender Composition

	Share Females in Class (Mean: .489)						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Female	0.0004 (0.0010)	0.0004 (0.0010)	0.0004 (0.0010)	0.0004 (0.0010)	0.0003 (0.0010)	0.0023 (0.0016)	0.0021 (0.0016)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	576,285	576,285	576,285	576,285	576,285	576,285	576,285
Schools	855	855	855	855	855	855	855

Note: The table shows the estimated relationship between the share of female peers and a student's own gender. Following Guryan, Kroft and Notowidigdo (2009), we control for the cohort-level leave-one-out class share of females. That is, we control for the share of females in the rest of the cohort. Other additional controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality as indicated. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

B.4 Class Fixed Effects

Table B.4: Stability of the Gender Difference when Accounting for Class Fixed Effects

	Mental Health Diagnosis [0,100]						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Males (Avg: 4.21)	1.72*** (0.39)	1.74*** (0.40)	1.85*** (0.39)	1.86*** (0.48)	2.00*** (0.47)		
Females (Avg: 5.16)	0.44 (0.42)	0.36 (0.42)	0.40 (0.42)	0.14 (0.51)	0.20 (0.50)		
Gender Difference	1.27** (0.54)	1.38** (0.54)	1.45*** (0.54)	1.72*** (0.65)	1.80*** (0.64)	1.28** (0.57)	1.24** (0.56)
School FE	X	X	X	-	-	-	-
Cohort FE	X	X	X	-	-	-	-
School Trends	-	X	X	-	-	-	-
Controls	-	-	X	-	X	-	X
School \times Cohort FE	-	-	-	X	X	-	-
Class FE	-	-	-	-	-	X	X
Observations	576,286	576,286	576,286	576,286	576,286	576,286	576,286
Schools	855	855	855	855	855	855	855
<i>R</i> -squared	0.01	0.02	0.02	0.04	0.04	0.06	0.06

Note: The table shows the estimated relationship between the incidence of mental health diagnoses in compulsory school and the share of female peers in the classroom. The estimates denoted by “Males” and “Females” come from interacting the female peer share with the gender dummy. Gender difference then captures the difference between the impact of the share of female peers on males vs. females based on the interaction term. Columns (6) and (7) show the specifications which include classroom fixed effects. The number of observations comes from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

B.5 Placebo Checks

Table B.5: Previous and Subsequent Cohorts and Mental Health

	Specifications				
	(1)	(2)	(3)	(4)	(5)
All Genders (t-1)	0.04 (0.40)	-0.03 (0.40)	0.34 (0.35)	0.53 (0.36)	0.47 (0.36)
All Genders (t+1)	-0.24 (0.34)	-0.41 (0.35)	0.01 (0.30)	-0.07 (0.31)	-0.07 (0.31)
Males (t-1)	-0.06 (0.46)	-0.02 (0.50)	0.41 (0.44)	0.69 (0.46)	0.59 (0.46)
Males (t+1)	-0.12 (0.41)	-0.11 (0.43)	0.32 (0.39)	0.12 (0.41)	0.12 (0.41)
Females (t-1)	0.04 (0.53)	-0.03 (0.55)	0.26 (0.51)	0.35 (0.53)	0.34 (0.53)
Females (t+1)	-0.46 (0.45)	-0.73 (0.46)	-0.31 (0.42)	-0.24 (0.45)	-0.25 (0.44)
School FE	-	X	X	X	X
Cohort FE	-	-	X	X	X
School Trends	-	-	-	X	X
Controls	-	-	-	-	X

Note: The table shows the estimated relationship between the incidence of mental health diagnoses and the share of female peers in the previous (t-1) and the subsequent cohorts (t+1). The first rows shows the results for the sample with both genders (All Genders). In these specifications, we interact all fixed effects and controls with gender. Males refers to the sample only consisting of boys, females refers to the sample only consisting of girls. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

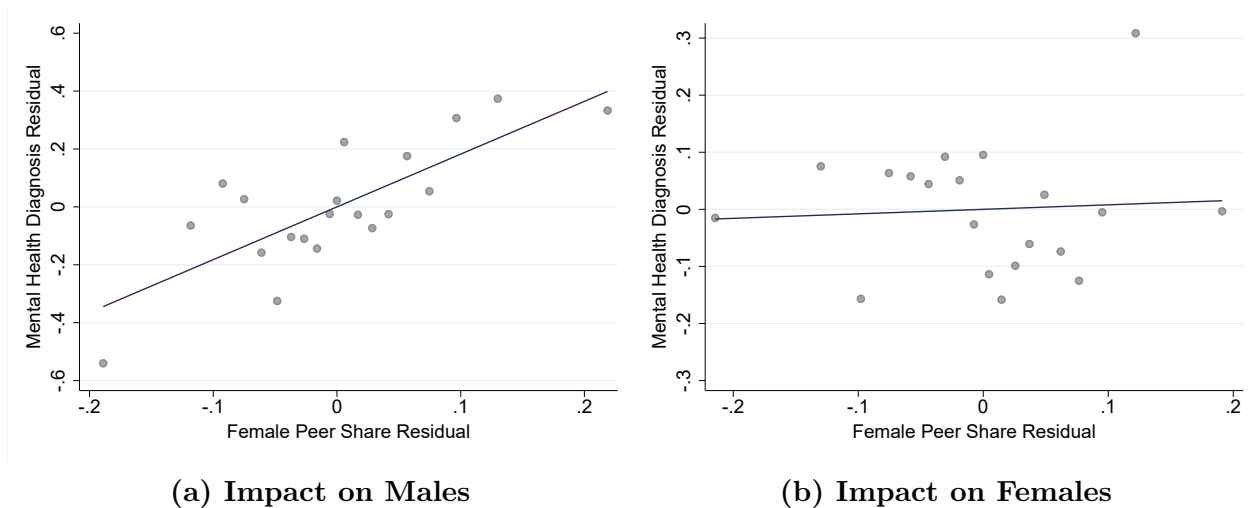
Table B.6: Gender Composition and Mental Health Problems with Biological Origins

Dependent variable: [0,100]	Specifications						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Mental Retardation (Avg: .04)	0.01 (0.02)	0.02 (0.03)	0.02 (0.03)	0.01 (0.03)	0.02 (0.03)	0.01 (0.03)	0.01 (0.03)
Organic Disorders (Avg: .02)	-0.01 (0.01)	-0.00 (0.02)	-0.00 (0.02)	-0.00 (0.02)	-0.00 (0.02)	0.01 (0.02)	0.00 (0.02)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	576,285	576,285	576,285	576,285	576,285	576,285	576,285
Schools	855	855	855	855	855	855	855

Note: The table shows the estimated relationship between the incidence of mainly biologically-driven mental health diagnoses and the share of female peers in the classroom. Each of the two rows shows the results for the sample with all genders. In this specification, we interact all fixed effects and controls with gender. The first row shows the effects on organic disorders, which include disorders induced by cerebral disease, brain injury, or other to cerebral dysfunction (F00-F09); the second row shows diagnoses related to mental retardation (F70-F79). Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

C Gender Composition and Mental Health

Figure C.1: Gender Composition and Mental Health, Results for Males and Females Separately



Note: The figure shows the relationships between the residual of mental health diagnoses and residual of the share of female peers in the classroom by gender. The residuals stem from regressions of the respective variables on cohort-by-school fixed effects. The dots show the binned averages across 20 quantiles. The line shows the linear fit from OLS regressions. The figure shows the relationship separately for boys and girls as shown in Table 1 rows (2) and (3).

D Robustness and Sample Checks

Table D.1: Inpatient and Outpatient Mental Health Diagnoses

	Mental Health Diagnosis [0,100]						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
All Diagnoses (Avg: 4.67)	1.04*** (0.30)	0.96*** (0.30)	1.05*** (0.29)	1.02*** (0.29)	1.10*** (0.29)	0.97*** (0.35)	1.06*** (0.35)
Outpatient (Avg: 4.01)	1.11*** (0.28)	0.99*** (0.27)	1.08*** (0.26)	1.07*** (0.26)	1.12*** (0.26)	0.90*** (0.31)	0.96*** (0.31)
Inpatient (Avg: 1.23)	0.20 (0.14)	0.19 (0.14)	0.20 (0.14)	0.20 (0.14)	0.23 (0.14)	0.29* (0.17)	0.36** (0.17)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	576,285	576,285	576,285	576,285	576,285	576,285	576,285
Schools	855	855	855	855	855	855	855

Note: The table shows the estimated relationship between the incidence of mental health diagnoses as recorded in the inpatient and the outpatient registers, and the share of female peers in the classroom. The first row shows the results for diagnoses from the outpatient and inpatient register. These correspond to the estimates in the first row of Table 1. The second row shows coefficient estimates using an indicator capturing a diagnosis in the outpatient register as the dependent variable. The third row shows the estimates based only on an indicator constructed from inpatient diagnoses. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Table D.2: Cohort Gender Composition and Mental Health

	Mental Health Diagnosis [0,100]				
	(1)	(2)	(3)	(4)	(5)
Diagnoses (Avg: 4.67)	1.04* (0.60)	0.69 (0.57)	1.11** (0.54)	1.00* (0.55)	1.04* (0.55)
Outpatient (Avg: 4.01)	1.34** (0.59)	0.88 (0.55)	1.31** (0.52)	1.28** (0.53)	1.31** (0.52)
Males (Avg: 4.21)	0.94 (0.73)	1.03 (0.75)	1.44** (0.72)	1.23 (0.76)	1.30* (0.76)
Females (Avg: 5.16)	0.63 (0.76)	0.35 (0.78)	0.78 (0.76)	0.88 (0.79)	0.90 (0.79)
School FE	-	X	X	X	X
Cohort FE	-	-	X	X	X
School Trends	-	-	-	X	X
Controls	-	-	-	-	X
Observations	576,286	576,286	576,286	576,286	576,286
Schools	855	855	855	855	855
<i>R</i> -squared	0.01	0.01	0.02	0.02	0.02

Note: The table shows the estimated relationship between the incidence of mental health diagnoses and the share of female peers in a student's cohort. The first row shows the results for the sample with both genders (All Genders). In this specification, we interact all fixed effects and controls with gender. The next row shows the results only for outpatient diagnoses. The final two rows show coefficient estimates from separate estimations based on samples including either only males or only females, using the combined data from outpatient and inpatient registers. The number of observations and the corresponding *R*-squared come from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Table D.3: Sample Restrictions

Diagnosis:	All		Males		Females	
	No	Yes	No	Yes	No	Yes
Last-Year Diagnosis	0.95*** (0.35)	0.79*** (0.25)	1.79*** (0.45)	1.09*** (0.30)	0.09 (0.48)	0.48 (0.37)
Without top/bottom 5%	0.95*** (0.35)	0.94* (0.49)	1.79*** (0.45)	1.53** (0.66)	0.09 (0.48)	0.33 (0.71)
Non-movers	0.95*** (0.35)	0.78 (0.79)	1.79*** (0.45)	2.55** (0.99)	0.09 (0.48)	-0.99 (1.12)

Note: The table shows the estimated relationship between the incidence of mental health diagnoses and the share of female peers in the classroom. The specification used for each estimate includes school-by-cohort fixed effects and the vector of individual- and school-level controls. The sample used is either all genders, only males, or only females. “No” refers to the estimates without any change to the sample or variable definition, “Yes” refers to the sample applying the sample restriction or the alternative definition of the dependent variable. For instance, specification 2, row 1, shows the effect of the share of female peers using all genders for estimation, but employs a different definition of the dependent variable. Specification 2, row 2, only uses observations that are not in the top or bottom 5% of the share of female peers. Last-Year Diagnosis means that the dependent variable refers to the indicator for a mental health diagnosis taking value 100 if there was a mental health code present in the last year of compulsory school and 0 otherwise. Without top/bottom 5% refers to dropping the classes with the 5% highest and lowest share of female peers. Non-movers refers to only including non-movers in the estimation. Non-movers are students who resided in the same neighbourhood for the entire nine years of compulsory schooling as well as the year before starting compulsory school. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Table D.4: Only Observations for Which We Observe Long-run Outcomes

	Mental Health Diagnosis [0,100]						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
All Genders (Avg: 3.25)	1.14*** (0.29)	1.05*** (0.28)	1.05*** (0.28)	0.94*** (0.28)	1.00*** (0.28)	0.82*** (0.32)	0.90*** (0.31)
Males (Avg: 2.86)	1.45*** (0.37)	1.46*** (0.37)	1.46*** (0.37)	1.28*** (0.38)	1.34*** (0.37)	1.24*** (0.43)	1.34*** (0.42)
Females (Avg: 3.66)	0.68* (0.39)	0.63 (0.39)	0.64* (0.38)	0.64 (0.39)	0.69* (0.39)	0.40 (0.45)	0.46 (0.44)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	414,716	414,716	414,716	414,716	414,716	414,716	414,716
Schools	890	890	890	890	890	890	890
<i>R</i> -squared	0.00	0.02	0.02	0.02	0.02	0.04	0.04

Note: The table shows the estimated relationship between the incidence of mental health diagnosis during compulsory schooling and the share of female peers in the classroom, with the sample restricted to students for whom we observe long-term mental health. The first row shows the results for the sample with both genders (All Genders). In this specification, we interact all fixed effects and controls with gender. The next two rows show coefficient estimates from separate estimations based on samples including either only males or only females. The number of observations and the corresponding *R*-squared come from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level.* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

E Mechanisms

E.1 Self-Harm

Table E.1: Gender Composition and Intentional Self-Harm

	Intentional Self-Harm [0,100]						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
All Genders (Avg: 0.49)	0.02 (0.08)	0.01 (0.09)	0.00 (0.09)	0.03 (0.09)	0.03 (0.09)	0.06 (0.10)	0.07 (0.10)
Males (Avg: .19)	0.17** (0.07)	0.18** (0.07)	0.18** (0.07)	0.17** (0.07)	0.18** (0.07)	0.16* (0.08)	0.16* (0.08)
Males (Avg: .81)	-0.26* (0.15)	-0.18 (0.16)	-0.18 (0.16)	-0.13 (0.16)	-0.12 (0.16)	-0.05 (0.19)	-0.03 (0.19)
School FE	-	X	X	X	X	-	-
Cohort FE	-	-	X	X	X	-	-
School Trends	-	-	-	X	X	-	-
Controls	-	-	-	-	X	-	X
School \times Cohort FE	-	-	-	-	-	X	X
Observations	576,285	576,285	576,285	576,285	576,285	576,285	576,285
Schools	855	855	855	855	855	855	855
<i>R</i> -squared	0.00	0.01	0.01	0.01	0.01	0.03	0.03

Note: The table shows the estimated relationship between the incidence of intentional self-harm and the share of female peers in the classroom. Intentional self-harm is an indicator variable based on whether a student has a diagnostic code the corresponding to the diagnostic category (ICD-10 codes X60 to X84). The first row shows the results for the sample with both genders (All Genders). In this specification, we interact all fixed effects and controls with gender. The next two rows show coefficient estimates from separate estimations based on samples including either only males or only females. The number of observations and the corresponding *R*-squared come from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

E.2 The Role of Educational Attainment and the Classroom Environment

Table E.2: Gender Composition and Compulsory School Grades

	Mean Standardized Grade				
	(1)	(2)	(3)	(4)	(5)
Males (Avg: -.16)	-0.19*** (0.03)	-0.21*** (0.03)	-0.21*** (0.02)	-0.23*** (0.03)	-0.25*** (0.03)
Females (Avg: .17)	0.13*** (0.03)	0.12*** (0.03)	0.09*** (0.03)	0.13*** (0.04)	0.08** (0.03)
School FE	X	X	X	-	-
Cohort FE	X	X	X	-	-
School Trends	-	X	X	-	-
Controls	-	-	X	-	X
School \times Cohort FE	-	-	-	X	X
Observations	576,286	576,286	576,286	576,286	576,286
School-Cohorts	855	855	855	855	855
<i>R</i> -squared	0.14	0.15	0.21	0.18	0.24

Note: The table shows the estimated relationship between the mean grade in compulsory school and the share of female peers in the classroom. The grades are standardized for the entire population of students (therefore, boys have a “negative” average as their grades are, on average, worse than girls). The first row shows the effect on males, and the second on females. The number of observations comes from the sample including both genders. Specifications in this table correspond to the ones in columns (3) - (7) in Table 1. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. Standard errors (in parentheses) are based on clustering at the school level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Table E.3: Gender Composition and Mental Health Conditional on Compulsory School Grades

	All		Male		Female	
	No	Yes	No	Yes	No	Yes
Compulsory-School Controls	1.06*** (0.35)	0.68** (0.34)	1.92*** (0.45)	0.88** (0.43)	0.17 (0.48)	0.52 (0.48)
Controls	X	X	X	X	X	X
School \times Cohort FE	X	X	X	X	X	X
Observations	576,285	576,285	294,601	294,601	281,684	281,684
Schools	855	855	855	855	855	855
<i>R</i> -squared	0.04	0.07	0.04	0.07	0.04	0.07

Note: The table shows the estimated relationship between the incidence of mental health diagnoses and the share of female peers in the classroom, controlling for one's grades in compulsory school. The sample used is either all genders, only males, or only females. Standard errors (in parentheses) are based on clustering at the school level. The female peer share has a strong effect on boys' academic performance, and grades can therefore explain part of the results. The specification used for each estimate includes school-by-cohort fixed effects and the vector of controls. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Table E.4: Controlling for Leave-one-out Grades and Mental Health in Classroom

	Mental Health Diagnosis [0,100]			
	(1)	(2)	(3)	(4)
All Genders (Avg: 4.67)	1.06*** (0.35)	1.14*** (0.34)	1.62*** (0.44)	1.96*** (0.43)
Males (Avg: 4.21)	1.92*** (0.45)	1.95*** (0.45)	2.48*** (0.52)	2.72*** (0.52)
Females (Avg: 5.16)	0.17 (0.48)	0.30 (0.47)	0.71 (0.56)	1.17** (0.56)
Controls	X	X	X	X
School \times Cohort FE	X	X	X	X
Leave-Out Class Grade	-	X	-	X
Class Mental Health Share	-	-	X	X
Observations	576,285	576,285	576,285	576,285
Schools	855	855	855	855
<i>R</i> -squared	0.04	0.04	0.05	0.05

Note: The table shows the estimated relationship between the incidence of mental health diagnosis during compulsory schooling and the share of female peers in the classroom, controlling for the leave-one-out mean grade and share of students with a mental health condition in the classroom. The first row shows the results for the sample with both genders (All Genders). In this specification, we interact all fixed effects and controls with gender. The next two rows show coefficient estimates from separate estimations based on samples including either only males or only females. The number of observations and the corresponding *R*-squared come from the sample including both genders. Controls include parental education, income, and mental health as well as class size, cohort size, and the number of schools in the municipality. The first column corresponds column (7) in Table 1. In each subsequent column, we incrementally add each of the two leave-one-out controls. Standard errors (in parentheses) are based on clustering at the school level.* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

E.3 Bullying

Table E.5: Gender Composition and Physical Harm

Dependent variable: [0,100]	All	Males	Females
	(1)	(2)	(3)
Assault	0.05 (0.08)	0.07 (0.14)	0.03 (0.07)
Physical Injury	-3.70*** (0.77)	-2.49** (1.14)	-4.92*** (0.88)
Controls	X	X	X
School \times Cohort FE	X	X	X
Observations	576,285	294,601	281,684
Schools	855	855	855
<i>R</i> -squared	0.06	0.05	0.05

Note: The table shows the estimated relationship between the incidence of physical assault/injury and the share of female peers in the classroom. The sample used is either all genders, only males, or only females. The Assault category captures assault by bodily force (Y04). We thereby exclude domestic assault. Physical Injury includes any injury sustained to any part of the body (S00 - S99). The specification used for the regressions above includes school-by-cohort fixed effects and the main vector of controls. Standard errors (in parentheses) are based on clustering at the school level.

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

E.4 The Role of Parents

Table E.6: The Role of Parents

Variable	Diagnosis [0,100]		
	(1)	(2)	(3)
Share Females	1.70** (0.77)	1.88*** (0.49)	1.53*** (0.55)
Parent Diagnosis	4.74*** (1.22)		
Parent Diagnosis × Share Females	2.07 (2.43)		
Parent College		-0.70** (0.35)	
Parent College × Share Females		-0.42 (0.71)	
Above-Median Income			-2.22*** (0.33)
Above-Median Income × Share Females			0.16 (0.65)
School × Cohort FE	X	X	X
Observations	576,285	576,285	576,285
Schools	855	855	855
<i>R</i> -squared	0.03	0.03	0.03

Note: The table shows the estimated relationship between the incidence of mental health diagnoses and the share of female peers in the classroom. The following two rows shows the difference in response to the gender peer share in students who have a parent with a concurrent mental health problem; the next two rows for students with a parent with a college degree; the following row for those whose family has above-median income. For family income, we construct a dummy variable indicating an above- or below-median value. Standard errors (in parentheses) are based on clustering at the school level.
 * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$