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INOUE, Atsushi

Nippon Institute for Research Advancement

TANAKA, Ryuichi RIETI



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The Rank of Socioeconomic Status within A Class and The Incidence of School Bullying and School Absence*

Atsushi Inoue[†]
Nippon Institute for Research Advancement
Ryuichi Tanaka[‡]
University of Tokyo / RIETI

Abstract

Does the relative wealth of students' households affect the incidence of risky behaviors of students in school? We estimate the effect of the rank of the socioeconomic status (SES) of the students' household within a class on the incidence of school bullying and school absence. We exploit the variation of SES rank within a class generated by the almost-random assignment of students to classes. Using the data from middle-school students in the Trends in International Mathematics and Science Study (TIMSS), we find that although the absolute level of SES is negatively associated with these incidences, students with a high SES rank within a class are more likely to be the victims of school bullying and to be absent from school.

Keywords: Rank; Socioeconomic Status; School Bullying; School Absence; TIMSS JEL classification: I21.

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[†] Nippon Institute for Research Advancement, Ebisu Garden Place Tower 34th Floor, 4-20-3 Ebisu, Shibuya-ku, Tokyo 150-6034, Japan. E-mail: ainoue@nira.or.jp.

^{*} Corresponding author. Institute of Social Science, The University of Tokyo, 7-3-1 Hongo, Bunkyo-ku, Tokyo 113-0033, Japan. E-mail: ryuichi.tanaka@iss.u-tokyo.ac.jp.

1. Introduction

In most advanced countries, primary and lower secondary schooling is compulsory. These levels of schooling are considered crucial for human capital formation because students learn basic skills, such as mathematics and languages, which are needed for further learning in higher education, and are required to accomplish tasks in the labor markets. Since education in these phases of compulsory schooling determines the quality of students' later life, the governments of most advanced countries are heavily responsible for providing high-quality compulsory education.

For compulsory education to work, it is imperative to maintain an effective learning environment in schools by mitigating students' risky behaviors such as school bullying and school absence observed globally. The Organisation for Economic Co-operation and Development (OECD) (2019) shows that 23% of students reported being bullied at least a few times a month, and 21% of students had skipped a whole day of school at least once in two weeks across OECD countries.⁴ Being bullied and absent from school adversely affect human capital accumulation (e.g., Le et al., 2005; Brown and Taylor, 2008; Ammermueller, 2012; Eriksen et al., 2014 for bullying, and Aucejo and Romano, 2016; Gershenson et al., 2017; Liu and Gershenson, 2021 for school absence). Hence, understanding the determinants of bullying victimization in schools and school absence is essential to mitigate these adverse effects on human capital formation.

Among the potential factors of school bullying and absence, the association between parental socioeconomic status (SES) and risky behaviors has been studied. In their meta-analysis, Tippett and Wolke (2014) find that the SES of a student's household is negatively associated with the likelihood of being bullied. Using data on Finnish youth from 2000 to 2015, Knaappila et al. (2018) find that bullying victimization and perpetration are associated with socioeconomic adversity. In addition, several studies have found that students with low SES have a high probability of absenteeism (e.g., Ready, 2010; Morrissey et al., 2014; Gottfried and Gee, 2017; Gennetian et al., 2018; Gubbels et al., 2019). These studies indicate that household SES matters not only for educational investment by households, but also for students' risky behaviors.

Although the relationship between the SES of students' households and these risky behaviors has been well studied, relatively little is known about how the *relative* SES of

⁴ Some studies show that disrupting face-to-face interactions during the Covid-19 pandemic reduces school bullying and cyberbullying. For example, Werner and Woessmann (2021) find that nearly half of German parents reported that their children were less likely to experience bullying victimization after school closed. Using U.S. data from Google Internet searches, Bacher-Hicks, Goodman, and Mulhern (2022) also obtain results suggesting a dramatically decreased incidence of in-person and online bullying.

students' households affects bullying victimization and absentee behavior. The relative SES of students' households may matter for these risky behaviors of students for at least two reasons. First, relative SES (or the rank of SES in school/class) captures the "quality" of school/classmates relative to their own SES. Many studies have examined the effects of peer characteristics such as gender, race, social and economic background, and classmates' abilities on various educational outcomes (e.g., Epple and Romano, 2011; Sacerdote, 2014). These studies indicate that peer quality is vital in shaping students' educational outcomes. In addition, several studies have observed the importance of peers in school bullying victimization and absenteeism (e.g., Hong and Espelage, 2012; Wolke and Lereya, 2015). Hence, SES rank as a proxy for peer quality relative to one's own quality may be related to the incidence of school bullying and absenteeism.

Second, relative SES is related to the concept of relative deprivation and thus may be systematically related to students' risky behaviors. Relative deprivation is caused by an imbalance in the allocation of resources within a given group, which can lead to feelings of unfairness, stress, and inferiority due to differences in social status (Wilkinson and Pickett, 2009). Mangyo and Park (2011) study how individual behavior and well-being are affected by relative deprivation, which is SES relative to a reference group, such as a class in school. Balsa, French, and Regan (2014) examine whether a student's relative SES directly affects their risky behaviors and find that relatively low SES is positively associated with the consumption of alcohol and cigarettes for males. Napoletano et al. (2016) find that relative affluence is negatively associated with bullying victimization among Canadian adolescents. Hence, SES rank may affect the incidence of school bullying and absenteeism through the channel of relative deprivation.

In this study, we examine the effect of the SES rank of students' households within a class on the incidence of school bullying and school absence. A common issue in identifying the effects of peer characteristics is that students may not be randomly grouped into classes. If a class assignment is non-random and unobserved student characteristics are associated with both SES rank and outcomes, the estimated effect of SES rank will be biased. To address this potential endogeneity, researchers often use random assignment (e.g., Sacerdote, 2001) and variation across cohorts (e.g., Lavy and Schlosser, 2011). We exploit the variation in SES rank within a class generated by the almost random assignment of students to classes within a school by restricting our sample to schools with multiple classes, where the average class SES is balanced across classes. Even if two students have the same SES, their SES ranks within a class can differ depending on which class they are assigned. Random assignment of students to a class generates random variation in SES rank within a class for students sharing the same

absolute level of SES. Our identification strategy is close to that of Ammermueller and Pischke (2009), which exploits the variation in peer ability within schools across classes to estimate the peer effect. We restrict our samples to those with no statistically significant differences in mean SES across classes and confirm that class arrangements are independent of household SES and as good as random for our restricted samples by employing Monte Carlo simulations.

Using the data obtained from the 2015 Trends in International Mathematics and Science Study (TIMSS), we construct the SES index following Hanushek et al. (2022) and estimate the SES rank effects on students' bullying victimization and absences from schools. We find that while the absolute level of SES is negatively associated with the incidence of school bullying and school absence, controlling for SES, eighth-grade students with a *high* SES rank within a class are *more* likely to be victims of school bullying and to be absent from school. We also find some heterogeneity in the SES rank effects: the effect is stronger for girls and countries with greater inequality. We examine the SES rank effect on other outcomes and find that students with higher SES ranks are likely to think that their teachers are unfair to them and that they cannot learn much in school. These SES rank-related changes may be attributed to absentee behavior. Importantly, we find weak effects of SES rank for fourth-grade students. These findings may be because elementary school students in friendships rarely recognize the SES rank, but become more aware of it as they age.

In addition to the literature discussed above, our study contributes to the literature on the effects of ranks within schools and classes on various educational outcomes. Murphy and Weinhardt (2020) find that the ranking of test scores in primary school affects test scores, confidence, and subject choice during secondary schooling. Similarly, Elsner and Isphording (2017) find that a student's ordinal academic rank in a high school cohort positively affects high school graduation and college attendance. Beyond academic achievements, Elsner and Isphording (2018) find a mitigating effect of ordinal ability rank in high school on risky behaviors such as smoking and alcohol drinking. These studies suggest that being ranked higher in academic achievement has beneficial effects on students, such as increasing their confidence, allowing them to associate with other highly ranked students, and providing them with more support from teachers and parents. As a closely related study to ours, Ballatore, Paccagnella, and Tonello (2020) examine the effect of age rank within schools on the incidence of school bullying as an outcome variable. They find that a higher rank in the age distribution decreases the chances of being bullied. We study the effect of SES rank as a new source that generates a rank effect on non-academic outcomes in schools.

The remainder of this paper is organized as follows. In the next section, we explain our empirical framework. Section 3 explains the study's data. We report the estimation results in section 4, and section 5 concludes the study.

2. Econometric Framework

To estimate the effect of SES rank within a class on school bullying and school absence, we estimate the following regression model:

$$Y_{ics} = \beta_0 + \beta_1 SESRank_{ics} + f(SES_{ics}) + \beta_2 X_{ics} + \lambda_s + \varepsilon_{ics}$$
 (1)

where Y_{ics} denotes school bullying and school absence for student i in class c of school s; $SESRank_{ics}$ is the percentile SES rank of student i in class c; $f(SES_{ics})$ is a quartic polynomial of the SES of student i; X_{ics} contains the gender, age, and dummy variables for missing values of student i; λ_s is a school fixed effect; and ε_{ics} is the error term. The coefficient of interest is β_1 , which captures the effect of SES rank within a class on school bullying and school absence. We use cluster-robust standard errors at the class level, accounting for the correlation in the level of school bullying and school absence within the same class.

We identify the SES rank effect by including SES in a flexible way and school fixed effects, which allow us to compare students that have the same absolute SES as their school cohort but result in different ordinal positions in their class-cohort because of differences in the shape of the respective class-cohort SES distributions. A threat to the identification strategy assumption is that students can be placed in classes by unobservable factors. Thus, β_1 may reflect the selection of students with certain characteristics rather than the effect of SES rank. To address this concern, we restrict the samples to those who belong to schools that organize their classes independently of SES. Specifically, we restrict our sample to schools with multiple classes and the same average SES.

3. Data

Our primary data source is TIMSS 2015, an international survey measuring the achievements of eighth- and fourth-grade students in mathematics and science and gathering information regarding students, households, teachers, schools, and curricula in participating countries worldwide. ⁵ In both grades, approximately 600,000 students and 20,000 schools from 62 countries and regions participated in TIMSS 2015. A unique feature of TIMSS is that it includes data from many developed and developing countries. This advantage allows us to examine

⁵ The most recent TIMSS 2019 survey has less information on school belonging used for other outcomes in this study than the 2015 survey. Therefore, we chose to use the 2015 data.

whether the effects of SES rank vary systematically according to a country's economic conditions, such as income inequality and economic growth.

Samples were extracted by employing the two-stage random sample design, where schools are selected as the first stage and classes are selected as the second stage in each country. In principle, all students in the selected classes are sampled. However, the number of students in the sample may be smaller than the actual number because of student non-participation in the survey. In all estimations, we use a rescaled student sampling weight to assign the same weight to each country because we use student samples from many countries.

The sample sizes included in the raw data are 285,190 eighth-grade and 286,189 fourthgrade students. We restrict our sample as follows. First, we drop samples of countries that did not ask about parental education necessary for our SES index, the construction of which will be explained later. Second, we drop samples of eighth- and fourth-grade students who did not respond to bullying victimization and absenteeism questions, which are the dependent variables in our study. Third, we drop students in classes with fewer than ten students, as interactions among students in small class sizes could be different from those in other classes. Fourth, we drop samples from schools where only one class was extracted. Finally, and most importantly, we employ the F-test to test whether the average SES is balanced within a school across classes. If classes are randomly formed, there should be no statistical difference in the mean SES across classes. Interestingly, we find that around 18% of the sample in both grades belongs to schools where the average SES across classes is not balanced; thus, we drop the corresponding sample. The final sample consists of 89,042 eighth-grade students from 4,066 classes in 1,994 schools, and 97,601 fourth-grade students from 5,073 classes in 2,499 schools. See Table A1 for the sample sizes dropped at each stage. The results are robust to analysis using data without these sample selections.

Furthermore, we employ Monte Carlo simulations to verify that students in the final restricted sample are assigned almost randomly to the class. Specifically, we randomly assign each student to classes in their school, and then calculate the percentile rank of the SES within their class. This process is repeated 1,000 times. Next, we construct a 95% confidence interval under the null hypothesis that class assignments are random within the school. We then confirm whether the observed SES rank is outside the confidence interval. Table 1 presents the results. In the sample with balanced SES means across classes, 3.8% of the eighth-grade student sample and 2.7% of the fourth-grade student sample reject this null hypothesis (See Panel A in Table 1). However, for samples where the mean SES across classes is not balanced, we find that 37.7% of the eighth-grade sample and 32.4% of the fourth-grade sample reject it (See Panel B in Table 1). As can be seen in Table 1, the rate of rejection of the null hypothesis varies greatly

depending on whether the observed SES averages are balanced across classes. Since the rejection rates are meager in samples where the SES means are balanced across classes, we conclude that student class placement is almost random in our final restricted sample.

For missing information, we impute any missing values of independent variables by the median of the smallest unit available among the class, school, and country units.⁶ The imputed sample proportions for eighth-grade students (fourth-grade students) are less than 1% (1%) for gender, age, the number of books at home, and six other household possessions (computer, study desk, own room, Internet access, etc.) commonly asked in each country, and 25% (15%) for parents' highest level of education.⁷ Our regression model includes dummy variables for the missing values of students for each variable. We confirm that our results are robust even when a sample with missing values is excluded from our analysis.

To construct the SES index, we follow the method of Hanushek et al. (2022). Their preferred SES measure is the first principal component of a full vector of dummy variables representing all available household resources and a vector of dummies corresponding to parental education level. Specifically, we use information on six home resources, the number of books (five categories), and parents' educational background (five categories), which are common questions in the TIMSS survey worldwide. Subsequently, we measure the percentile rank of SES within the class using the following formula:

$$SES$$
- $Rank_{ics} = \frac{n_{ics}-1}{N_{Cs}-1}$

where SES- $Rank_{ics}$ is the percentile SES rank of student i in class c of school s; N_{cs} is the number of students in class c of school s; n_{ics} is the ordinal rank of the SES of student i in class c of school s, which increases SES to a maximum of N_{cs} . Figure 1 shows the percentile rank of SES within classes for each SES decile using data from eighth- and fourth-grade students. This figure shows sufficient variation in the SES percentile ranks among students in the same SES deciles for both grades.

⁶ Wößmann (2003) and Fuchs and Wößmann (2007) impute for missing fundamental attribution such as gender, age, grade, parental education level, and the number of books at home by using the mean or median of the smallest available group (class, school, country).

⁷ If the respondent answered "I don't know" for the item of parent's highest level of education, it is treated as a missing value and imputed. Note that parental education is asked of students in the survey of eighth-grade students and parents in that of fourth-grade students. As a result, the proportion of missing values is smaller for fourth-grade students than for eighth-grade students

⁸ Hanushek et al. (2022) calculate four alternative SESs using information from these home environments and parental education levels. We calculate these alternative SESs and find that the results are robust when we use any SES measures.

⁹ Specifically, we use responses to the following six items about household possession: (1) A computer or tablet of your own; (2) A computer or tablet that is shared with other people at home; (3) Study desk/table for your use; (4) Your own room; (5) Internet connection; (6) Your own mobile phone.

¹⁰ In creating the rank variable for SES, the number of students in a class is assumed to be the number of students who participated in the TIMSS.

To capture the incidence of victimization caused by school bullying, we use student responses to a series of questions about school bullying and the student bullying scale constructed in TIMSS. There are nine questions on bullying victimization for eighth-grade students and eight for fourth-grade students: (1) Made fun of me or called me names, (2) Left me out of their games or activities, (3) Spread lies about me, (4) Stole something from me, (5) Hit or hurt me, (6) Made me do things I didn't want to do, (7) Shared embarrassing information about me, (8) Posted embarrassing things about me online, and (9) Threatened me. ¹¹ For each item, students choose one of four options: 1 "Never," 2 "A few times a year," 3 "Once or twice a month," and 4 "At least once a week." The student bullying scale is standardized to have a mean of 10 and a standard deviation of two across all countries by employing Item Response Theory scaling methods, specifically the Rasch partial credit model (Martin et al., 2016). 12 Note that the more frequent the bullying victimization, the lower the value of the student bullying scale. Information on school absence can be obtained from students' responses to a four-level ordinal scale ranging from "more than once a week" to "never or rarely." We construct a dummy variable indicating absence once or more than once every two weeks as a dependent variable.

Table 2 reports the summary statistics. The summary statistics of our preferred restricted sample are in columns (3) and (6). Columns (1) and (4) show the statistics before dropping samples belonging to schools in which only one class was extracted, and columns (2) and (5) show the statistics before dropping samples belonging to schools where SES is not balanced across classes. In both cases, we confirm that the statistics are similar to those in columns (3) and (6).

Regarding the measure of bullying victimization, the average scores for each item ranged between 1 ("Never") and 2 ("A few times a year"). For some items, such as "Left me out of their games or activities," "Hit or hurt me," and "Threatened me," fourth-grade students were more likely than eighth-grade students to be victims of bullying. As for absences, in both grades, the percentage of students who were absent more than once every two weeks was around 15%. These two measures of students' risky behaviors are correlated with each other. Table A2 shows the correlation coefficients for SES, the Bullying Scale, and dummy variable for absence. The Bullying Scale and dummy variable for absence are negatively correlated, suggesting that

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¹¹ Note that item (8) was not asked for fourth-grade students.

¹² Regarding the magnitude of the student bullying scale, considering the TIMSS in 2015, a score of 9.3 or higher means that the students are almost never bullied, a score of 7.3 or lower means that the students experience such behaviors approximately every week, and a score between 7.3 and 9.3 means that the students are bullied approximately every month. Therefore, an increase of one standard deviation, an increase of two points on the student bullying scale, means that the students who experience bullying approximately every week will almost never experience them. See Martin et al. (2016) for details.

students who are frequently victimized by bullying are frequently absent from school.

To investigate potential channels of the rank effects on school bullying and absence, we use the students' subjective responses to questions about their sense of school belonging as other outcome variables. This set of questions consists of seven items for both grades: (1) I like being in school, (2) I feel safe when I am at school, (3) I feel like I belong at this school, (4) I like to see my classmates at school, (5) Teachers at my school are fair to me, (6) I am proud to go to this school, and (7) I learn a lot in school. For each question, students choose one of four options: 1 "Disagree a lot," 2 "Disagree a little," 3 "Agree a little," and 4 "Agree a lot." For all these questions, the average response is between 3 ("Agree a little") and 4 ("Agree a lot").

4. Results

This section presents the estimation results obtained using Eq. (1). First, we explain the main results of the effects of SES rank on bullying victimization and school absence. We then discuss the robustness of the main findings.

4.1 The correlation between SES and the incidence of school bullying and school absence

First, we estimate the correlation between SES and the incidence of school bullying and school absence. Table A3 shows the results. As shown in columns (1) and (3), the coefficient of SES on the bullying scale is positive and statistically significant at the 1% level for both grades, suggesting that students with higher SES are less likely to be bullied. In addition, as shown in columns (2) and (4), the coefficient of SES on school absence is negative and statistically significant at the 1% level for both grades, suggesting that students with higher SES are less likely to be absent from school.

4.2 The SES rank effect on school bullying and school absence

Next, we employ regression model (1) to examine the effect of SES rank within the class on bullying victimization and absenteeism. Table 3 shows the results for both grades. We find that the coefficient of the SES rank of eighth-grade students on the bullying scale is negative and statistically significant at the 5% level, as shown in column (1). This result suggests that eighth-grade students with higher SES ranks within the class are *more* likely to be victims of school bullying. Furthermore, we find that the coefficient of SES rank on school absence is positive and statistically significant at the 1% level, as shown in column (2). This result suggests that students with higher ranks within the class are *more* likely to be absent from school. On the other hand, for fourth-grade students, the effect of SES rank on bullying victimization is much smaller than that of eighth-grade students and not statistically significant (see column (3)). In

addition, the effect on school absence is somewhat positive (see column (4)), but not statistically significant.

In Table A4, we report the estimated effects of SES rank on each bullying victimization variable used to construct the bullying scale. In general, the results are consistent with those presented in Table 3. For example, regarding the eighth-grade results, we find that SES rank has an adverse effect on all nine bullying victimization items, seven of which are statistically significant at the 5% significance level. Only the outcomes of "Left me out of their games or activities" and "Stole something from me" are insignificantly affected. On the other hand, in the fourth-grade results, only two of the eight bullying victimization items (i.e., "Made fun of me or called me names" and "Spread lies about me") are statistically significantly affected by SES rank at the 5% significance level. These findings indicate that the adverse effect of SES rank on school bullying is weaker for fourth-grade students than for eighth-graders, but that SES rank has an adverse effect on less violent bullying even for fourth graders.

Overall, our findings suggest that middle school students with higher SES ranks within the class are more likely to be victims of school bullying or to be absent from school. However, this is only weakly observed in the case elementary school students. These findings may be because elementary school students in friendships rarely recognize SES rank, but become more aware of it as they age. ¹³

4.3. Robustness: Impact of sample selection

To exploit the variation of within-class SES ranks generated by the almost-random assignment of students to classes, we restrict our sample to students in schools with no statistically significant differences in mean SES across classes. Owing to this restriction, slightly less than 70% of the sample in the raw data is dropped from our analysis (See Table A1). To check the effect of sample selection, we also add samples belonging to schools whose SES means are not balanced across classes and estimate model (1) (see columns (2) and (5) in Table A5 for the results). Furthermore, we add the sample belonging to schools from which only one class was extracted in the TIMSS, and estimate (see columns (3) and (6) of Table A5 for the results).

Columns (1) and (4) in Table A5 are reprints of the results in Table 3. The results in columns (2) and (5) of Table A5 show that the adverse effects of SES rank are larger when adding samples belonging to schools where the mean SES is not balanced across classes. The results suggest that in schools where classes are organized by SES, students assigned to classes

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¹³ Similar results are obtained when we use the variable "once or more than once in a week" instead of "once or more than once in every two weeks" as the variable for school absence.

with lower average SES are more likely to be bullied and absent from school. This result could be explained by the higher frequency of bullying victimization in lower SES classes, reinforcing the adverse effects of SES rank. Conversely, students assigned to classes with higher average SES, where the frequency of bullying victimization is lower, are less likely to be victimized by bullies and less likely to be absent from school. That is, in schools where classes are organized by SES, relatively high-SES students in classes with low SES averages would be most strongly affected by the adverse effects of SES rank. Furthermore, the results in columns (3) and (6) of Table A5 show that the absolute value of the coefficient of SES rank is smaller than that in columns (2) and (5) of Table A5 when students belonging to schools where only one class is extracted are added. ¹⁴

In summary, while the results vary somewhat with sample selection and, in particular, the adverse effect of SES rank is enhanced when the mean SES is not balanced across classes, we can conclude that the results obtained in Table 3 are robust.

4.4. Other related outcomes

The above results indicate that students with relatively higher SES suffered from victimization due to school bullying and school absence. To further understand why SES rank in class adversely affects these risky behaviors, we examine the effects of SES rank on other outcomes related to bullying victimization and absenteeism. Good academic performance and good relationships with classmates and teachers seem to be essential for students to develop an attachment to school and a willingness to attend school every day (OECD, 2019). As discussed in the data section, the TIMSS 2015 asked students about their sense of school belonging. We estimate the SES rank effects on these variables regarding students' sense of school belonging. This array of questions has seven items (i.e., "Like being in school"; "Safe at school"; "Belong to school"; "Like to see classmates"; "Fair teachers"; "Proud of school"; and "Learn a lot in school").

The results are reported in Table A6, columns (1) through (7). The table shows that all seven items for eighth-grade students have an adverse SES rank effect at the 5% significance level. Students with higher SES ranks do not think that the school is safe, that teachers are fair to them, and that they learn a lot at school. These SES rank-related changes can be attributed to absenteeism. However, none of the seven items in the fourth-grade results shows an adverse effect of SES rank at the 5% significance level. These results are consistent with the weak SES

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¹⁴ The reason for the results in columns (3) and (6) of Table A4 is not apparent, but there may be some systematic differences; for example, schools with one class extracted are less likely to organize their classes according to SES rank.

rank effect for primary school students, as presented in Table 3.

4.5. Heterogeneity

Finally, we examine how the SES rank effect varies by student gender, national income inequality, and national economic level. Confirming differences in the effect of SES ranks by individual characteristics helps to find and support students at high risk of these risky behaviors. It would also be interesting to ascertain whether the effect of SES ranks differs depending on the country's economic situation. Income inequality is known to be associated with school bullying worldwide (e.g., Elger et al., 2009, 2019). This may be because countries with greater income inequality are more likely to experience relative deprivation, which may strengthen the adverse effects of SES rank. In addition, more affluent countries are likely to have more resources to spend on schools and education, which may help mitigate the effects of the SES rank effect by providing more support to alleviate bullying victimization and absenteeism. We estimate regression Eq. (1) with interaction terms between SES rank and the corresponding variables.

Table 4 presents the estimation results with the interaction terms. In eighth grade, the effect of SES rank on bullying victimization differs between boys and girls. The adverse effect of SES rank on bullying victimization is larger for girls and statistically no different from zero for boys (see column (1) of Table 4). For absences, on the other hand, the effect of SES rank does not differ by gender (see column (4) of Table 4).

We also find evidence that the effect of SES rank varies with national income inequality. Column (2) of Table 4 shows that SES rank has a smaller adverse effect on bullying victimization for students from countries with GINI coefficients of 0.3-0.4 than for those with GINI coefficients of 0.4 or higher (countries with high-income inequality). Furthermore, according to column (5) of Table 4, SES rank has a smaller adverse effect on absenteeism for students from countries with GINI coefficients below 0.3 than those with GINI coefficients above 0.4. Moreover, even in the fourth-grade results, according to column (11) of Table 4, students from countries with GINI coefficients of 0.3-0.4 and less than 0.3 have a smaller adverse effect of SES rank on absenteeism than students from countries with GINI coefficients of 0.4 or more. These results suggest that countries with greater income inequality have a larger adverse effect on SES rank. When we check the heterogeneity of the SES rank effect by country economic level, we find no difference in the effect of SES rank depending on whether a country is an OECD member or a non-OECD member.

5. Conclusion

This study examined the effects of SES rank on bullying victimization and absenteeism among middle and elementary school students worldwide using TIMSS data. We exploited the variation in SES rank within a class generated by the almost random assignment of students to classes within a school by restricting our sample to schools with multiple classes, where the average class SES is balanced across classes. We found that eighth-grade students with a high SES rank within the class were more likely to be victims of school bullying and be absent from school. The results are noteworthy because they contrast with the negative association between absolute SES and the incidence of bullying and absenteeism. Our results highlight the importance of not only the absolute SES but also the relative SES of students' households in class as determinants of bullying victimization and absenteeism.

Interestingly, our results contrast with the findings of Napoletano et al. (2016), who show that the relative affluence of Canadian youth is negatively correlated with bullying victimization. Use of different datasets may be a potential reason of the discrepancy. As our heterogeneity analysis revealed, the effects of SES rank vary by a country's economic situation. In particular, we found that the adverse effects of SES rank are reinforced in countries with high income inequality. Similarly, as our robustness check revealed, the adverse effects of SES rank are strong in schools where the SES mean is not balanced across classes. These results suggest that societies with greater inequality reinforce a sense of inequity, stress, and feelings of inferiority due to differences in social status (Wilkinson and Pickett, 2009). In addition, we found weak effects of SES rank on fourth-grade students. This "insignificant" result is intuitive because elementary school students in friendships rarely recognize SES rank, but they become more aware of it as they age.

Our findings on the effects of SES rank on students' sense of school belonging provide suggestive evidence for a potential mechanism of the adverse effect on bullying victimization and absenteeism. We found that as SES rank increases, students dislike school more, feel unsafe, have a lower sense of school belonging, and are less likely to want to see their friends. Moreover, we found that students with higher SES ranks feel they are treated unfairly by their teachers and recognize that they do not learn much at school. These results suggest that students with high SES feel uncomfortable with the learning environment in schools. Although our analysis is silent about causality, the sense of school belonging is negatively correlated with the incidence of school bullying. In addition, we confirmed that students who are more likely to be victims of bullying are more likely to be absent from school. Similar to previous studies that have reported that bullying victimization can be a factor in avoiding school (e.g., Hutzell and Payne, 2012, 2018), our results indicate that school bullying is a determinant of school absence.

Our findings have ample implications for educational policy. First, whereas most literature examining the relationship between SES and student risky behavior focuses on the absolute level of SES, our study provides evidence of the effect of SES rank. Second, we provide evidence of heterogeneity in the effects of SES rank by student grade, gender, and income inequality in the country. These findings provide a new perspective on educational policies and interventions to mitigate students' risky behaviors. While students with low absolute levels of SES may have been the focus of support in the past, our results highlight the need for policies and interventions for different targets.

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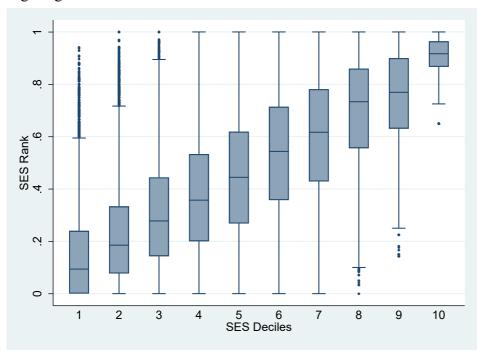
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Figure 1. Box plot of SES rank

Eighth-grade students



Fourth-grade students

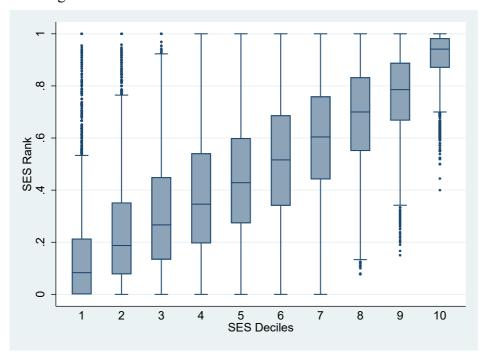


Table 1. Verification of the randomness of student class placement using Monte Carlo simulation.

	Eighth-grade student			Fourth-grade student		
		(1)			(2)	
	Mean	S.D.	Obs.	Mean	S.D.	Obs.
Panel A (Samples where the observed						
SES is balanced across classes)						
Reject	0.038	0.191	89,042	0.027	0.163	97,601
Panel B (Samples where the observed						
SES is not balanced across classes)						
Reject	0.377	0.485	21,726	0.324	0.468	21,207

Notes: This table shows the results of examining whether students are randomly assigned to classes. First, using Monte Carlo simulation, each student is randomly assigned to a class within a school, and the percentile rank of SES within that class is calculated. This process is repeated 1,000 times. Then, a 95% confidence interval is constructed under the null hypothesis that the class assignment of students is random. Next, we check whether the observed SES ranks lie outside the confidence interval. The variable "Reject" is a dummy variable that takes one if the null hypothesis is rejected. Panel A shows the results for samples where the observed SES is balanced across classes at the 5% significance level, while Panel B shows the results for samples where it is not.

Table 2. Summary statistics.

				Eighth-	grade	student							Fourth-	grade s	student			
		(1)			(2)			(3)			(4)			(5)			(6)	
	Mean	S.D.	Obs.	Mean	S.D.	Obs.	Mean	S.D.	Obs.	Mean	S.D.	Obs.	Mean	S.D.	Obs.	Mean	S.D.	Obs.
Dependent variable:																		
Bullying Scale	10.13	1.98	244,994	10.10	1.96	110,768	10.10	1.96	89,042	10.10	1.96	227,764	10.10	1.99	118,808	10.10	1.99	97,601
Made fun of me or called me names	1.97	1.13	244,994	1.98	1.11	110,768	1.99	1.12	89,042	1.92	1.14	227,764	1.90	1.12	118,808	1.90	1.12	97,601
Left me out of their games or activities	1.55	0.94	244,994	1.57	0.92	110,768	1.57	0.92	89,042	1.80	1.09	227,764	1.80	1.08	118,808	1.80	1.08	97,601
Spread lies about me	1.67	0.95	244,994	1.66	0.93	110,768	1.67	0.93	89,042	1.73	1.05	227,764	1.73	1.04	118,808	1.73	1.04	97,601
Stole something from me	1.49	0.88	244,994	1.45	0.83	110,768	1.46	0.83	89,042	1.46	0.88	227,764	1.46	0.88	118,808	1.46	0.88	97,601
Hit or hurt me	1.44	0.85	244,994	1.44	0.83	110,768	1.44	0.84	89,042	1.73	1.03	227,764	1.73	1.03	118,808	1.74	1.03	97,601
Made me do things I didn't want to do	1.31	0.73	244,994	1.30	0.70	110,768	1.31	0.71	89,042	1.44	0.88	227,764	1.44	0.87	118,808	1.44	0.87	97,601
Shared embarrassing information about me	1.44	0.82	244,994	1.44	0.81	110,768	1.45	0.81	89,042	1.53	0.93	227,764	1.53	0.93	118,808	1.53	0.93	97,601
Posted embarrassing things about me online	1.18	0.58	244,994	1.19	0.57	110,768	1.19	0.57	89,042									
Threatened me	1.24	0.67	244,994	1.24	0.65	110,768	1.24	0.65	89,042	1.43	0.88	227,764	1.43	0.88	118,808	1.44	0.88	97,601
Absence_Once every two weeks or more	0.15	0.36	244,994	0.15	0.36	110,768	0.15	0.36	89,042	0.14	0.35	227,764	0.14	0.35	118,808	0.14	0.35	97,601
Like being in school	3.13	0.85	244,068	3.04	0.86	110,433	3.03	0.87	88,770	3.33	0.86	226,786	3.32	0.86	118,308	3.31	0.86	97,182
Safe at school	3.28	0.82	243,603	3.25	0.82	110,248	3.25	0.82	88,606	3.49	0.77	226,261	3.48	0.77	118,001	3.48	0.77	96,939
Belong at school	3.23	0.87	241,560	3.15	0.89	109,266	3.15	0.89	87,844	3.47	0.83	225,050	3.46	0.83	117,352	3.46	0.84	96,423
Like to see classmates	3.62	0.67	243,005	3.59	0.69	109,937	3.59	0.70	88,349	3.79	0.53	226,177	3.81	0.51	117,978	3.81	0.51	96,919
Fair teachers	3.22	0.85	242,835	3.18	0.84	109,940	3.17	0.85	88,364	3.57	0.74	225,816	3.58	0.72	117,824	3.57	0.73	96,784
Proud to go to this school	3.23	0.89	243,341	3.20	0.88	110,152	3.20	0.88	88,540	3.55	0.77	226,048	3.57	0.74	117,929	3.56	0.74	96,867
Learn a lot in school	3.44	0.73	243,938	3.37	0.75	110,336	3.37	0.75	88,686	3.75	0.55	226,666	3.75	0.54	118,256	3.75	0.54	97,138
Independent variable of interest:																		
SES Rank	0.50	0.30	244,994	0.50	0.30	110,768	0.50	0.30	89,042	0.50	0.30	227,764	0.50	0.30	118,808	0.50	0.30	97,601
Independent variable of others:																		
SES	0.01	1.60	244,994	0.45	1.24	110,768	0.48	1.21	89,042	0.07	1.43	227,764	0.31	1.27	118,808	0.32	1.27	97,601
Girl	0.50	0.50	244,994	0.50	0.50	110,768	0.50	0.50	89,042	0.50	0.50	227,764	0.50	0.50	118,808	0.50	0.50	97,601
Age	14.2	0.73	244,994	14.1	0.61	110,768	14.1	0.62	89,042	10.2	0.58	227,764	10.2	0.58	118,808	10.2	0.58	97,601
Gini>0.4	0.29	0.45	205,392	0.36	0.48	88,612	0.36	0.48	70,566	0.16	0.37	216,160	0.10	0.30	113,041	0.08	0.27	92,845
Gini0.3-0.4	0.57	0.50	205,392	0.43	0.49	88,612	0.44	0.50	70,566	0.60	0.49	216,160	0.59	0.49	113,041	0.61	0.49	92,845
Gini<0.3	0.14	0.35	205,392	0.22	0.41	88,612	0.20	0.40	70,566	0.24	0.42	216,160	0.31	0.46	113,041	0.30	0.46	92,845
OECD	0.42	0.49	244,994	0.45	0.50	110,768	0.45	0.50	89,042	0.31	0.46	227,764	0.25	0.43	118,808	0.25	0.43	97,601
(Fourth grade only) Drop if the sample of countries that		_						_			Yes			Yes			Yes	
have not conducted questionnaires for parents.											1 03			1 03			103	
Drop if responses to questions about bullying		Yes			Yes			Yes			Yes			Yes			Yes	
victimization and absenteeism are missing.		108			1 08			1 03			1 05			1 08			1 03	
Drop if samples are in classes with less than ten		Yes			Yes			Yes			Yes			Yes			Yes	
Drop if samples belong to schools in which only one class was extracted.		No			Yes			Yes			No			Yes			Yes	
Drop if samples belong to schools where SES is not balanced across classes.		No			No			Yes			No			No			Yes	

Notes: Sample: Students of participating countries in the TIMSS 2015 survey. SES rank: Percentile rank of SES within the class. All tabulations give the same weight to each country.

Table 3. The SES rank effect on school bullying and school absence.

	Eighth-gra	ide student	Fourth-gra	ade student	
		Absence_Once		Absence_Once	
	Bullying Scale	every two weeks	Bullying Scale	every two weeks	
_		or more		or more	
	(1)	(2)	(3)	(4)	
SES Rank	-0.1757**	0.0499***	-0.1070	-0.0056	
	(0.0726)	(0.0125)	(0.0654)	(0.0123)	
SES	Yes	Yes	Yes	Yes	
School fixed effects	Yes	Yes	Yes	Yes	
Student controls	Yes	Yes	Yes	Yes	
Observations	89,042	89,042	97,601	97,601	
R-squared	0.156	0.117	0.173	0.098	

Notes: Sample: Students from schools with more than two classes were extracted in the TIMSS 2015 survey and balanced SES averages across classes. The Bullying Scale indicates that lower values mean more bullying victimization. All specifications control a quartic polynomial of students' SES, students' gender, age, dummy variables for missing values, and school-fixed effects and give the same weight to each country. Standard errors are clustered-robust at the class level and reported in parentheses. * p < 0.10. ** p < 0.05. *** p < 0.01.

Table 4. Heterogeneous effects.

			Eighth-gr	ade student		
		ullying Scale		Absence_O	nce every tw	o weeks or
		unying Scale			more	
	(1)	(2)	(3)	(4)	(5)	(6)
SES Rank * Girl	-0.1218**			0.0031		
	(0.0484)			(0.0094)		
SES Rank * Gini0.3-0.4		0.1813***			-0.0131	
		(0.0625)			(0.0123)	
SES Rank * Gini<0.3		0.0217			-0.0365**	
		(0.0841)			(0.0149)	
SES Rank * OECD		()	-0.0513		()	-0.0086
SES Raik OECD			(0.0525)			(0.0097)
~~~ ~ .		0.00000000	` ′	0.0404444	0.0==0.04.4	` ′
SES Rank	-0.1107	-0.3266***	-0.1469*	0.0491***	0.0758***	0.0547***
	(0.0771)	(0.0920)	(0.0776)	(0.0137)	(0.0167)	(0.0140)
SES	Yes	Yes	Yes	Yes	Yes	Yes
School fixed effects	Yes	Yes	Yes	Yes	Yes	Yes
Student controls	Yes	Yes	Yes	Yes	Yes	Yes
Observations	89,042	70,566	89,042	89,042	70,566	89,042
R-squared	0.156	0.155	0.156	0.117	0.124	0.117
			Fourth-gr	ade student		
	В	ullying Scale		Absence_O	nce every tw	o weeks or
			(0)	(10)	more	(10)
GEG D 1 * G'1	(7)	(8)	(9)	(10)	(11)	(12)
SES Rank * Girl	-0.0407			-0.0050		
	(0.0449)			(0.0088)		
SES Rank * Gini0.3-0.4		0.1267			-0.0448***	
		(0.0919)			(0.0168)	
SES Rank * Gini<0.3		0.0279			-0.0412**	
		(0.0961)			(0.0173)	
SES Rank * OECD			0.0154			-0.0013
			(0.0507)			(0.0091)
SES Rank	-0.0858	-0.2122**	-0.1120*	-0.0030	0.0348*	-0.0052
SES Kalik						
SES	(0.0695) Yes	(0.1070) Yes	(0.0678) Yes	(0.0130) Yes	(0.0193) Yes	(0.0128)
School fixed effects						Yes
	Yes	Yes	Yes	Yes	Yes	Yes
Student controls	Yes	Yes	Yes	Yes	Yes	Yes
Observations	97,601	92,845	97,601	97,601	92,845	97,601
R-squared	0.173	0.165	0.173	0.098	0.098	0.098

Notes: Sample: Students from schools with more than two classes were extracted in the TIMSS 2015 survey and balanced SES averages across classes. The Bullying Scale indicates that lower values mean more bullying victimization. All specifications control a quartic polynomial of students' SES, students' gender, age, dummy variables for missing values, and school-fixed effects and give the same weight to each country. Standard errors are clustered-robust at the class level and reported in parentheses. * p < 0.10. *** p < 0.05. **** p < 0.01.

Table A1. Sample selection.

	Eighth-grade student	Fourth-grade student
Baseline	285,190	286,189
(1) (Fourth-grade only) Drop if the sample of		
countries that have not conducted		272,154
questionnaires for parents.		
(2) Drop if responses to questions about		
bullying victimization and absenteeism are	249,793	241,375
missing.		
(3) Drop if samples are in classes with less	244,994	227,764
than ten students.	244,994	227,704
(4) Drop if samples belong to schools in which	110,768	118,808
only one class was extracted.	110,700	110,000
(5) Drop if samples belong to schools where	89,042	97,601
SES is not balanced across classes.	69,042	97,001

Note: Sample: TIMSS 2015.

Table A2. Correlation coefficients among SES, Bullying Scale, and the dummy variable for absence.

	E	Eighth-grade student			Fourth-grade student			
-	SES	Bullying Absence_Once Scale every two weeks or more		SES	Bullying Scale	Absence_Once every two weeks or more		
SES	1.00			1.00				
Bullying Scale	0.07	1.00		0.05	1.00			
Absence_Once every two weeks or more	-0.14	-0.07	1.00	-0.15	-0.08	1.00		

Note: The number of observations is 89,042 in Eighth-grade and 97,601 in Fourth-grade.

Table A3. Relationship between SES and bullying victimization and absences.

	Eighth-grad	de student	Fourth-gra	Fourth-grade student			
		Absence_Once		Absence_Once			
	Bullying Scale	every two	Bullying Scale	every two			
		weeks or more		weeks or more			
	(1)	(2)	(3)	(4)			
SES	0.0293***	-0.0125***	0.0198***	-0.0256***			
	(0.0087)	(0.0016)	(0.0071)	(0.0015)			
School fixed effects	Yes	Yes	Yes	Yes			
Student controls	Yes	Yes	Yes	Yes			
Observations	89,042	89,042	97,601	97,601			
R-squared	0.155	0.117	0.173	0.098			

Notes: Sample: Students from schools with more than two classes were extracted in the TIMSS 2015 survey and balanced SES averages across classes. The Bullying Scale indicates that lower values mean more bullying victimization. All specifications control students' gender, age, dummy variables for missing values, and school-fixed effects and give the same weight to each country. Standard errors are clustered-robust at the class level and reported in parentheses. * p < 0.10. *** p < 0.05. *** p < 0.01.

Table A4. The SES rank effect on individual bullying victimization items.

				Eig	hth-grade stud	lent			
	Made fun of me or called me names	Left me out of their games or activities	Spread lies about me	Stole something from me	Hit or hurt me	Made me do things I didn't want to do	Shared embarrassing information about me	Posted embarrassing things about me online	Threatened me
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
SES Rank	0.0911**	0.0411	0.1019***	0.0217	0.0958***	0.0515**	0.0728**	0.0676***	0.0835***
	(0.0406)	(0.0342)	(0.0358)	(0.0295)	(0.0316)	(0.0256)	(0.0306)	(0.0224)	(0.0248)
SES	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
School fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Student controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	89,042	89,042	89,042	89,042	89,042	89,042	89,042	89,042	89,042
R-squared	0.155	0.086	0.085	0.161	0.115	0.084	0.076	0.072	0.099
				Fou	ırth-grade stud	lent			
	Made fun of me or called me names	Left me out of their games or activities	Spread lies about me	Stole something from me	Hit or hurt me	Made me do things I didn't want to do	Shared embarrassing information about me	Threatened me	
	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	
SES Rank	0.0944**	-0.0135	0.1242***	0.0032	0.0413	0.0307	0.0044	0.0493*	
	(0.0374)	(0.0361)	(0.0346)	(0.0284)	(0.0332)	(0.0288)	(0.0304)	(0.0295)	
SES	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
School fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Student controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Observations	97,601	97,601	97,601	97,601	97,601	97,601	97,601	97,601	
R-squared	0.116	0.098	0.098	0.155	0.123	0.101	0.101	0.123	

Notes: Sample: Students from schools with more than two classes were extracted in the TIMSS 2015 survey and balanced SES averages across classes. All specifications control a quartic polynomial of students' SES, students' gender, age, dummy variables for missing values, and school-fixed effects and give the same weight to each country. Standard errors are clustered-robust at the class level and reported in parentheses. * p < 0.10. ** p < 0.05. *** p < 0.01.

Table A5. Robustness check: Impact of sample selection.

	Eigl	nth-grade stud	dent	Fourth-grade student			
	(1)	(2)	(3)	(4)	(5)	(6)	
Bullying Scale	-0.1757**	-0.3137***	-0.2565***	-0.1070	-0.1512***	-0.0876**	
	(0.0726)	(0.0572)	(0.0442)	(0.0654)	(0.0533)	(0.0438)	
Absence_Once every	0.0499***	0.0769***	0.0606***	-0.0056	0.0140	0.0124	
two weeks or more	(0.0125)	(0.0103)	(0.0085)	(0.0123)	(0.0097)	(0.0080)	
Observations	89,042	110,768	244,994	97,601	118,808	227,764	
Drop if samples belong to							
schools in which only one	Yes	Yes	No	Yes	Yes	No	
class was extracted.							
Drop if samples belong to							
schools where SES is not	Yes	No	No	Yes	No	No	
balanced across classes.							

Notes: Sample: TIMSS 2015. The Bullying Scale indicates that lower values mean more bullying victimization. All specifications control a quartic polynomial of students' SES, students' gender, age, dummy variables for missing values, and school-fixed effects and give the same weight to each country. Standard errors are clustered-robust at the class level and reported in parentheses. * p < 0.10. ** p < 0.05. *** p < 0.01.

Table A6. The SES rank effect on related outcomes.

		Eighth-grade student										
	Like being in school	Sate at school B		Sate at school Belong at sch		Like to see classmates	Fair teachers	Proud to go to this school	Learn a lot in school			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)					
SES Rank	-0.0658**	-0.1114***	-0.1076***	-0.0711***	-0.0736**	-0.0648**	-0.1124***					
	(0.0321)	(0.0291)	(0.0342)	(0.0254)	(0.0315)	(0.0322)	(0.0279)					
SES	Yes	Yes	Yes	Yes	Yes	Yes	Yes					
School fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes					
Student controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes					
Observations	88,770	88,606	87,844	88,349	88,364	88,540	88,686					
R-squared	0.182	0.184	0.131	0.211	0.157	0.187	0.197					

Fourth-grade student Like being in Like to see Proud to go to Learn a lot in Fair teachers Safe at school Belong at school classmates this school school school (8) (9) (10)(12)(13)(14)(11)SES Rank 0.0523* -0.0270 -0.0320 -0.0312* -0.0191 -0.0070 -0.0124 (0.0275)(0.0248)(0.0287)(0.0165)(0.0235)(0.0248)(0.0174)SES Yes Yes Yes Yes Yes Yes Yes School fixed effects Yes Yes Yes Yes Yes Yes Yes Student controls Yes Yes Yes Yes Yes Yes Yes Observations 97.182 96,939 96,423 96,919 96,784 96,867 97,138 R-squared 0.191 0.115 0.122 0.073 0.131 0.132 0.090

*Notes:* Sample: Students from schools with more than two classes were extracted in the TIMSS 2015 survey and balanced SES averages across classes. All specifications control a quartic polynomial of students' SES, students' gender, age, dummy variables for missing values, and school-fixed effects and give the same weight to each country. Standard errors are clustered-robust at the class level and reported in parentheses. * p < 0.10. ** p < 0.05. *** p < 0.01.