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New Evidence on Class Size Effects : A Pupil Fixed Effects Approach

by
Nadir Altinok and Geeta Kingdon

Abstract

The impact of class size on student achievement remains a thorny question for educational decision makers. Meta-analyses of empirical studies emphasise the absence of class-size effects but detractors have argued against such pessimistic conclusions because many of the underlying studies have not paid attention to the endogeneity of class-size. This paper uses a stringent method to address the endogeneity problem using TIMSS data on 45 countries. We measure the class size effect by relating the difference in a student's achievement across subjects to the difference in his/her class-size across subjects. This (subject-differenced) within-pupil achievement production function avoids the problem of the non-random matching of children to specific schools, and to classes within schools. The results show a statistically significant effect of class size for 16 countries but in only 10 of them is the effect negative, and the effect size is very small in most cases. Several robustness tests are carried out, including control for students' subject-specific ability and subject-specific teacher characteristics, and correction for possible measurement error. Thus, our stringent approach to addressing the problem endogeneity confirms the findings of meta-analyses that find little support for class size effects. We find that class-size effects are smaller in resource-rich countries than in developing countries, supporting the idea that the adverse effect of larger classes increases with class-size. We also find that class size effects are smaller in regions with higher teacher quality.

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Author contacts

*Corresponding author, Nadir Altinok, BETA (Bureau d'Economie Théorique et Appliquée), 61 avenue de la Forêt Noire, 67085 Strasbourg Cedex, France. Tel: +33 387 54 71 38

Fax: +33 390 24 20 70. Email: nadir.altinok@gmail.com

Geeta Kingdon, Institute of Education, University of London, 36 Gordon Square, London, WC1H 0PD. Tel: 020 3073 8316. Email g.kingdon@ioe.ac.uk

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1. Introduction

Increasing teacher inputs is popularly thought to be an effective way of increasing student learning and teachers are mostly in favour of smaller classes. However, reducing class size is usually an expensive policy. In most developed countries, demographic changes have mechanically involved important reductions in class size. This evolution strongly contributes to the increase of the cost per pupil over a long period because teacher salaries consume the dominant share of recurring education expenses, estimated between 66% and more than 90% in OECD countries.

The key empirical challenge in identifying class size effects arises due to the potential non-random matching of students to schools and, within schools, to particular classes. For instance, if higher ability students are systematically placed in the larger classes within their grade in the school, a positive coefficient for class size can appear in a standard estimate of the education production function but this would not represent a causal effect. In order to circumvent this kind of endogeneity bias two major approaches have been used in the literature.

The first uses randomized experiments. The Project STAR (Student/Teacher Achievement Ratio) carried out in Tennessee in the mid-1980s is probably the most important study of this kind. Krueger (1999) found that the advantage of experimentally reducing class size were relatively small: a reduction of around 8 students per class raised test score by less than 0.02 standard-deviations, making class size reduction a very costly way of achieving small learning gains. An important limitation of randomized trials is that participants may react to being in an experiment. Hoxby (2000) argues that teachers may exhibit behavioural change in an experiment because the outcome of the experiment has implications for the future funding of their schools. As Todd and Wolpin (2003) point out, experiment based studies have a very high cost and new experimental studies are necessary in order to check for the robustness of the results found in Project STAR. Other studies have tried the same kind of experiment and Kremer (2003) provides a synthesis of these analyses in developing countries. Yet other studies use other exogenous

variations in class-size given by ‘natural’ experiments. For instance, Hoxby (2000) uses the fact that large rivers make it difficult for parents to access schools. These natural restrictions make it possible to purge endogeneity bias in a rather intuitive way. However, truly natural ‘natural’ experiments are few (Rosenzweig and Wolpin, 2000).

Another approach to dealing with endogeneity bias is using valid instruments which induce exogenous variation in class size. If class size varies due to some exogenously given administrative rules (rather than due to student or school choices), the measurement of the class size effect can be free of endogeneity bias. The solution consists of seeking a variable correlated with class size but not otherwise correlated with pupil achievement. Angrist and Lavy (1999) exploit a rule of the Israeli school system whereby the mandated maximum class size is fixed at 40 pupils. Consequently, when number of pupils in a grade increases from 40 to 41 pupils, the grade is split into two classes, with mean class size of 20.5, and the change in class size is not due to the non-random choice of pupils or school. Angrist and Lavy apply this principle and obtain a significant effect from a reduction in class size on student achievement. However, even when an exogenously given (e.g. government-mandated) maximum class-size rule exists in a developing country, it is rarely adhered to in practice. Thus a discontinuity design of the type used by Angrist and Lavy may not yield valid instruments for class size in developing country contexts.

Within the instrumental variables (IV) approach, Wößmann and West (2006) estimate the class size effect on the school performance for 11 countries by combining school fixed effects and the IV method, in order to identify the random variation of the class size between two consecutive grades inside the same school. They regress the change in performance of pupils between grades 7 and 8 on the change in class size between grades 7 and 8 in the same school. The authors combine school fixed effects and IV estimation to identify random class-size variation between two adjacent grades within individual schools. It should be noted that they do not use a *pupil fixed effects* approach but rather a *school fixed effects* technique. The authors show

that conventional estimates of class-size effects are strongly skewed by the non-random distribution of pupils into small and large class sizes in schools. The instruments used are the average number of pupils for each grade within the school. While Wößmann and West (2006) find large and significant effects coefficients for Greece and Iceland, they show small or no effects for the other 9 countries.

In this paper, we examine the effect of class size on pupil achievement by using the traditional achievement production function, but the innovation is to allow for pupil fixed effects in cross section data. Across-subject rather than across-time differencing is used. This is possible because the TIMSS database provides a pupil's score in several evaluated subjects for some countries (and in at least mathematics and science for all countries). This approach enables us to control all subject-invariant student and family unobservables. Our cross-section data make it possible to check if the class size in different subjects in a given year is correlated with the student's marks across those subjects within the grade in the school. In other words, we estimate a within-pupil across-subject equation of the achievement production function rather than a within-pupil across-time one. The idea is exactly the same as in panel data estimates of the achievement production function but we will show below that our approach is superior to the panel data approach both methodologically and in terms of cost-effectiveness. Other studies have used this estimation technique in order to evaluate teacher gender effects on pupil achievement (Dee, 2005; Holmlund and Sund, 2005; Ammermüller and Dolton, 2007) and the effect of teacher characteristics on pupil achievement (Kingdon, 2006; Aslam and Kingdon, 2007). To our knowledge, our paper is the first to use this regression technique for examining the effect of class size on educational achievement.

2. Estimation technique

The objective is to estimate an educational production function in a consistent manner. The standard achievement function is specified as follows:

$$A_{ik} = \alpha + \beta X_{ik} + \delta S_k + \mu_i + \eta_k \quad (1)$$

where the achievement level of student i in school k is determined by a vector of his/her personal characteristics (X) and by a vector of school and teacher characteristics (S). μ_i captures all student level and η_k all school level unobservables. Typically in the literature, class size is included in the vector S , i.e. class size variables have a k subscript. However, we have data on student achievement and separately on class-size *by subject* for all students of grade 8 in the sample schools. That is, for each student in grade 8, there are as many rows of data as there are number of subjects. Thus, there can be variation in class size for a student within the school since lessons for each subject are taught in different classes, and it is possible to include class size as an explanatory variable in its own right in a school fixed effects equation. With this approach it is also possible to include pupil fixed effects whereby the only variables retained in the achievement equation will be class size and teacher characteristics since it is only these that vary within pupil (across subject). This is the approach we follow. We estimate a simple pupil fixed effects equation of achievement:

$$A_{ijk} = \alpha + \beta X_{ik} + \gamma C_{jk} + \delta S_k + (\mu_{ij} + \varepsilon_{jk} + \eta_{jk}) \quad (2)$$

A_{ijk} is the achievement of student i in subject j in school k , X is a vector of characteristics of pupil i , C is the class size and teacher characteristics for subject j and S a vector of characteristics of school k . The composite error term is in brackets. μ_{ij} , ε_{jk} and η_{jk} represent respectively the unobserved characteristics of the student, the subject class and the school. A pupil fixed effects model implies, for the simplified case of two subjects, 1 and 2:

$$(A_{i2k} - A_{i1k}) = \beta(C_{2k} - C_{1k}) + \{(\mu_{i2} - \mu_{i1}) + (\varepsilon_{k2} - \varepsilon_{k1}) + (\eta_{k2} - \eta_{k1})\} \quad (3)$$

Pupil fixed effects implies within-school estimation since a student necessarily studies within a single school. If school unobservables are not subject specific (η does not have a j subscript) and if pupil unobservables are also not subject specific (μ does not have a j subscript) then within school k , we have:

$$(A_{i2} - A_{i1}) = \beta(C_2 - C_1) + (\varepsilon_2 - \varepsilon_1) \quad (4)$$

and regressing *difference* in a pupil's test scores across subjects on the *difference* in class size across subjects nets out the effect of all student unobserved characteristics. However, if student ability is subject-varying, it not netted out but $(\mu_{i2} - \mu_{i1})$ remains in the error term. But its presence in the error term alone would not cause omitted variable bias *unless it were correlated with* $(C_2 - C_1)$. But for that to happen, students should be able to match to specific classes of a subject within their grade in the school (e.g. students who are bright in a subject systematically match to the smaller – or the larger – classes of that subject within their grade). In our approach, by construction this is ruled out since each subject row has the *averaged* class size of all classes of that subject in grade 8 in the school. Thus, the presence of subject-varying pupil ability is not a source of bias in our approach. However, subject-specific *school* unobservables $(\eta_{k2} - \eta_{k1})$ remain in the error term and may in principle be correlated with $(C_2 - C_1)$. However such bias is unlikely to be important since subject varying aspects of school should be captured in class size, i.e. if a school emphasises a subject, it is often reflected in (or is because of) small class sizes in that subject and this is likely to show up in the subject specific class size effect.

However, for consistent estimation of the class size effect, it is also required that *class level* unobserved characteristics be unrelated to the included class size variable:

$$E[(\varepsilon_2 - \varepsilon_1)(C_2 - C_1)] = 0 \quad (5)$$

Since omitted class level variables in ε_1 , ε_2 may be correlated both with class size C_1 , C_2 and with student achievement A_1 , A_2 , we cannot say that pupil fixed effects estimation of

achievement permits us to interpret the effects of class size as causal. While across-subject pupil fixed effects estimation solves one source of endogeneity (the correlation between μ and C), it does not solve the second potential source of endogeneity (the possible correlation between ε and C). This is analogous to the situation with standard panel data analysis where class unobservables remain in the error term.

4. Data

The estimation strategy presented above requires a specific database. Two conditions are necessary. First of all, it is necessary to have students' achievement scores for different subjects. If we have only one subject for each pupil, it is not possible to use the pupil fixed effects estimator. Another condition is that there be reasonable variation in class size between subjects.

A database that answers both these conditions is the TIMSS survey (Trends in International Mathematics and Science Study) which was carried out in 1995, 1999 and 2003. In our study, we use the data from TIMSS 2003. TIMSS was led by the IEA (International Association for the Evaluation of Educational Achievement) and constitutes the largest international survey of pupil achievement, encompassing approximately 50 countries, covering grades 4 and 8 of the educational cycle. Since the number of evaluated subjects varies only in grade 8, we chose this grade for our estimations. Each participating country administered the test with a representative sample of schools with grade 8 pupils. The tests are standardized by the IEA and the results are thus comparable across countries. The pupils were tested in mathematics and the sciences. According to the structure of the education systems, the number of science subjects varied from 2 to more than 8. Typically, there are two types of education systems: the 'unified' system where science is taught as a single subject in grade 8 and the 'diversified' system where the different sciences are taught as separate subjects. For example, we can differentiate England where only 2 scientific subjects are taught at grade 8 level (Mathematics and Science) and

Hungary where there are 5 various scientific subjects (Mathematics, Physics, Biology, Chemistry and Geography).

Of the 45 countries taking part in TIMSS for pupils at grade 8, we retained 33 countries which met the two conditions mentioned above. For 8 countries, class size does not vary between the subjects (Indonesia, Israel, Italy, Japan, Philippines, Singapore, Syria, Taiwan). In some other countries, the variations in class size are too small. In general, when the class size difference between subjects is less than 10% apart – when averaged across the sample schools within a country - we do not retain that country. Four countries are in this situation (Saudi Arabia, Ghana, Japan, Norway). Finally we do not retain Morocco because data is lacking for a large number of schools and because the education system consists of two parts: an "integrated system" (where only one science matter is taught) and a "diversified system" (where several science subjects are taught). Although for some countries, more than 5 subjects are available, we do not have specific achievement scores in all the subjects. This is especially the case in countries of Central Asia and Eastern Europe for which 5 or 6 scientific subjects are taught.

Descriptive statistics are presented in Appendix Tables A1 and A2. In graphs 1 to 4, we present the kernel densities of achievement mark by subject for 4 countries (Australia, US, Bulgaria and Egypt) as an illustration. It is clear that the distribution of achievement scores is not similar across subjects. For example, the science scores in Egypt are higher and less dispersed than scores in mathematics. In order to use the difference of achievement scores between subjects as the dependent variable, it is necessary to standardize the scores. We standardize the score in each subject by the average of the score in the subject in the country, i.e. we use the z-scores of achievement score. The z-score is the score of a pupil in a given subject minus the national average score in that subject, divided by the standard deviation of the national score on this subject. By construction then, the average of the standardized score in each subject is equal to 0

and its standard deviation is 1. The right parts of graphs 1 to 4 show that the distribution of standardized scores is more similar across subjects than the distribution of raw scores.

5. Results

We start by discussing the main results. Then, we perform various robustness tests and finally make a synthesis of the effects obtained.

5.1. Overall results

The results of the regressions are presented in Tables 1 to 7. In order to simplify the presentation, we divide the countries into three categories: (i) Developed Countries, (ii) Eastern European and Central Asian (EECA) countries and (iii) Developing Countries. Only the coefficients relating to the class size variable are presented. Details about the other included variables are given in the notes below each table. To avoid losing observations due to missing values of control variables, we use the dummy variable approach for missing values. This is likely to reduce biases due to omitted variables and also helps to avoid reducing the sample size.

The estimates use pupils' results in all subjects tested in a country. We start with weighted least squares regression. Our estimations are thus weighted by students' sampling weights to ensure that the contribution of the students from each stratum in the sample to the parameter estimates is the same as would have been obtained in a complete census enumeration (DuMouchel and Duncan, 1983 ; Wooldridge, 2001). This method can only explain the differences between pupils across schools and carries the risk that coefficients are biased due to the correlation of class-size with students' unobserved characteristics, such as ability. To correct such endogeneity bias, we first re-estimate the WLS achievement equations on the reduced sample of only those classes which are not grouped by student ability (Table 2) and then attempt to address endogeneity by instrumenting class-size (not presented). In Table 3 we present results

from School fixed effects estimation. Table 4 presents our preferred pupil fixed effects estimates of the achievement equation where identification of the class size effect comes only from across-subject difference in class-size ‘within a pupil’. Since class-size varies sufficiently across subjects in only 33 countries, the pupil fixed effects estimates are confined to this sub-sample of countries. Tables 5 to 8 present some robustness tests and synthesis of results. As we pool data across all subjects, a dummy variable for each subject is included in all the regressions. In addition, a series of pupil and school level variables are introduced as control variables (age, gender, speaks national language at home, number of books at home, possesses a computer, parents born in country, parents' highest education, city size and number of computers in the school). The last variable drops out of the school fixed effects equations and all these controls drop out of the pupil fixed effects equations.

We start with WLS regressions in Table 1. The first column imposes the restriction that class-size has a linear relationship with achievement (class-size is entered linearly). In the second set of columns, we allow the class-size effect to be non-linear by introducing both the linear and quadratic terms of class-size. In the majority of linear estimates, class size seems to have a significant coefficient. For instance, it is statistically significant in 32 out of 45 countries. However, the sign of the class size coefficient differs across country groups. Looking at the first column, class size is mostly positively correlated with pupil achievement in developed countries: in 11 of the 14 countries in this group, there is a positive and significant class-size ‘effect’. For the United States, there is no significant effect. The size of the coefficient varies by country: the effects are strong for Hong-Kong and Netherlands (where a 1 SD increase in class size increases achievement score by 0.25 and 0.22 SD respectively), while they are rather weak for Italy and Taiwan. A positive class size effect can arise if better quality schools attract more students and thus have higher class size, or if schools group low ability students into smaller classes so that high ability students are in larger classes. The second set of columns show that the class size-

achievement relationship is non-linear in 10 out of the 14 developed countries – concave in 5 countries (with an average turning point at 40 students per class) – and convex in others.

The second group of countries (EECA) is analyzed next. Of the 13 countries included in this group, 5 have significant positive class-size coefficients (Estonia, Hungary, Latvia, Moldova and Slovakia) and 4 have significant negative coefficients (Bulgaria, Macedonia, Romania and Russian Federation). The size of the coefficients is very small, except for Hungary and Slovakia. The relationship is quadratic in 6 of the 13 countries. The last group of countries – namely developing countries – demonstrate a mostly negative relation between class size and student performance. The class-size coefficient is negative and statistically significant in 10 out of the 14 countries for which there is a statistically significant class-size ‘effect’. Ghana, Indonesia, and Lebanon have a positive and significant class-size coefficient. Where the relationship is non-linear, it is mostly convex i.e. achievement falls with higher class size but at a decreasing rate. Where the convexity has an upward sloping part, it occurs at very high levels of class-size (on average above 49 students per class, and only 8% of developing country classes have more than 49 students per class). Thus, for most of the relevant range of class-sizes in developing countries, the relationship of class-size with achievement is negative.

Thus it appears that, in general, the class size effect varies according to the economic level of the countries. While it is generally positive in the developed countries, it seems to be mostly negative in developing countries. The effect is mixed for the EECA countries. The fact that the correlation of class-size with pupil achievement across schools is generally negative in developing countries but not so elsewhere accords with the notion that class-size matters negatively when classes are large, since mean class-size is substantially larger in developing countries than in the EECA and developed countries, as seen in Appendix Table A2. However, these are naïve WLS results and, as such suffer from potential endogeneity bias.

We try several different ways of reducing endogeneity bias. Firstly, we can partly address endogeneity concerns by estimating achievement equations only for that subset of classes where students are not grouped by ability. The TIMSS survey asked schools whether students were grouped by ability within their maths classes and within their science classes. Appendix Table A3 shows the proportion of maths and science classes that are grouped by ability in each country, and how sample size changes when we consider only students not grouped by ability. Table 2 presents WLS results for the reduced sample of students not grouped by ability. In the developed country group, the result of removing the effect of student ability in this way is generally to cause the class-size effect (CSE) to become more negative. Comparing column 1 in Tables 1 and 2, the CSE falls in 9 out of 13 countries, though rarely statistically significantly. In four of these 9 (Australia, Italy, Sweden and USA) the coefficient turns from positive to negative; in the case of Sweden it turns from significant positive to significant negative and in the US it turns from insignificant positive to significant negative. Changes are apparent in the quadratic specification as well, e.g. for England where the sign of the linear term turns negative. In both the EECA group and the developing country groups, the result of removing ability-set classes is more mixed and generally more inconsequential¹.

A second way we attempted to address the endogeneity of class size, is by using Instrumental Variable (IV) estimation. The TIMSS survey asked teachers to what extent high student teacher ratio limited how they taught. This is highly correlated with class size but perhaps should not otherwise be correlated to student achievement. If we believe in the instrument's validity, this correction for endogeneity makes quite a lot of difference. Appendix A4 presents the results. Whereas Table 1 showed that in the developed country group, the CSE was positive and significant in 10 out of the 14 countries, IV estimates show the CSE to be

¹ The CSE falls in 4 and rises in 10 out of the 14 EECA countries, though never statistically significantly. In Latvia it changes from significant to insignificant positive, in Macedonia from significant to insignificant negative and in the Russian Federation from insignificant to significant negative. The CSE falls in 11 and rises in 6 out of the 17 developing countries, but never significantly. In Malaysia and Tunisia, CSE turns negative, in Syria the significant negative CSE turns insignificant.

negative and significant in 5 of these 14 countries. In Belgium, New Zealand and Taiwan the significant positive CSE turns to a significant negative effect and the same also happens in Indonesia and Lebanon in the developing country group, though there are less substantial changes in the EECA country group.

When we move from across school estimation to school fixed effects estimation in Table 3, the results again change much. For developed countries, the positive CSE remains statistically significant only for 5 countries, down from 10 countries in Table 1, when we focus only on variations inside schools. There is a similar change for the two other groups of countries. On the whole, 6 of the EECA countries had statistically significant effects with the school fixed effects estimator. The coefficient also seems to be reduced for developing countries where only 4 countries still have significant coefficients, down from 13 in Table 1.

However, the school fixed effects estimator will produce biased estimates of the CSE if pupils are sorted into smaller or larger classes within their grade in the school on the basis of their unobserved characteristics. For instance, even if the school does not have a deliberate policy of grouping students by ability, more ambitious/able parents may insist on having their children placed in smaller classes within the grade. This would lead to a negative correlation between ability and class-size within the school and exaggerate the size of the expected negative coefficient on class size. If this occurs, the coefficient on class-size will be biased (i.e. be a bigger negative) even in a school fixed effects estimator since ability is systematically correlated with both class size and with student performance *within* the school.

To address the problem of the endogeneity of class size even within the school, we now use the pupil fixed effects estimator. Under this estimator, the identification of effects comes only from across-subject variation in class-size and in achievement score *for the same pupil* and, as such, this is an extremely stringent test for the presence of a class-size effect since it nets out the

effect of all subject-invariant pupil and school unobservables. Providing we believe that pupils who are bright in math are also bright in science (i.e. if ability is not subject specific within the maths-science subject set), this estimator takes care of all student and school level unobservables. In order to check for any non-linearity of the class size effect, both linear and quadratic specifications are presented.

When variations in achievement and class-size across subject within the same pupil are taken into account in Table 4, the CSE is diminished in all countries, compared with previous estimations. This suggests that part of the CSE previously observed across and even within school was spurious, arising due to the correlation of class-size with pupil unobservables. However, even though diminished, there remains a statistically significant class size effect for 13 out of the 33 countries for which it was possible to use the pupil fixed effects estimator, though in only 8 of these does it have the expected negative sign. One explanation for the diminished coefficients on class-size could be attenuation bias, which is exacerbated in differenced estimation. However, this is not an important cause for worry because it seems unlikely that measurement error in class-size will be subject-specific within a grade/school. As long as measurement error in class-size for the different subjects is equal and in the same direction, differencing will not cause attenuation bias. In any case, we attempt to address any attenuation bias by instrumenting class size later in Table 5.

In general, the CSE in Table 4 is positive for the developed countries, negative for the developing countries and mixed for the EECA countries. A significant relation persists only in 4 developed countries (positive in Australia, Scotland and Sweden and negative in Belgium). The CSE in Belgium becomes negative once pupil fixed effects are taken into account, but the effect is modest: a 1 SD reduction in class size (by approximately 4 pupils) would increase student performance by 0.05 SD, representing an expensive way of raising student achievement. For four other countries (England, Hong-Kong, Netherlands, New Zealand), the CSE disappears when

pupil fixed effects are taken into account. The effect differs from country to country for the EECA group of countries. For Bulgaria, the effect is more quadratic than linear, with a convex relationship and a turning point occurring at 28 pupils ($0.0508/(0.0009 \times 2) = 28.22$). The various thresholds are presented in brackets in the last column of Table 4. No significant CSE is found for Cyprus, Estonia, Lithuania, Macedonia, Romania and Slovenia (and it cannot be estimated for Slovakia and the Russian Federation) but for the other EECA countries, a CSE persists. Lastly, a CSE exists for 4 countries in the developing country group, while it was significant for 12 of them with the Least Squares estimator in Table 1. In all these 4 countries, the CSE is negative and linear. For Chile, the pupil fixed effects estimator increases the (negative) CSE, suggesting that measurement error bias is not a problem. The size of the CSE is substantial in Chile: a 1 SD reduction in class size (by 7.3 students) raises student achievement by 0.16 SD. However, in the other developing countries where there is a statistically significant CSE (Jordan and Palestine), the effect of a 1 SD reduction in class size would be to raise achievement by 0.07 and 0.05 SD respectively.

5.2. Robustness

In this section, we check the robustness of our pupil fixed effects results. A potential criticism of the pupil fixed effects approach is that its efficacy depends on the assumption that pupil unobservables do not vary by subject, i.e. that differencing a student's marks in different subjects nets out all aspects of the student's unobserved characteristics. But it may be that students are more able in, or more motivated to study, particular subjects. If students who are specially interested/able in maths are systematically allocated to smaller (or larger) maths classes – either because they lobby or because of school policy – then subject-specific class-size may be systematically correlated with student unobservables. While such a concern may seem somewhat far-fetched, and while we have already sought to address it in Table 2 where we restrict

estimation to students not class-setted by ability, we seek to further control for the possibility of different abilities for different subjects.

If pupil ability differs by subject, the ideal way to control for this would be to have indicators of ability for each subject. While such indicators are not available, it is possible to build a *proxy* for subject-specific ability based on the TIMSS database. An indicator of subject-specific ability has been constructed from five questions asked directly of the pupil. These are detailed in Appendix 1. The first two columns of Table 5 present the results of pupil fixed effects achievement equations fitted for each country, which include the subject varying pupil ability variable. Although we do not report the underlying achievement regressions for each country, the coefficients on the ability variables were positive and significant, i.e. subject-specific ability enables pupils to have better results in achievement tests in that subject, as might be expected. But what interests us more is the effect of the inclusion of the ability measure on the CSE. Compared with the results in Table 4, the CSE here remains roughly identical for the majority of the countries. Thus, it appears that subject-varying ability is not systematically or strongly correlated with class size. This could be because within a grade in a school students may not be able to engineer to sort themselves into smaller classes for those subjects in which they are more capable, while remaining in larger classes for the other subjects.

A potential omitted variable bias occurs if other class-level variables are correlated with class size but are omitted from the equation. These can be variables concerning the teachers. Indeed, typically, other than class-size, only the teacher characteristics can vary between the classes of the same grade in a school. A possible example of bias is then the possibility that there is a correlation between the class size and teacher characteristics. For instance, if more experienced teachers are, for some reason, systematically allocated to the smaller classes and teacher experience affects student achievement, then class-size could ‘pick up’ the effect of omitted teacher experience. But in order for this to apply, the correlation of teacher characteristics

with class-size would have to be subject-specific, e.g. more experienced maths teachers are systematically assigned to teach small (or large) maths classes and more experienced science teachers are systematically assigned to small (or large) science classes, something that appears unlikely. In any case, in order to check for such omitted variable bias, we introduce four teacher characteristics into the estimation: teacher's gender, age, possession of a master-level or higher diploma in the subject, and experience entered in quadratic form. Since teacher characteristics vary by subject, it is possible to include these variables in pupil fixed effect estimations. The results are presented in the columns (3) and (4) of Table 5. Neither the size nor the significance of the class size coefficients changes, or when it does, generally the coefficients are higher. We can thus conclude that the previous class-size effects are not biased.

A final robustness test we perform is to see if measurement error is driving the pupil fixed effects results. To do this, we instrument class-size with the average class-size in the subject across all grades in the school where this is possible (in some countries there is data only on one class per school). This is identical to the methodology followed in Woessmann and West (2006) but is done here in the context of a pupil fixed effects regression. The results are presented in Table 6. It shows that while the CSE is unaffected by instrumenting in most countries (compared with Table 4) the coefficient on class size increases significantly in Australia, Belgium, Scotland and Hungary. Thus measurement error did downward bias results. The CSE roughly doubles in Australia and Belgium. However, despite this it remains the case that the class size effect is statistically significant and of the expected (negative) sign in only 7 out of the 19 countries for which it was possible to estimate pupil fixed effects achievement production functions with IV. Moreover, the size of the effect is at best modest, as we discuss in the next section.

5.3 Synthesis

The magnitude of the CSE estimated in our various regressions is synthesized in Table 7. The first column shows the mean and (in parentheses) SD of class size in the country. Columns 2, 4, 6, 8 and 10 respectively show the effect, on pupil achievement, of a 1 SD reduction in class size. The odd numbered columns represent the class size above which the direction of the class-size – achievement relationship is reversed in countries where the relationship is quadratic.

It is clear that neither the direction nor the size of the class-size effects are similar across the countries. In addition, it is clear that where effects are present, their size is very small in the majority of countries. In the 33 countries for which it was possible to estimate pupil fixed effects achievement equations (column 8), a statistically significant effect exists in only 16 and among these, the CSE is negative in only 9 countries (it is quadratic in another 2). Finally, among these 9 countries, the magnitude of the CSE, following a 1 SD decrease in class size, is greater than 0.10 SD of achievement in only 1 country (Chile). A 1 SD reduction in class-size implies a large increase in number of teachers, which is costly. In the developed group of countries, only Belgium, Scotland and Sweden have a class-size effect of any magnitude but the *sign* of the effect is positive for Scotland and Sweden. When measurement error is corrected for (column 10's results), class size reductions are beneficial for student achievement in 8 out of 23 countries for which estimates could be made but again, the effect size is small – a 1 SD reduction in class size results in a 0.10 SD increase in achievement in only 2 out of these 8 countries (Belgium and Chile).

We can compare our results with previous studies. There are hundreds of econometric studies seeking to estimate educational production functions. The meta-analyses carried out by Hanushek (1997, 2003, 2006) synthesise findings on the impact of class size on pupil performance. Of the 276 US estimates of the class-size effect listed by Hanushek (2003), only 14% found a significant negative class-size effect. Even when he takes only the value-added studies estimated within a state, 17% find a significant negative class-size effect. Developing

country studies also yield pessimistic conclusions, with only about a quarter (27%) finding a significant negative effect. Hanushek and Luque (2003) using TIMSS 1995 data on performance of grade 8 pupils in mathematics and sciences, show that 52% of the class size effects were positive, while 42% were not statistically significant. Although, Krueger (2003) criticises meta-analyses for including papers that pay little attention to the endogeneity of class-size in achievement production functions, our paper using a stringent methodology to address endogeneity confirms the findings of meta-analyses.

Lastly, we attempt to understand and explain our findings with the help of Table 8. In particular we ask whether CSE are smaller in richer countries, supporting a ‘diminishing returns’ explanation. We also ask if CSE are smaller in countries where teachers are more skilled. We present the results for the three groups of countries as before – the developed countries, the EECA countries and the developing countries. We measure resources by per capita GDP in purchasing power parity dollars (column 6), and also by per pupil education expenditure (both in absolute terms and as a proportion of per capita GDP, column 7). Column 4 shows the mean and SD of class-size for each country group. Column 3 summarises the CSE in the three country types and we trust the pupil fixed effects (PFE) result the most. These ‘effects’ represent the number of SD by which student achievement would fall if there was a 1 SD increase in class-size. It is clear that the CSE is much greater in developing countries than in the EECA countries and that in turn is greater than in the developed countries where, if anything, the CSE is positive albeit weak. This follows the hierarchy of per student expenditure on education and also country resources. In developing countries, where resources for education are more limited, reducing class size has the largest effect on achievement: a 1 SD reduction in class-size (by 8.2 pupils per class) would increase student achievement by 0.06 SD^2 ; in EECA countries where resources are

² This is likely to be an over-estimate since it represents the mean of the class-size effect taken only over those developing countries where there was a statistically significant class-size effect. If we add the developing countries where the class-size effect was not significantly different from zero, i.e. assign a zero class-size effect in those

more plentiful, the CSE is small. Thus there is some support for diminishing returns to resources. Secondly, teacher quality is measured by teacher education level (column 9), in particular by the percentage of teachers with MA qualifications and training. Again in regions where teachers are more skilled / higher quality (the EECA and developed countries), the CSE is smaller than in the region with the lowest quality teachers, namely developing countries.

6. Conclusion

The impact of class size on student achievement remains a thorny question for politicians and educational decision makers. If class-size reductions could bring a sizeable gain in achievement for pupils, decision makers would be tempted to act in this direction. However, recent literature in this field leaves researchers as perplexed as policy makers. In his review work, Hanushek (1997, 2003, 2006) shows the lack of any consistent or strong relationship between class size and pupil performance across a large range of studies.

In this study, we have sought to check the veracity of these findings using a new technique that avoids arguably the most important sources of endogeneity bias. We measure the class size effect by relating the difference in achievement across subjects to the difference in class-size across subjects. A subject-differenced achievement production function helps to address the non-random matching of children to specific schools and to classes within schools.

We proceeded in several stages. Weighted least squares estimates showed significant and large class-size effects in a high proportion of the countries studied, though many of the 'effects' were positive for developed and transition countries. While WLS effects were negative for developing countries, they were small in size. Pupil fixed effects estimation of the achievement production function powerfully nets out the effect of all subject-invariant pupil, class and school unobservable variables. Several tests of robustness were carried out in order to confirm the results

countries and take the mean across all developing countries, the class-size effect on student achievement is only about -0.03 SD rather than -0.06 SD, for a one SD increase in class-size.

obtained. In particular, we controlled for students' subject-specific ability, (subject-specific) teacher characteristics, and also controlled for measurement error. The results show a statistically significant effect of class size for 16 out of 33 countries for which pupil fixed effects achievement equation could be fitted. However, in only 10 countries is the effect of the expected negative sign. Moreover, the magnitude of the class size effect is small in most cases. At best, an expensive 1 SD reduction in class size in developing countries (equal to reducing 8.2 students per class) allows pupil scores to increase by only 0.06 standard deviations.

Our study confirms one of the principal conclusions of Hanushek (2003, 2006) in his meta-analyses, namely that class size does not have a systematic or substantial effect on pupil performance. Detractors (e.g. Krueger, 2003) have argued that meta-analyses based conclusions are flawed because many of the studies that go into meta-analyses are of low quality, for instance, across-state or across-school rather than within-state or within-school studies. In the present paper, Hanushek's conclusions are corroborated using a much more stringent methodology that accounts for endogeneity in the pupil achievement class-size relationship. We conclude that, given the economic implications, class-size reductions do not appear to be a cost-effective strategy for raising student achievement levels in developed, developing or transition countries. We find that class-size effects are smaller in resource-rich countries than in developing countries, supporting the idea of diminishing returns. We also find that class size effects are smaller in regions with higher teacher quality, suggesting that more skilful teachers cope better with larger classes than less skilful teachers.

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Appendix 1 : **Evaluation of pupils' "facilities" as a proxy for ability**

As mentioned in the robustness section (Section 5.2) of the text, we construct and use a proxy for ability for each pupil in each subject. This is calculated starting from 4 questions put directly to the pupil for each evaluated subject. As the pupil must answer for each subject, the ‘capability’ level built varies between the subjects.

The scale of our capability proxy varies between 0 and 10 and it depends on the 4 variables detailed below. The possible answers of the pupil are based on a Likert scale of four occurrences (agree a lot, agree a little, disagree a little, disagree a lot). :

- "I usually do well in [subject]"
- "[Subject] is more difficult for me than of many of my classmates"
- "I enjoy learning [subject]"
- "I learn things quickly in [subject]"

A score between 1 and 5 is given for each question, the highest value corresponding to the response “agree a lot”. The coefficient of “facility” is calculated by carrying out the arithmetic mean of the scores obtained. Hence, as we have 4 different variables, a score between 4 and 16 is computed. Lastly, for having a score between 0 and 10, we simply multiply this score by 10/16. It should be noted that, by construction, the minimum of a score is in fact 2.5 and not 0, but this is not a problem for our study.

Figure 1 : Kernel Densities of Scores for Australia

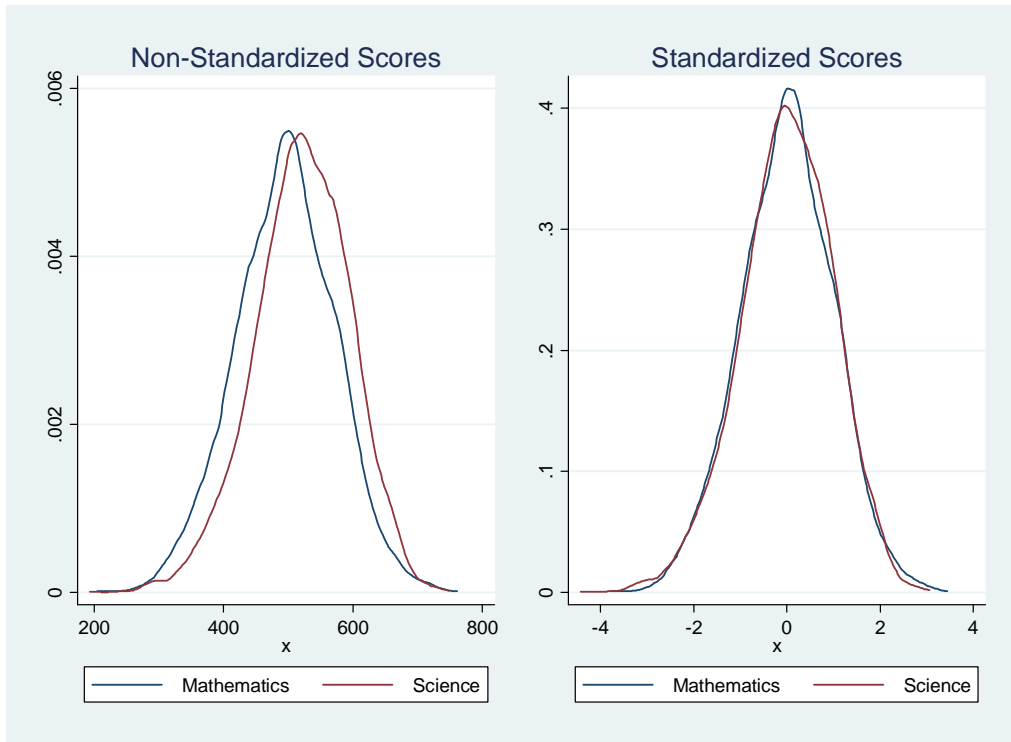


Figure 2 : Kernel Densities of Scores for the USA

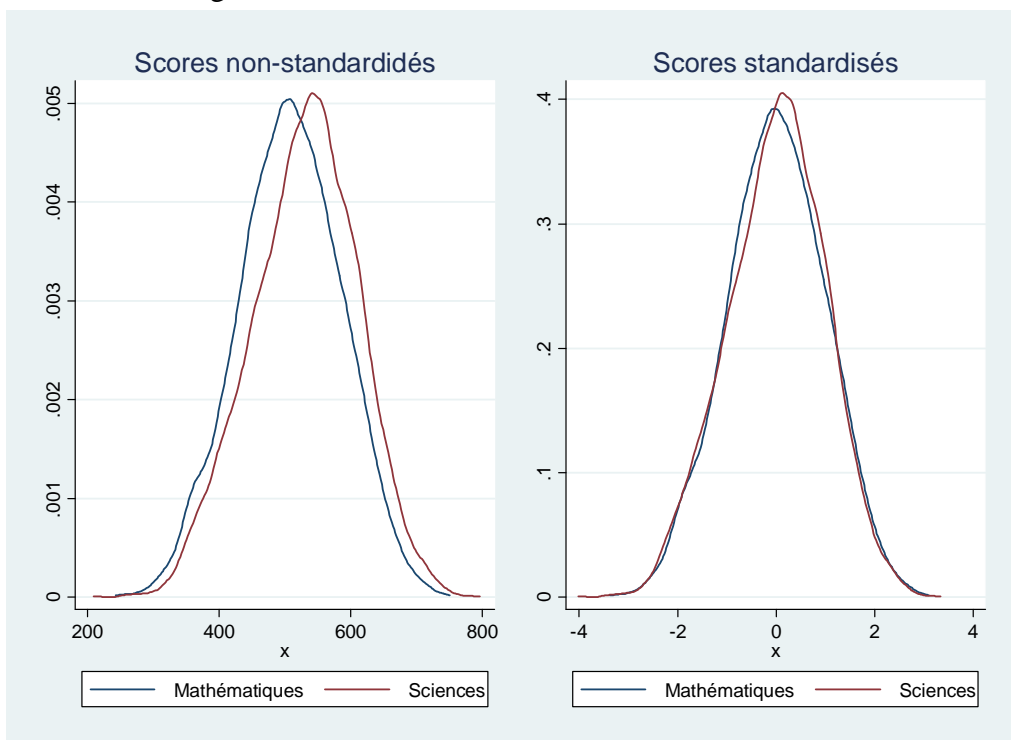


Figure 3 : Kernel Densities of Scores for Bulgaria

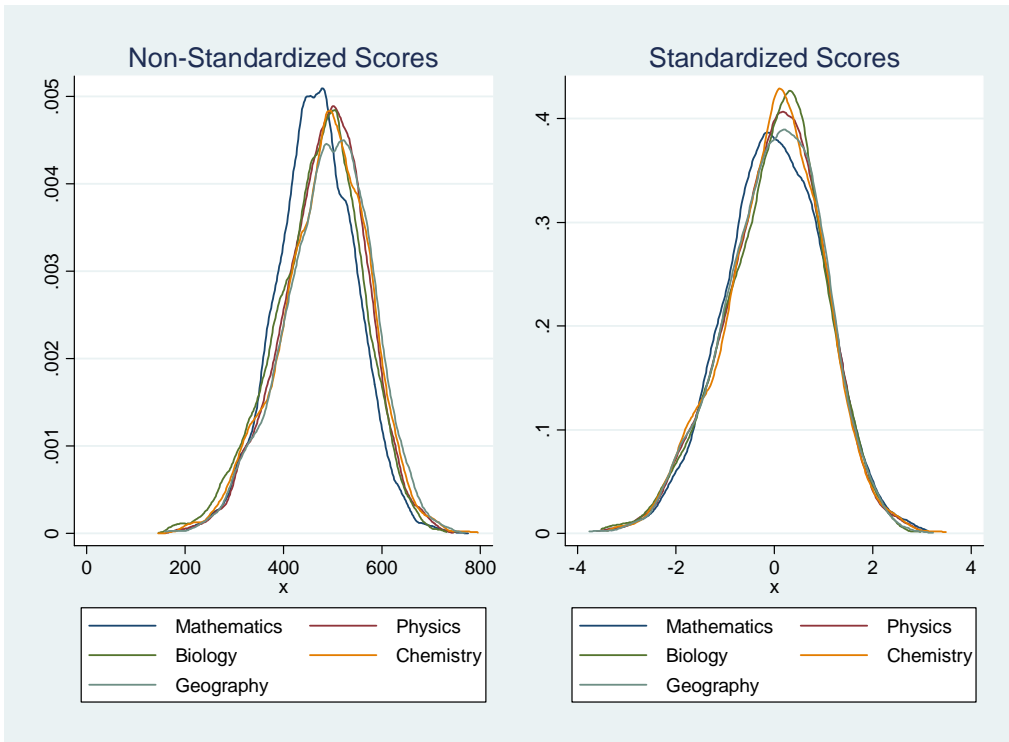


Figure 4 : Kernel Densities of Scores for Egypt

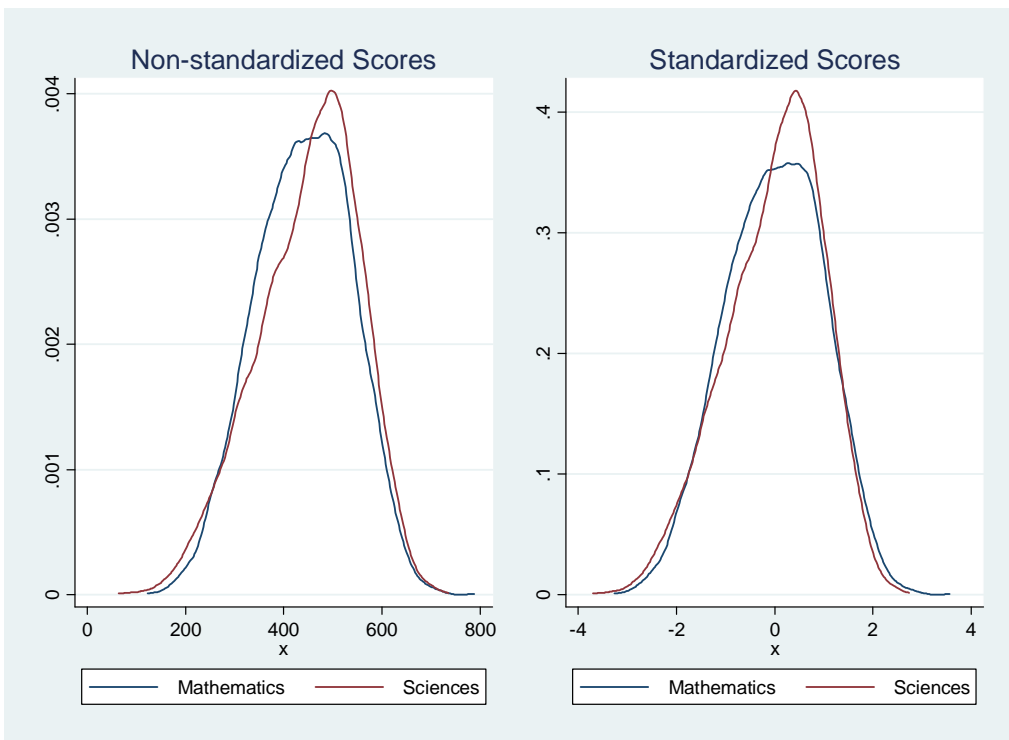


Table 1 : Weighted Least Squares – Full Sample

Country	Class size Mean (SD)	(1)	(2)	(3)	(4)	(5)	(6)	(7)
		Linear Form		Quadratic Form				
		Coef.	t Stat	Linear Coef.	t Stat	Quadratic Coef.	t Stat	Turning point
Australia	24.16 (4.66)	0.0354*	(9.38)	0.0777*	(3.65)	-0.0009†	(2.09)	42.7
Belgium (Fl.)	20.77 (4.15)	0.0440*	(15.08)	0.1546*	(7.98)	-0.0029*	(5.95)	26.5
England	27.30 (5.55)	0.0402*	(10.15)	0.1260*	(8.03)	-0.0015*	(5.85)	41.6
Hong-Kong	39.59 (4.06)	0.0631*	(18.35)	-0.1021*	(5.26)	0.0024*	(8.59)	20.9
Italy	22.05 (3.65)	0.0111*	(2.73)	-0.0192	(0.50)	0.0007	(0.81)	
Japan	35.57 (3.86)	0.0029	(0.71)	-0.0540*	(3.67)	0.0014*	(4.64)	29.5
Korea	37.16 (5.26)	-0.0052	(1.60)	0.0020	(0.09)	-0.0001	(0.33)	
Netherlands	26.01 (3.74)	0.0605*	(13.83)	0.0063	(0.21)	0.0011‡	(1.86)	
New Zealand	25.88 (4.90)	0.0319*	(8.71)	-0.0204	(1.04)	0.0010*	(2.65)	
Norway	26.03 (3.82)	-0.0094†	(2.27)	0.0205	(1.35)	-0.0007	(2.24)	
Scotland	24.08 (7.31)	0.0451*	(13.21)	0.0958*	(9.88)	-0.0009*	(5.13)	51.5
Sweden	21.13 (7.19)	0.0220*	(10.56)	0.0649*	(10.62)	-0.0009*	(7.79)	37.2
Taiwan	37.24 (5.47)	0.0171*	(7.57)	-0.0790*	(5.55)	0.0013*	(7.15)	31.1
USA	24.76 (6.81)	0.0013	(0.90)	-0.0067	(1.04)	0.0002	(1.42)	
Bulgaria	23.15 (5.20)	-0.0214*	(5.84)	-0.0196	(1.23)	-0.0004	(0.11)	
Cyprus	25.65 (4.17)	-0.0009	(0.30)	0.0358	(1.52)	-0.0008	(1.52)	
Estonia	29.22 (6.92)	0.0048†	(2.46)	0.0053	(1.24)	0.0000	(0.18)	
Hungary	23.74 (5.56)	0.0114*	(4.33)	-0.1021*	(5.26)	0.0024*	(8.59)	18.9
Latvia	29.40 (12.29)	0.0033*	(2.87)	0.0037	(0.83)	0.0000	(0.10)	
Lithuania	26.39 (4.52)	0.0016	(0.51)	-0.0051	(0.40)	0.0001	(0.61)	
Macedonia	28.14 (5.19)	-0.0067*	(2.75)	-0.0116	(0.85)	0.0001	(0.37)	
Moldova	25.78 (6.48)	0.0082*	(3.22)	0.0193†	(1.93)	-0.0002	(1.29)	
Romania	25.06 (5.45)	-0.0118*	(3.86)	-0.1102*	(6.93)	0.0020*	(6.98)	27.2
Russian Fed.	24.40 (5.22)	-0.0044	(1.41)	-0.0530*	(3.16)	0.0011*	(3.10)	23.4
Serbia	26.66 (6.29)	0.0024	(1.18)	0.0132†	(2.12)	-0.0002†	(2.08)	39.2
Slovakia	25.80 (5.08)	0.0116*	(4.65)	0.1029*	(4.78)	0.0023*	(5.45)	21.9
Slovenia	22.05 (4.10)	-0.0052	(1.15)	-0.0633*	(2.59)	0.0015†	(2.46)	21.6
Armenia	31.79 (11.66)	-0.0053*	(5.28)	-0.0056	(0.89)	0.0000	(0.05)	
Bahrain	30.70 (5.16)	-0.0194*	(6.22)	-0.0579*	(3.82)	0.0006*	(2.53)	45.0
Botswana	37.13 (4.84)	-0.0182*	(6.52)	-0.2361*	(7.62)	0.0031*	(7.07)	38.7
Chile	35.46 (7.32)	-0.0016	(1.07)	0.0133	(1.55)	-0.0002‡	(1.82)	
Egypt	37.40 (7.86)	-0.0025†	(1.96)	-0.0031	(0.46)	0.0000	(0.10)	
Ghana	38.36 (11.76)	0.0093*	(8.04)	0.0100‡	(1.92)	0.0000	(0.14)	
Indonesia	40.57 (5.92)	0.0120*	(5.48)	-0.0141	(1.15)	0.0004†	(2.16)	
Iran	29.70 (6.78)	-0.0012	(0.69)	0.0091	(1.20)	-0.0002	(1.45)	
Jordan	26.37 (8.23)	-0.0117*	(6.84)	-0.0481*	(5.19)	0.0006*	(4.11)	43.4
Lebanon	29.16 (7.90)	0.0054*	(3.36)	-0.1020	(1.39)	0.0002†	(2.24)	
Malaysia	37.10 (5.10)	0.0008	(0.31)	-0.2325*	(11.93)	0.0032*	(12.04)	36.2
Morocco	42.04 (13.74)	-0.0024	(0.82)	-0.0171	(1.28)	0.0002	(1.17)	
Palestine	39.52 (8.63)	-0.0090*	(5.84)	-0.0121‡	(1.86)	0.0000	(0.48)	
Philippines	54.75 (10.81)	-0.0080*	(7.04)	-0.0237*	(2.93)	0.0001†	(2.03)	81.8
Saudi Arabia	29.23 (8.46)	-0.0050†	(2.34)	-0.0040	(0.40)	-0.0000	(0.09)	
South Africa	45.04 (14.68)	-0.0055*	(8.69)	-0.0092*	(2.88)	0.0000	(1.21)	
Syrian A.R.	37.73 (9.42)	-0.0062†	(2.52)	-0.0283*	(2.98)	0.0003†	(2.36)	45.6
Tunisia	34.09 (3.92)	0.0016	(0.55)	0.0107	(0.48)	-0.0001	(0.41)	

Significance levels (based on clustering-robust standard errors): *1%. †5%. ‡10%. Dependent variable: achievement z-score. Controlling for student, school and family background variables. Numbers in parenthesis represents robust absolute t statistics. In order to simplify the lecture, coefficients significant at the 10 percent level or better are in bold. Simple least squares (without weighting) gives approximately same results. Logging class-size also gives results that are not too dissimilar.

Table 2: Weighted Least Squares – Reduced Sample (Without Ability Grouped Classes)

Country		(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Class size	Linear Form		Quadratic Form				
	Mean (SD)	Coef.	t Stat	Linear Coef.	t Stat	Quadratic Coef.	t Stat	Turning point
Australia	24.16 (4.66)	-0.0024	(0.37)	0.0526	(0.98)	-0.0011	(1.04)	
Belgium (Fl.)	20.77 (4.15)	0.0409*	(10.18)	0.1848*	(6.82)	-0.0037*	(5.56)	
England	27.30 (5.55)	0.0486*	(4.38)	-0.1032*	(3.56)	0.0035*	(5.12)	14.9
Hong-Kong	39.59 (4.06)	0.0705*	(12.15)	-0.1644*	(4.27)	0.0033*	(5.86)	25.2
Italy	22.05 (3.65)	-0.0012	(0.30)	-0.1029*	(2.69)	0.0024*	(2.71)	21.8
Japan	35.57 (3.86)	0.0213*	(5.22)	-0.0901*	(4.33)	0.0026*	(5.68)	32.0
Korea	37.16 (5.26)	-0.0079	(0.98)	-0.0218	(0.25)	0.0002	(0.16)	
Netherlands	26.01 (3.74)	0.0651*	(12.04)	-0.0580‡	(1.64)	0.0025*	(3.68)	11.5
New Zealand	25.88 (4.90)	0.0287*	(5.83)	0.0230	(0.85)	0.0001	(0.21)	
Norway	26.03 (3.82)	n	n	n	n	n	n	
Scotland	24.08 (7.31)	0.0594*	(7.89)	-0.0756†	(1.99)	0.0029*	(3.61)	13.0
Sweden	21.13 (7.19)	-0.0108*	(2.66)	-0.0056	(0.34)	-0.0001	(0.36)	
Taiwan	37.24 (5.47)	0.0139*	(5.75)	-0.0831*	(5.50)	0.0013*	(6.87)	31.5
USA	24.76 (6.81)	-0.0052†	(2.07)	-0.0167	(1.40)	0.0002	(1.12)	
Bulgaria	23.15 (5.20)	-0.0181*	(3.30)	0.0082	(0.25)	-0.0006	(0.82)	
Cyprus	25.65 (4.17)	0.0006	(0.18)	-0.0036	(0.12)	0.0001	(0.13)	
Estonia	29.22 (6.92)	0.0118*	(5.10)	0.0273*	(4.86)	-0.0002*	(3.70)	69.7
Hungary	23.74 (5.56)	0.0213*	(5.22)	-0.0901*	(4.33)	0.0026*	(5.68)	17.7
Latvia	29.40 (12.29)	0.0022	(1.63)	-0.0009	(0.18)	0.0004	(0.70)	
Lithuania	26.39 (4.52)	0.0053	(1.32)	0.0136	(0.89)	-0.0002	(0.63)	
Macedonia	28.14 (5.19)	-0.0015	(0.54)	0.0023	(0.14)	-0.0001	(0.24)	
Moldova	25.78 (6.48)	0.0120*	(3.20)	0.0705*	(4.13)	-0.0010*	(3.81)	36.4
Romania	25.06 (5.45)	-0.0056	(1.30)	-0.0925*	(4.05)	0.0017*	(4.34)	27.2
Russian Fed.	24.40 (5.22)	-0.0094†	(2.04)	-0.0227	(1.07)	0.0003	(0.67)	
Serbia	26.66 (6.29)	-0.0004	(0.17)	0.0099	(1.33)	-0.0001‡	(1.63)	
Slovakia	25.80 (5.08)	0.0121*	(4.05)	-0.1291*	(5.38)	0.0029*	(6.04)	22.3
Slovenia	22.05 (4.10)	-0.0028	(0.42)	-0.1911*	(3.41)	0.0044*	(3.46)	21.9
Armenia	31.79 (11.66)	-0.0069*	(6.07)	-0.0214*	(2.98)	0.0002†	(2.16)	54.3
Bahrain	30.70 (5.16)	-0.0200*	(5.36)	-0.1180*	(6.50)	0.0016*	(5.43)	36.2
Botswana	37.13 (4.84)	-0.0216*	(7.21)	-0.2006*	(6.14)	0.0025*	(5.51)	39.6
Chile	35.46 (7.32)	-0.0058*	(3.36)	-0.0103	(0.94)	0.0001	(0.43)	
Egypt	37.40 (7.86)	0.0005	(0.36)	-0.0232*	(2.67)	0.0003*	(2.82)	42.7
Ghana	38.36 (11.76)	0.0077*	(5.91)	0.0133†	(2.03)	-0.0001	(0.89)	
Indonesia	40.57 (5.92)	0.0132*	(5.25)	0.0193	(1.08)	-0.001	(0.35)	
Iran	29.70 (6.78)	-0.0044	(1.07)	0.0739†	(2.07)	-0.0013†	(2.23)	27.8
Jordan	26.37 (8.23)	-0.0098*	(5.44)	-0.0392*	(3.76)	0.0004*	(2.90)	45.3
Lebanon	29.16 (7.90)	0.0065*	(3.79)	-0.0039	(0.49)	0.0002	(1.34)	
Malaysia	37.10 (5.10)	-0.0045	(1.16)	-0.1773*	(6.47)	0.0025*	(6.23)	35.7
Morocco	42.04 (13.74)	-0.0049	(1.42)	0.0007	(0.03)	-0.0001	(0.22)	
Palestine	39.52 (8.63)	-0.0068*	(3.85)	-0.0085	(1.14)	0.0000	(0.24)	
Philippines	54.75 (10.81)	-0.0135*	(9.74)	-0.0697*	(9.01)	0.0005*	(7.53)	67.4
Saudi Arabia	29.23 (8.46)	-0.0060†	(2.54)	-0.0001	(0.01)	-0.0001	(0.55)	
South Africa	45.04 (14.68)	-0.0064*	(9.15)	-0.0079†	(2.06)	0.0000	(0.40)	
Syrian A.R.	37.73 (9.42)	-0.0028	(0.96)	-0.0215‡	(1.70)	0.0002	(1.51)	
Tunisia	34.09 (3.92)	-0.0019	(0.50)	0.0223	(0.80)	-0.0004	(0.88)	

Significance levels (based on clustering-robust standard errors): *1%. †5%. ‡10%. Dependent variable: achievement z-score. Controlling for student, school and family background variables. Numbers in parenthesis represents robust absolute t statistics. In order to simplify the lecture, coefficients significant at the 10 percent level or better are in bold. Only pupils who are not present in classes grouped by ability are included in the regression (WLS reduced sample regression). See text for more details.

Table 3 : School Fixed Effect Estimation

Country	Class size Mean (SD)	(1)	(2)	(3)	(4)	(5)	(6)	(7)
		Linear Form		Quadratic Form				Turning point
		Coef.	t Stat	Coef.	t Stat	Coef.	t Stat	
Australia	24.16 (4.66)	0.0174*	(2.86)	-0.0194	(0.68)	0.0009	(1.31)	
Belgium (Fl.)	20.77 (4.15)	-0.0001	(0.03)	-0.0694*	(4.10)	0.0019*	(4.29)	18.1
England	27.30 (5.55)	0.0391*	(8.17)	0.1376*	(7.91)	-0.0017*	(5.95)	40.7
Hong-Kong	39.59 (4.06)	0.0035	(0.95)	0.0206	(0.90)	-0.0003	(0.75)	
Italy	22.05 (3.65)	-0.0209*	(2.69)	-0.1765*	(2.53)	0.0036†	(2.27)	24.3
Japan	35.57 (3.86)	0.0029	(0.17)	0.0669	(0.52)	-0.0010	(0.51)	
Korea	37.16 (5.26)	0.0001	(0.01)	-0.0489	(1.34)	0.0007	(1.22)	
Netherlands	26.01 (3.74)	-0.0011	(0.16)	-0.0464	(1.25)	0.0011	(1.19)	
New Zeland	25.88 (4.90)	0.0041	(0.82)	-0.0705*	(3.33)	0.0017*	(3.73)	20.9
Norway	26.03 (3.82)	0.0005	(0.05)	0.0106	(0.32)	-0.0002	(0.28)	
Scotland	24.08 (7.31)	0.0122*	(4.06)	0.0317*	(3.50)	-0.0003†	(2.10)	96.4
Sweden	21.13 (7.19)	0.0216*	(8.88)	0.0740*	(11.52)	-0.0010*	(8.93)	35.9
Taiwan	37.24 (5.47)	0.0898	(0.78)					
USA	24.76 (6.81)	0.0043†	(2.06)	0.0033	(0.43)	0.0000	(0.14)	
Bulgaria	23.15 (5.20)	-0.0141	(1.49)	-0.0748†	(2.27)	0.0014*	(2.45)	26.1
Cyprus	25.65 (4.17)	-0.0007	(0.23)	0.0039	(0.15)	-0.0001	(0.18)	
Estonia	29.22 (6.92)	0.0066*	(2.93)	0.0161*	(3.00)	-0.0001†	(2.11)	72.2
Hungary	23.74 (5.56)	-0.0026	(0.83)	-0.0023	(0.11)	0.0000	(0.01)	
Latvia	29.40 (12.29)	0.0028†	(2.10)	0.0056	(0.85)	-0.0000	(0.48)	
Lituania	26.39 (4.52)	0.0055‡	(1.67)	0.0109	(0.84)	-0.0001	(0.44)	
Macedonia	28.14 (5.19)	0.0192*	(3.55)	0.0833*	(3.35)	-0.0010*	(2.64)	40.5
Moldova	25.78 (6.48)	0.0047	(1.14)	0.0220	(1.27)	-0.0003	(1.10)	
Romania	25.06 (5.45)	0.0036	(1.05)	0.0563*	(3.21)	-0.0009*	(3.29)	31.2
Russian Fed.	24.40 (5.22)	-0.0228	(0.75)	0.2310	(1.06)	-0.0050	(1.20)	
Serbia	26.66 (6.29)	-0.0022	(0.97)	-0.0055	(0.70)	0.0000	(0.42)	
Slovakia	25.80 (5.08)	0.0221*	(2.78)	0.0091	(0.17)	0.0003	(0.25)	
Slovenia	22.05 (4.10)	-0.0066‡	(1.69)	-0.0102	(0.54)	0.0001	(0.19)	
Armenia	31.79 (11.66)	-0.0009	(0.97)	-0.0051	(0.85)	0.0001	(0.71)	
Bahrain	30.70 (5.16)	-0.0117†	(1.97)	-0.0835*	(2.69)	0.0012†	(2.35)	35.3
Botswana	37.13 (4.84)	0.0033	(0.23)	-0.0220	(0.17)	0.0004	(0.20)	
Chile	35.46 (7.32)	-0.0212‡	(1.89)	0.0075	(0.19)	-0.0005	(0.64)	
Egypt	37.40 (7.86)	-0.0024	(0.63)	0.0113	(0.36)	-0.0002	(0.46)	
Ghana	38.36 (11.76)	-0.0053	(0.78)	0.0140	(0.28)	-0.0002	(0.39)	
Indonesia	40.57 (5.92)	0.0030	(0.53)	0.0848	(0.48)	-0.0010	(0.46)	
Iran	29.70 (6.78)	-0.0006	(0.18)	0.0080	(0.60)	-0.0001	(0.67)	
Jordan	26.37 (8.23)	-0.0121	(1.38)	-0.3859	(0.85)	0.0005	(0.64)	
Lebanon	29.16 (7.90)	0.0016	(0.52)	-0.0088	(0.40)	0.0001	(0.50)	
Malaysia	37.10 (5.10)	0.0426*	(2.98)	-0.0193	(0.14)	0.0008	(0.45)	
Morocoo	42.04 (13.74)	0.0151	(1.14)	-0.01425	(1.33)	0.0023	(1.52)	
Palesti. N.A.	39.52 (8.63)	-0.0055†	(2.10)	-0.0070	(0.65)	0.0000	(0.14)	
Philippines	54.75 (10.81)	-0.0004	(0.11)	0.0125	(0.39)	-0.0001	(0.41)	
Saudi Arabia	29.23 (8.46)	-0.0030	(0.41)	-0.1289*	(2.52)	0.0019*	(2.60)	33.7
South Africa	45.04 (14.68)	-0.0012	(1.01)	-0.0067	(0.82)	0.0000	(0.68)	
Syrian A.R.	37.73 (9.42)	-0.0024	(0.32)	-0.0350	(1.20)	0.0007	(1.25)	
Tunisia	34.09 (3.92)	0.0005	(0.09)	-0.0132	(0.33)	0.0002	(0.33)	

Significance levels (based on clustering-robust standard errors): *1%. †5%. ‡10%. Dependent variable: Actual class size. Controlling for student, school and family background variables. Numbers in parenthesis represent robust absolute t statistics. In order to simplify the lecture, coefficients significant at the 10th percent level are indicated in bold. Controlling for school fixed effects and student- and family-background.

Table 4: Pupil Fixed Effect Estimation

Country	Class size Mean (SD)	(1)	(2)	(3)	(4)	(5)	(6)	(7)
		Linear Form		Quadratic Form				
		Coef.	t Stat	Linear Coef.	t Stat	Quadratic Coef.	t Stat	Turning point
Australia	24.16 (4.66)	0.0122*	(2.71)	0.0065	(0.32)	0.0001	(0.27)	
Belgium (Fl.)	20.77 (4.15)	-0.0140*	(2.43)	0.0330	(1.00)	-0.0012	(1.46)	
England	27.30 (5.55)	-0.0007	(0.16)	-0.0038	(0.20)	0.0001	(0.19)	
Hong-Kong	39.59 (4.06)	0.0013	(0.37)	-0.0207	(0.85)	0.0004	(0.93)	
Italy	22.05 (3.65)	n	n	n	n	n	n	
Japan	35.57 (3.86)	n	n	n	n	n	n	
Korea	37.16 (5.26)	0.0021	(0.47)	-0.0156	(0.50)	0.0002	(0.57)	
Netherlands	26.01 (3.74)	-0.0025	(0.41)	-0.0400	(1.27)	0.0009	(1.17)	
New Zealand	25.88 (4.90)	-0.0045	(0.83)	-0.0340	(1.39)	0.0006	(1.21)	
Norway	26.03 (3.82)	n	n	n	n	n	n	
Scotland	24.08 (7.31)	0.0065†	(2.26)	0.0106	(0.95)	-0.0001	(0.39)	
Sweden	21.13 (7.19)	0.0074*	(3.59)	0.0211*	(3.33)	-0.0003†	(2.45)	42.0
Taiwan	37.24 (5.47)	n	n	n	n	n	n	
USA	24.76 (6.81)	0.0008	(0.68)	-0.0015	(0.35)	0.0000	(0.57)	
Bulgaria	23.15 (5.20)	-0.0146†	(2.42)	-0.0630*	(3.08)	0.0011*	(3.07)	27.7
Cyprus	25.65 (4.17)	0.0013	(0.63)	0.0273	(1.22)	-0.0006	(1.17)	
Estonia	29.22 (6.92)	-0.0002	(0.18)	0.0014	(0.42)	-0.0000	(0.64)	
Hungary	23.74 (5.56)	-0.0051†	(2.31)	-0.0159	(1.21)	0.0002	(0.88)	
Latvia	29.40 (12.29)	0.0020†	(2.42)	0.0019	(0.52)	0.0000	(0.04)	
Lithuania	26.39 (4.52)	-0.0020	(0.92)	-0.0103	(1.51)	0.0002	(1.36)	
Macedonia	28.14 (5.19)	0.0048	(1.23)	-0.0161	(0.70)	0.0003	(0.95)	
Moldova	25.78 (6.48)	0.0109†	(2.55)	0.0306‡	(1.77)	-0.0003	(1.24)	
Romania	25.06 (5.45)	-0.0008	(0.36)	0.0138	(1.08)	-0.0002	(1.25)	
Russian Fed.	24.40 (5.22)	n	n	n	n	n	n	
Serbia	26.66 (6.29)	-0.0029†	(2.43)	-0.0014	(0.26)	-0.0000	(0.29)	
Slovakia	25.80 (5.08)	n	n	n	n	n	n	
Slovenia	22.05 (4.10)	-0.0012	(0.41)	-0.0241‡	(1.68)	0.0007‡	(1.65)	16.8
Armenia	31.79 (11.66)	-0.0022*	(3.50)	-0.0006	(0.15)	-0.0000	(0.43)	
Bahrain	30.70 (5.16)	-0.0006	(0.11)	-0.1038†	(1.98)	0.0016†	(1.97)	31.7
Botswana	37.13 (4.84)	n	n	n	n	n	n	
Chile	35.46 (7.32)	-0.0218†	(2.32)	0.0082	(0.25)	-0.0008	(0.82)	
Egypt	37.40 (7.86)	-0.0023	(0.95)	-0.0013	(0.07)	-0.0000	(0.05)	
Ghana	38.36 (11.76)	n	n	n	n	n	n	
Indonesia	40.57 (5.92)	n	n	n	n	n	n	
Iran	29.70 (6.78)	0.0018	(0.66)	0.0127	(1.30)	-0.0002	(1.22)	
Jordan	26.37 (8.23)	-0.0084†	(1.97)	-0.0324	(1.31)	0.0004	(1.01)	
Lebanon	29.16 (7.90)	0.0010	(0.39)	-0.0160	(0.87)	0.0002	(0.95)	
Malaysia	37.10 (5.10)	n	n	n	n	n	n	
Morocco	42.04 (13.74)	n	n	n	n	n	n	
Palestine	39.52 (8.63)	-0.0053*	(3.08)	-0.0044	(0.64)	-0.0000	(0.13)	
Philippines	54.75 (10.81)	-0.0007	(0.27)	0.0011	(0.05)	-0.0000	(0.08)	
Saudi Arabia	29.23 (8.46)	n	n	n	n	n	n	
South Africa	45.04 (14.68)	-0.0015	(1.28)	-0.0102	(1.45)	0.0001	(1.25)	
Syrian A.R.	37.73 (9.42)	0.0022	(0.43)	-0.0235	(1.22)	0.0005	(1.45)	
Tunisia	34.09 (3.92)	-0.0003	(0.06)	-0.0075	(0.22)	0.0001	(0.22)	

Significance levels (based on clustering-robust standard errors): *1%. †5%. ‡10%. Dependent variable: achievement z-score. Numbers in parenthesis represent robust absolute t statistics. In order to simplify the presentation, coefficients significant at the 10 percent level or better are in bold. Controlling for pupil fixed effects.

Table 5: **Robustness check for the Pupil Fixed Effect estimation**

Country		(1)	(2)	(3)	(4)
	Class size Mean (SD) 24.16 (4.66)	With ability variable		With teacher controls	
		Coef.	t Stat	Coef.	t Stat
Australia	20.77 (4.15)	0.0127*	(2.83)	0.0132*	(2.90)
Belgium (Fl.)	27.30 (5.55)	-0.0128†	(2.22)	-0.0146*	(2.51)
England	39.59 (4.06)	0.0007	(0.18)	-0.0005	(0.12)
Hong-Kong	22.05 (3.65)	0.0010	(0.29)	0.0017	(0.46)
Japan	35.57 (3.86)	n	n	n	n
Korea	37.16 (5.26)	0.0037	(0.87)	0.0024	(0.51)
Netherlands	26.01 (3.74)	-0.0029	(0.49)	-0.0032	(0.51)
New Zealand	25.88 (4.90)	-0.0019	(0.37)	-0.0022	(0.39)
Norway	26.03 (3.82)	n	n	n	n
Scotland	24.08 (7.31)	0.0075*	(2.57)	0.0060†	(2.09)
Sweden	21.13 (7.19)	0.0067*	(3.32)	0.0087*	(3.83)
Taiwan	37.24 (5.47)	n	n	n	n
USA	24.76 (6.81)	0.0006	(0.52)	0.0007	(0.62)
Bulgaria	23.15 (5.20)	-0.0146†	(2.44)	-0.0150†	(2.47)
Cyprus	25.65 (4.17)	0.0018	(0.86)	0.0023	(1.07)
Estonia	29.22 (6.92)	-0.0001	(0.05)	-0.0005	(0.37)
Hungary	23.74 (5.56)	-0.0052†	(2.38)	-0.0053†	(2.40)
Latvia	29.40 (12.29)	0.0022*	(2.71)	0.0018†	(2.15)
Lithuania	26.39 (4.52)	-0.0773‡	(1.67)	-0.0801‡	(1.72)
Macedonia	28.14 (5.19)	0.0039	(1.01)	0.0062	(1.58)
Moldova	25.78 (6.48)	0.0108*	(2.52)	0.0106*	(2.47)
Romania	25.06 (5.45)	-0.0010	(0.49)	-0.0017	(0.82)
Russian Fed.	24.40 (5.22)	n	n	n	n
Serbia	26.66 (6.29)	-0.0027†	(2.28)	-0.0030†	(2.44)
Slovakia	25.80 (5.08)	n	n	n	n
Slovenia	22.05 (4.10)	-0.0244‡	(1.69)q	-0.0240‡	(1.66)
Armenia	31.79 (11.66)	-0.0022*	(3.51)	-0.0025*	(3.86)
Bahrain	30.70 (5.16)	-0.0899‡	(1.70)q	-0.0840	(1.55)q
Botswana	37.13 (4.84)	n	n	n	n
Chile	35.46 (7.32)	-0.0157‡	(1.66)	-0.0250*	(2.65)
Egypt	37.40 (7.86)	-0.0026	(1.08)	-0.0012	(0.47)
Ghana	38.36 (11.76)	n	n	n	n
Indonesia	40.57 (5.92)	n	n	n	n
Iran	29.70 (6.78)	0.0004	(0.14)	0.0022	(0.81)
Jordan	26.37 (8.23)	-0.0084†	(1.97)	-0.0127*	(2.66)
Lebanon	29.16 (7.90)	0.0009	(0.38)	0.0013	(0.48)
Malaysia	37.10 (5.10)	n	n	n	n
Morocco	42.04 (13.74)	0.0191	(0.74)	0.0164	(0.62)
Palestine N.A.	39.52 (8.63)	-0.0053*	(3.13)	-0.0046†	(2.50)
Philippines	54.75 (10.81)	-0.0000	(0.02)	-0.0021	(0.80)
Saudi Arabia	29.23 (8.46)	n	n	n	n
South Africa	45.04 (14.68)	-0.1186†	(1.91)l	-0.1084‡	(1.71)l
Syrian A.R.	37.73 (9.42)	0.0030	(0.58)	-0.0364‡	(0.23)q
Tunisia	34.09 (3.92)	-0.0001	(0.01)	0.0007	(0.16)

Significance levels (based on clustering-robust standard errors): *1%. †5%. ‡10%. Dependent variable: achievement z-score. Numbers in parenthesis represent robust absolute t statistics. In order to simplify the presentation, coefficients significant at the 10 percent level or better are in bold. Controlling for pupil fixed effects. For some countries, we show the quadratic relation instead of the simple linear form (marked by a 'q') because coefficients from linear estimation are not statistically significant.

Table 6: Pupil Fixed Effect Estimation with IV technique

Country		(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Class size Mean (SD)	Linear Form		Quadratic Form				
		Coef.	t Stat	Linear Coef.	t Stat	Quadratic Coef.	t Stat	Turning point
Australia	24.16 (4.66)	0.0210*	(4.27)	0.5474‡	(1.61)	-0.0116	(1.57)	23.7
Belgium (Fl.)	20.77 (4.15)	-0.0303*	(3.89)	-0.4583*	(2.74)	0.0112*	(2.64)	
England	27.30 (5.55)	0.0007	(0.15)	0.0319	(0.37)	-0.0005	(0.37)	
Hong-Kong	39.59 (4.06)	-0.0001	(0.03)	0.1087	(0.83)	-0.0018	(0.81)	
Italy	22.05 (3.65)	n	n	n	n	n	n	
Japan	35.57 (3.86)	n	n	n	n	n	n	
Korea	37.16 (5.26)	0.0010	(0.25)	0.2100	(0.96)	-0.0028	(0.96)	
Netherlands	26.01 (3.74)	na	na	na	na	na	na	
New Zealand	25.88 (4.90)	-0.0059	(0.91)	-0.1481	(0.66)	0.0030	(0.65)	
Norway	26.03 (3.82)	n	n	n	n	n	n	
Scotland	24.08 (7.31)	0.0093*	(3.26)	0.0554‡	(2.28)	-0.0008‡	(2.04)	33.7
Sweden	21.13 (7.19)	0.0058*	(2.87)	0.0081	(0.66)	-0.0000	(0.20)	
Taiwan	37.24 (5.47)	n	n	n	n	n	n	
USA	24.76 (6.81)	0.0009	(0.75)	-0.0009	(0.04)	0.0000	(0.09)	
Bulgaria	23.15 (5.20)	-0.0147‡	(2.45)	-0.0652*	(2.87)	0.0012*	(2.81)	27.6
Cyprus	25.65 (4.17)	0.0026	(0.96)	0.0635	(1.29)	-0.0015	(1.29)	
Estonia	29.22 (6.92)	-0.0001	(0.06)	0.0028	(0.65)	-0.0000	(0.84)	
Hungary	23.74 (5.56)	-0.0092*	(3.64)	-0.1483*	(3.35)	0.0031*	(3.26)	24.0
Latvia	29.40 (12.29)	0.0019‡	(2.37)	0.0012	(0.28)	0.0000	(0.20)	
Lithuania	26.39 (4.52)	-0.0034	(1.36)	-0.0921‡	(1.82)	0.0015‡	(1.80)	30.9
Macedonia	28.14 (5.19)	0.0048	(1.22)	-0.0160	(0.59)	0.0003	(0.79)	
Moldova	25.78 (6.48)	0.0115*	(2.78)	0.0394‡	(2.18)	-0.0004‡	(1.68)	46.5
Romania	25.06 (5.45)	0.0007	(0.21)	0.0206	(0.89)	-0.0003	(0.98)	
Russian Fed.	24.40 (5.22)	n	n	n	n	n	n	
Serbia	26.66 (6.29)	-0.0027‡	(2.12)	0.0088	(0.85)	-0.0001	(1.12)	
Slovakia	25.80 (5.08)	n	n	n	n	n	n	
Slovenia	22.05 (4.10)	-0.0012	(0.42)	-0.0338‡	(1.71)	0.0010‡	(1.68)	16.8
Armenia	31.79 (11.66)	-0.0021*	(3.01)	0.0015	(0.18)	-0.0000	(0.45)	
Bahrain	30.70 (5.16)	0.0050	(0.84)	0.9047	(1.02)	-0.0142	(1.02)	
Botswana	37.13 (4.84)	n	n	n	n	n	n	
Chile	35.46 (7.32)	-0.0248‡	(2.06)	0.0125	(0.16)	-0.0005	(0.43)	
Egypt	37.40 (7.86)	na	na	na	na	na	na	
Ghana	38.36 (11.76)	n	n	n	n	n	n	
Indonesia	40.57 (5.92)	n	n	n	n	n	n	
Iran	29.70 (6.78)	na	na	na	na	na	na	
Jordan	26.37 (8.23)	na	na	na	na	na	na	
Lebanon	29.16 (7.90)	na	na	na	na	na	na	
Malaysia	37.10 (5.10)	n	n	n	n	n	n	
Morocco	42.04 (13.74)	n	n	n	n	n	n	
Palestine	39.52 (8.63)	-0.0054*	(2.94)	-0.0079	(0.37)	0.0000	(0.12)	
Philippines	54.75 (10.81)	na	na	na	na	na	na	
Saudi Arabia	29.23 (8.46)	n	n	n	n	n	n	
South Africa	45.04 (14.68)	na	na	na	na	na	na	
Syrian A.R.	37.73 (9.42)	na	na	na	na	na	na	
Tunisia	34.09 (3.92)	na	na	na	na	na	na	

Significance levels (based on clustering-robust standard errors): *1%. †5%. ‡10%. Dependent variable: achievement z-score. Numbers in parenthesis represents robust absolute t statistics. In order to simplify the presentation, coefficients significant at the 10 percent level or better are in bold. The instrument used is the average class-size in the subject across all grades in the school. For some countries, we cannot perform this estimation because only one class per school is tested (countries marked by an asterisk: 'na' means 'not applicable'). See table A.5 for background data and table A.6 for first-stage estimates.

Table 7: Synthesis of results

Country	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
	Class size	WLS		WLS RS		SFE		PFE		PFE IV	
		Lin.	Quad.	Lin.	Quad.	Lin.	Quad.	Lin.	Quad.	Lin.	Quad.
Australia	26.16 (4.66)	0.16		- ns		0.08		0.06		-0.33	23.7
Belgium (Fl.)	20.77 (4.15)	0.18		0.16		0.05	18.1	-0.06		-0.13	
England	27.30 (5.55)	0.22		0.51	14.9	0.24	40.7	- ns		+ ns	
Hong-Kong	39.59 (4.06)	0.37	20.9	0.41	25.2	+ ns		+ ns		- ns	
Italy	22.05 (3.65)	0.04		0.02	21.8	-0.05	24.3	n		n	
Japan	35.57 (3.86)	0.18	29.5	0.38	32	+ ns		n		n	
Korea	37.16 (5.26)	- ns		- ns		+ ns		+ ns		+ ns	
Netherlands	26.01 (3.74)	0.23		0.28	11.5	- ns		- ns		- ns	
New Zealand	25.88 (4.90)	0.16		0.14		0.09	20.9	- ns		- ns	
Norway	26.03 (3.82)	-0.04		n		+ ns		n		n	
Scotland	24.08 (7.31)	0.33		0.49	13	0.12	96.4	0.05		0.12	33.7
Sweden	21.13 (7.19)	0.16		-0.06		0.22	35.9	0.06	42	0.04	
Taiwan	37.24 (5.47)	0.10	31.1	0.08	31.5	+ ns		n		n	
USA	24.76 (6.81)	+ ns		-0.04		0.03		+ ns		+ ns	
Bulgaria	23.15 (5.20)	-0.11		-0.09		-0.04	26.1	-0.06	27.7	-0.04	27.6
Cyprus	25.65 (4.17)	- ns		+ ns		- ns		+ ns		+ ns	
Estonia	29.22 (6.92)	0.03		0.08		0.07	72.2	- ns		- ns	
Hungary	23.74 (5.56)	0.08	18.9	0.20	17.7	- ns		-0.03		-0.05	
Latvia	29.40 (12.29)	0.04		+ ns		0.03		0.02		0.02	
Lithuania	26.39 (4.52)	+ ns		+ ns		0.02		- ns		-0.05	30.9
Macedonia	28.14 (5.19)	-0.03		- ns		0.14	40.5	+ ns		+ ns	
Moldova	25.78 (6.48)	0.05		0.08		+ ns		0.07		0.12	46.5
Romania	25.06 (5.45)	-0.06		-0.03	27.2	0.06	31.2	- ns		+ ns	
Russian Fed.	24.40 (5.22)	0.01	23.4	-0.05		- ns		n		n	
Serbia	26.66 (6.29)	0.01	39.2	- ns		- ns		-0.02		-0.02	
Slovakia	25.80 (5.08)	0.06		0.12	22.3	0.11		n		n	
Slovenia	22.05 (4.10)	0.04	21.6	0.03	21.9	-0.03		0.03	16.8	0.05	16.8
Armenia	31.79 (11.66)	-0.06		-0.08		- ns		-0.03		-0.02	
Bahrain	30.70 (5.16)	-0.10		-0.11		-0.04	35.3	-0.02	31.7	+ ns	
Botswana	37.13 (4.84)	-0.09		-0.11		+ ns		n		n	
Chile	35.46 (7.32)	- ns		-0.04		-0.16		-0.16		-0.18	
Egypt	37.40 (7.86)	-0.02		0.00	42.7	- ns		- ns		n	
Ghana	38.36 (11.76)	0.11		0.09		- ns		n		n	
Indonesia	40.57 (5.92)	0.07		0.08		+ ns		n		n	
Iran	29.70 (6.78)	- ns		-0.03	27.8	- ns		+ ns		n	
Jordan	26.37 (8.23)	-0.10		-0.15	45.3	- ns		-0.07		n	
Lebanon	29.16 (7.90)	0.04		0.05		+ ns		+ ns		n	
Malaysia	37.10 (5.10)	0.04	36.2	0.05	35.7	0.22		n		n	
Morocco	42.04 (13.74)	- ns		- ns		+ ns		n		n	
Palestine.	39.52 (8.63)	-0.08		-0.06		-0.05		-0.05		-0.05	
Philippines	54.75 (10.81)	-0.09		-0.16	67.4	- ns		- ns		n	
Saudi Arabia	29.23 (8.46)	-0.04		-0.05		-0.13	33.7	n		n	
South Africa	45.04 (14.68)	-0.08		-0.10		- ns		-0.02		n	
Syrian A.R.	37.73 (9.42)	-0.06		-0.20		- ns		+ ns		n	
Tunisia	34.09 (3.92)	+ ns		- ns		+ ns		- ns		n	

Note : Coefficients present in the columns (2), (4), (6), (8) and (10) represent the effect, on pupil achievement, of a 1 SD reduction in class size. "ns" indicates that the relation is not statistically significant. A "+" or "-" before the "ns" indicates that the relation is respectively positive and negative but insignificant. "n" indicates that the regression is not possible for this country. The numbers in columns (3), (5), (7), (9) and (11) represent the class size above which the direction of the class-size/achievement relationship is reversed. See text for more information. WLS means Weighted Least Square regression (see table 1) ; WLS-RS: Weighted Least Squares with Reduced Sample (see table 2) ; SFE : School Fixed Effect (see table 3) ; PFE: Pupil Fixed Effect (see table 4) ; PFE-IV: Pupil Fixed Effect and Instrumental Variable technique (see table 6).

Table 8: Country characteristics and the existence of class-size effects

(1)	(2)	(3)		(4)		(5)		(6)	(7)		(8)			(9)	(10)
		Class size effect		Class size		Test Score		GDP per capita PPP\$	Expenditure per student		Teacher salary			Teachers' education	% of teachers reporting that large Class Size limits teaching
	No . of countries	WLS	PFE	Mean	SD	Mean	SD		Mean	Rel. To GDPpc	SEC Mean	SEC Rel. To GDPpc	PRI Rel. To GDPpc	% of Teachers with MA & training	
Developed countries	14	0.17	0.03	28.59	5.14	532	74	29347	7071	24	38465	1.36	2.10	22	18
EEC countries	13	0.01	0.00	26.26	6.39	494	79	10544	3486	24	10958	0.76	1.39	18	21
Developing countries	18	-0.03	-0.06	36.47	8.17	397	85	7653	1351	20	na	na	2.66	6	33
World mean (total)	45	0.05	-0.01	30.74	6.64	471	80	14846	4208	23	36173	1.31	2.12	15	24

(3): Weighted Least Squares estimates. *Source:* Table 7.; Pupil Fixed Effects estimates. *Source:* Table 7. Note that these 'effects' represent the number of SD by which student achievement would fall if there was a 1 SD increase in class-size. Thus, a 1 SD reduction in class-size in developing countries (by a large 8.2 pupils per class) would increase student achievement by 0.06 SD.

(4) – (5): Simple mean and SD of class size; Simple mean of mathematics and science test score. *Source:* Tables A.1 and A.3.

(6): GDP per capita, 2003 (PPP \$US). *Source:* UIS-Unesco (2008).

(7): Real government expenditure per pupil at secondary school, 2003. *Source:* UIS-Unesco (2008).

(8): Teacher salaries in lower secondary education, 2003 (Annual statutory salaries at 15 years experience in public institutions; PPP equivalent US dollars). *Source:* OECD (2004). And average real salary of primary school teachers relative to GDP per capita, 1990 (PPP-adjusted 1985 international dollars). *Source:* Lee and Barro (2001).

(9): % of TIMSS (grade 8 maths & science) teachers with MA and some training. *Source:* TIMSS teacher data. Some countries omitted or modified options in accordance with their national systems.

(10): % of TIMSS (grade 8 maths & science) teachers reporting that their teaching is limited "a great deal" by a high student/teacher ratio. *Source:* TIMSS teacher data.

Note: The developed country group includes 13 countries (Australia, Belgium, England, Hong-Kong, Italy, Japan, Republic of Korea, Netherlands, New Zealand, Norway, Scotland, Sweden, Taiwan and United States). The Eastern European (EEC) group includes 13 countries (Bulgaria, Cyprus, Estonia, Hungary, Latvia, Lithuania, Macedonia, Moldova, Romania, Russian Federation, Serbia, Slovenia and Sloak Republic). The developing countries group includes 18 countries (Armenia, Bahrain, Botswana, Chile, Egypt, Ghana, Indonesia, Iran, Jordan, Lebanon, Malaysia, Morocco, Palestinian N.A., Philippines, Saudi Arabia, South Africa, Syrian Arab Republic and Tunisia). The world group include all the 45 countries cited above. It should be noted that all these countries are not included in the CSE with Pupil Fixed Effect estimation, see table 4.

Appendix Table A1 : Mean of Raw (Non-Standardized) Scores

Country	Subjects	Average Scores					
		MAT	SCI	PHY	BIO	CHE	GEO
Australia	2	496 (78)	521 (74)				
Belgium (Fl.)	3	547 (67)			545 (56)		522 (63)
England	2	541 (84)	499 (78)				
Hong-Kong	2	587 (70)	557 (65)				
Italy	2	483 (76)	489 (78)				
Japan	2	568 (79)	551 (70)				
Korea	2	588 (83)	561 (70)				
Netherlands	3	541 (68)			543 (59)		541 (57)
New Zealand	2	489 (75)	517 (75)				
Norway	2	461 (70)	490 (71)				
Saudi Arabia	2	366 (81)	397 (73)				
Scotland	2	504 (73)	518 (74)				
Sweden	2	497 (72)	521 (77)				
Taiwan	2	585 (98)	576 (81)				
USA	2	505 (79)	530 (80)				
Bulgaria	5	484 (85)		480 (83)	469 (88)	484 (89)	487 (87)
Cyprus	4	461 (80)		449 (76)		440 (78)	446 (77)
Estonia	5	534 (69)		547 (64)	546 (64)	553 (59)	557 (72)
Hungary	5	531 (78)		540 (76)	538 (72)	562 (76)	538 (75)
Latvia	4	517 (72)		515 (62)	516 (65)	523 (73)	
Lithuania	5	507 (76)		520 (62)	520 (70)	536 (70)	515 (76)
Macedonia	5	437 (88)		457 (82)	450 (95)	466 (100)	441 (100)
Moldova	3	459 (79)		479 (75)			467 (79)
Romania	5	480 (88)		475 (81)	477 (90)	472 (94)	472 (94)
Russian Fed.	5	510 (76)		512 (73)	516 (75)	529 (80)	519 (78)
Serbia	5	479 (89)		474 (86)	468 (85)	475 (90)	473 (94)
Slovak Rep.	5	522 (87)		531 (76)	526 (79)	529 (79)	537 (86)
Slovenia	4	494 (70)		508 (58)	521 (70)	534 (70)	
Armenia	5	480 (83)		482 (74)	454 (85)	467 (102)	464 (91)
Bahrain	2	405 (76)	439 (74)				
Botswana	2	365 (70)	361 (86)				
Chile	2	399 (85)	421 (86)				
Egypt	2	440 (97)	450 (104)				
Ghana	2	283 (91)	258 (120)				
Indonesia	3	419 (88)		439 (84)	433 (75)		
Iran	2	415 (75)	455 (73)				
Jordan	2	422 (88)	472 (89)				
Lebanon	4	439 (66)		434 (80)		448 (89)	376 (106)
Malaysia	2	508 (74)	510 (67)				
Morocco(1)	2	388 (68)	399 (69)				
Palestine	2	391 (91)	437 (92)				
Philippines	2	379 (85)	379 (99)				
South Africa	2	269 (105)	249 (129)				
Syrian A.R.	2	361 (78)	413 (79)				
Tunisia	2	411 (60)	402 (60)				

Note : MAT = maths ; SCI = science ; PHY = physics ; CHE = chemistry ; BIO = biology; GEO = geography.
For Lebanon, score in last column is for "Life and Earth Science" and not geography.

Appendix Table A2: Descriptive statistics on control variables (continued on next page)

	ARM	AUS	BFL	BGL	BHR	BWA	CHL	CYP	EGY	ENG	EST	GHA	HKG	HUN	IDN	IRN	ITA	JOR	JPN	KOR	LBN	LIT	LTV
Age (in months)	245	180	176	186	176	195	184	172	179	178	189	193	179	183	180	181	174	174	183	183	181	186	187
Sex (1= Female)	53	51	53	49	49	51	50	49	48	49	49	45	50	51	51	43	49	52	49	48	57	50	51
Speaks national language at home																							
1=Always	80	84	78	82	66	5	87	78	57	90	89	23	77	95	21	55	94	72	94	70	5	88	78
2=Often	16	10	11	11	15	6	9	13	17	8	8	11	15	4	12	10	2	13	4	29	12	10	15
3=Sometimes	3	5	8	6	14	80	4	6	23	2	2	61	7	0	58	21	3	12	1	1	69	2	5
4=Never	1	1	3	1	4	9	0	2	3	0	1	5	1	0	10	14	1	3	0	0	14	1	2
Books at home																							
Less than 10	14	5	12	12	11	50	21	11	26	12	2	33	28	4	33	39	13	23	13	15	21	9	3
11-25	24	10	24	14	26	29	36	27	35	17	11	34	28	14	44	30	28	33	22	10	35	27	11
26-100	29	29	35	25	31	13	29	35	22	29	22	16	27	29	19	18	25	27	32	32	27	36	31
101-200	14	24	16	18	14	4	9	16	9	19	18	6	9	22	3	5	14	8	17	23	8	16	26
More than 200	19	31	12	31	17	4	6	11	8	23	47	11	9	32	1	8	20	9	17	20	8	12	30
Possess a computer	19	95	96	41	81	15	44	83	35	95	69	25	97	75	17	28	84	40	82	2	61	51	47
Mother born in country	93	70	89	98	81	95	99	89	89	92	81	91	50	97	97	97	93	75	99	99	93	95	81
Father born in country	94	69	88	98	84	93	98	91	87	91	80	89	53	98	97	97	94	71	99	99	94	93	80
Student born in country	96	87	94	95	89	96	96	88	79	95	96	69	74	97	88	99	95	75	99	99	88	96	96
Parents' highest education																							
University	52	29	27	33	35	9	21	29	45	25	42	11	13	37	9	11	21	33	44	36	20	39	46
Technical post-second.	21	27	27	35	6	13	10	14	Na	12	38	17	12	na	6	10	5	15	17	15	23	31	na
Upper Secondary	24	26	30	26	22	15	31	36	9	54	18	24	35	49	25	14	39	32	36	40	19	28	32
Lower Secondary	2	15	11	5	20	20	28	14	21	9	2	36	25	13	21	22	29	12	2	7	15	2	22
No more than Primary	1	3	4	1	17	43	10	7	25	na	0	12	15	0	39	43	5	8	na	2	24	1	0
City size																							
More than 500,000	17	32	2	13	7	2	25	na	26	17	4	19	36	10	23	27	13	22	24	48	21	8	30
100,001-500,000	9	20	8	22	4	3	30	23	13	22	27	13	53	15	6	19	13	7	39	25	13	23	8
50,001-100,000	7	12	16	17	11	11	14	21	12	25	13	7	3	8	12	11	15	13	12	9	8	7	6
15,001-50,000	15	19	55	15	33	26	15	14	17	13	14	10	7	25	18	10	24	18	17	11	19	20	13
3,001-15,000	35	12	18	17	35	34	9	43	24	19	27	35	na	25	34	15	30	30	8	5	25	24	24
Less than 3,000	17	5	Na	16	10	24	7	na	8	4	14	17	na	17	8	18	4	11	0	2	14	17	19
Computers in school	10	121	47	12	17	15	17	23	10	200	25	1	93	20	11	1	20	16	40	54	18	22	25

Note : ARM = Armenia ; AUS = Australia ; BFL = Belgium (Flemish) ; BGL = Bulgaria ; BHR = Bahrain ; BWA = Botswana ; CHL = Chile ; CYP = Cyprus ; EGY = Egypt ; ENG = England ; EST = Estonia ; GHA = Ghana ; HKG = Hong-Kong ; HUN = Hungary ; IDN = Indonesia ; IRN = Iran ; ITA = Italy ; JOR = Jordan ; JPN = Japan ; KOR = Korea ; LBN = Lebanon ; LIT = Lithuania ; LTV = Latvia.

Appendix Table A2: Descriptive statistics on control variables (continued)

	MAR	MKD	MOL	MYS	NLD	NOR	NZL	PHL	PSE	ROM	RUS	SAU	SCO	SER	SVK	SLO	SWE	SYR	TUN	TWN	USA	ZAF
Age (in months)	189	183	186	186	179	174	182	186	177	187	178	175	172	185	179	173	186	176	185	178	178	194
Sex (1= Female)	53	50	51	58	49	50	50	57	52	52	50	45	49	49	49	50	51	54	53	48	52	50
Speaks national language at home																						
1=Always	31	88	68	50	83	84	80	2	74	86	88	100	92	93	80	81	80	52	50	44	84	20
2=Often	17	4	19	15	12	11	13	4	11	9	9	0	5	5	13	11	11	12	13	36	10	9
3=Sometimes	42	5	12	28	4	3	6	80	12	4	2	0	2	2	6	6	7	31	28	19	5	57
4=Never	10	2	1	7	1	1	1	14	2	2	0	0	1	0	1	2	2	5	9	1	1	14
Books at home																						
Less than 10	30	16	22	17	9	6	8	43	26	16	3	23	15	20	4	7	7	26	23	17	12	44
11-25	39	39	36	40	18	11	15	34	36	24	15	34	20	38	21	26	14	41	45	24	17	32
26-100	22	29	25	28	32	32	30	16	25	31	33	24	30	27	40	39	28	22	22	30	28	14
101-200	4	8	9	9	20	22	21	4	6	15	27	9	17	9	20	15	20	6	7	14	18	5
More than 200	6	8	8	6	21	28	25	3	7	14	22	10	18	6	15	13	30	5	4	15	24	5
Possess a computer	21	44	20	57	98	97	90	21	42	36	33	54	91	45	70	87	98	30	22	91	93	36
Mother born in country	94	94	91	96	86	89	76	98	91	99	90	87	94	85	96	85	78	94	99	98	86	92
Father born in country	92	94	89	97	85	88	74	97	93	99	88	86	95	87	96	84	78	92	99	98	85	90
Student born in country	90	92	88	95	93	92	86	97	76	98	91	91	95	89	97	96	89	92	99	98	91	67
Parents' highest education																						
University	16	24	38	11	23	68	29	18	28	20	44	26	48	20	39	25	49	19	10	17	57	10
Technical post-second.	na	19	17	20	32	15	30	22	12	17	27	Na	26	68	na	31	17	na	12	11	9	13
Upper Secondary	16	41	20	27	42	11	33	33	36	47	23	12	21	2	59	35	22	13	16	46	25	29
Lower Secondary	16	11	15	24	na	4	5	13	18	10	6	18	5	9	1	8	9	30	17	21	6	19
No more than Primary	52	5	10	18	3	2	3	14	6	6	0	44	0	1	0	1	4	38	44	6	3	29
City size																						
More than 500,000	3	13	20	10	4	8	18	13	7	16	24	33	8	9	3	3	9	12	2	25	13	11
100,001-500,000	2	5	4	16	25	16	19	15	13	24	19	10	13	16	9	12	22	10	3	36	12	8
50,001-100,000	10	19	na	17	22	10	9	25	21	10	11	5	12	13	14	1	14	7	13	18	14	11
15,001-50,000	24	17	16	26	43	34	18	16	18	15	15	9	23	30	29	13	26	23	44	18	30	24
3,001-15,000	39	37	38	30	5	29	24	17	32	14	15	15	39	22	23	40	24	44	38	3	22	34
Less than 3,000	22	9	21	2	na	3	12	15	9	21	17	27	5	10	22	30	5	4	na	0	9	12
Computers in school	8	5	21	16	65	23	90	18	9	12	11	7	196	8	9	16	32	5	8	74	91	7

Note : MAR = Morocco ; MKD = Macedonia ; MOL = Moldova ; MYS = Malaysia ; NLD = Netherlands ; NOR = Norway ; NZL = New Zealand ; PHL = Philippines ; PSE = Palestinian Nat. Auth. ; ROM = Romania ; RUS = Russian Federation ; SAU = Saudi Arabia ; SCO = Scotland ; SER = Serbia ; SVK = Slovak Republic ; SLO = Slovenia ; SWE = Sweden ; SYR = Syrian Arab Republic ; TUN = Tunisia ; TWN = Taiwan ; USA = United States of America ; ZAF = South Africa. "na" mean "not applicable"

Appendix Table A3 : Analysis of grouping classes with same ability

Country	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Mean Class Size (Standard Deviation)		Percentage of classes grouped by ability		Sample (Number of Observations)		
	Full sample	Reduced	Math.	Science	Total	Full	Reduced
Australia	26.16 (4.66)	26.04 (3.97)	49.04	16.25	10224	7606	3633
Belgium (Flemish)	20.77 (4.15)	21.12 (3.84)	30.65	25.03	15228	12605	8211
England	27.30 (5.55)	26.83 (5.44)	61.32	54.28	7221	2975	708
Hong-Kong	39.59 (4.06)	39.96 (3.43)	34.66	11.94	9944	9629	5641
Italy	22.05 (3.65)	22.17 (3.74)	10.55	08.60	8598	8240	7112
Japan	35.57 (3.86)	35.69 (3.85)	11.87	01.33	9712	9678	8398
Korea	37.16 (5.26)	36.85 (4.52)	64.51	46.08	11237	6389	2047
Netherlands	26.01 (3.74)	26.13 (3.75)	18.08	10.83	7320	7320	5594
New Zealand	25.88 (4.90)	26.44 (4.81)	35.41	18.92	7626	6787	3964
Norway	26.03 (3.82)	n	n	n	8266	7175	n
Scotland	24.08 (7.31)	24.21 (6.54)	74.48	10.65	7791	4494	681
Sweden	21.13 (7.19)	23.02 (5.79)	66.45	09.87	10338	5441	1835
Taiwan	37.24 (5.47)	37.32 (5.57)	15.41	11.09	10808	10579	8879
USA	24.76 (6.81)	24.94 (6.95)	53.85	14.41	17905	13923	5138
Bulgaria	23.15 (5.20)	22.96 (5.21)	43.30	39.80	15275	15275	7187
Cyprus	25.65 (4.17)	25.90 (4.31)	26.52	14.42	14842	11672	7517
Estonia	29.22 (6.92)	29.19 (7.07)	32.71	09.02	16437	15687	10470
Hungary	23.74 (5.56)	22.74 (5.51)	45.07	15.48	14814	14652	7348
Latvia	29.40 (12.29)	28.63 (11.77)	20.15	05.55	11016	11016	7834
Lithuania	26.39 (4.52)	26.62 (4.28)	39.77	31.40	20805	20805	10979
Macedonia	28.14 (5.19)	28.08 (5.10)	12.00	08.35	17612	17612	14957
Moldova	25.78 (6.48)	24.90 (6.39)	31.61	30.98	9401	7978	3364
Romania	25.06 (5.45)	25.47 (5.65)	36.82	33.75	18797	18797	10291
Russian Federation	24.40 (5.22)	24.82 (4.88)	42.95	31.79	23322	22461	12358
Serbia	26.66 (6.29)	26.42 (7.04)	32.74	33.87	18879	18711	11746
Slovakia	25.80 (5.08)	26.03 (5.16)	29.63	15.18	17162	1712	11788
Slovenia	22.05 (4.10)	22.86 (3.08)	30.14	10.68	13234	13102	7863
Armenia	31.79 (11.66)	32.94 (12.16)	19.39	18.93	19460	19460	12308
Bahrain	30.70 (5.16)	30.09 (5.74)	37.04	25.53	8398	8076	4570
Botswana	37.13 (4.84)	36.98 (4.89)	1.88	02.44	10300	9001	7868
Chile	35.46 (7.32)	36.12 (7.07)	16.77	15.65	13992	10806	8544
Egypt	37.40 (7.86)	37.42 (7.75)	31.19	27.65	14190	13760	9196
Ghana	38.36 (11.76)	38.24 (11.20)	19.00	17.98	10466	8213	6389
Indonesia	40.57 (5.92)	40.43 (5.74)	17.42	14.19	16707	15494	12354
Iran	29.70 (6.78)	29.61 (6.40)	61.29	70.32	9884	9656	2563
Jordan	26.37 (8.23)	36.58 (8.29)	9.03	11.43	8978	8795	7718
Lebanon	29.16 (7.90)	29.92 (7.85)	19.88	14.17	8895	8895	7086
Malaysia	37.10 (5.10)	36.46 (4.93)	45.87	39.94	10628	10500	5468
Morocco	42.04 (13.74)	43.39 (14.89)	13.83	31.41	2943	1184	833
Palestine	39.52 (8.63)	39.90 (8.37)	7.68	05.33	10714	10268	9084
Philippines	54.75 (10.81)	53.78 (10.72)	34.09	38.80	13834	13482	7981
Saudi Arabia	29.23 (8.46)	29.28 (8.50)	15.84	14.69	8590	7342	6051
South Africa	45.04 (14.68)	46.00 (15.17)	11.96	09.42	17949	12336	9250
Syrian A.R.	37.73 (9.42)	38.77 (9.16)	7.70	12.58	6102	4564	3612
Tunisia	34.09 (3.92)	33.73 (4.27)	44.63	45.43	9862	9034	4727

Appendix Table A4: Weighted Least Squares with Instrumental Variable technique (WLS-IV)

Country	(1)	(2)	(5)	(6)	(7)	(8)	(9)
	Linear Form		Quadratic Form				
	Coef.	t Stat	Coef.	t Stat	Coef.	t Stat	Turning point
Australia	-0.0139	(1.14)	0.5466*	(3.85)	-0.0107*	(3.64)	25.6
Belgium (Fl.)	-0.1808†	(2.03)	4.9243	(1.25)	-0.1266	(1.23)	
England	0.0293†	(2.42)	-0.0224	(0.18)	0.0009	(0.44)	
Hong-Kong	-0.0018	(0.12)	-0.6670*	(3.71)	0.0107*	(4.10)	31.3
Italy	-0.0017	(0.18)	-0.5788‡	(1.53)	0.0136‡	(1.56)	21.3
Japan	-0.0826*	(6.38)	2.74149*	(2.89)	-0.0428*	(2.98)	32.0
Korea	-0.0087	(0.55)	0.1337	(0.24)	-0.0019	(0.25)	
Netherlands	-0.0103	(0.65)	1.4972*	(3.71)	-0.0291*	(3.57)	25.7
New Zealand	-0.0160‡	(1.55)	-4.5588‡	(1.67)	0.0885‡	(1.67)	25.8
Norway	-0.0114	(1.16)	-0.0108	(0.16)	-0.0000	(0.01)	
Scotland	0.0384*	(3.40)	0.0346	(1.06)	0.0001	(0.17)	
Sweden	0.0107‡	(1.60)	-0.0062	(0.25)	0.0004	(0.89)	
Taiwan	-0.0609*	(2.70)	-0.2858*	(3.68)	0.0041*	(3.94)	35.1
USA	-0.0303*	(5.38)	-0.2256*	(5.32)	0.0042*	(5.32)	27.0
Bulgaria	-0.0184*	(2.82)	0.2034	(0.66)	-0.0047	(0.74)	
Cyprus	0.0041	(0.74)	0.1443	(1.20)	-0.0032	(1.22)	
Estonia	0.0080†	(1.90)	0.0172	(1.59)	-0.0002	(1.34)	
Hungary	0.0017	(0.25)	0.3738†	(1.97)	-0.0076†	(1.90)	24.7
Latvia	0.0149*	(2.31)	0.0394†	(2.01)	-0.0004‡	(1.85)	49.2
Lithuania	-0.0060	(0.76)	-0.1048	(1.08)	0.0021	(1.10)	
Macedonia	0.0084	(0.85)	-0.5428	(1.24)	0.0101	(1.23)	
Moldova	0.0215	(0.92)	0.1566	(0.57)	-0.0025	(0.54)	
Romania	-0.0317*	(3.71)	-0.2492*	(3.26)	0.0047*	(3.17)	26.3
Russian Fed.	-0.0083†	(2.09)	-1.1316	(1.41)	0.0249	(1.40)	
Serbia	0.0066	(1.07)	0.0166	(0.97)	-0.0002	(0.90)	
Slovakia	0.0085*	(2.37)	-0.3454‡	(1.67)	0.0073‡	(1.73)	23.8
Slovenia	-0.0068	(0.61)	0.1703	(0.43)	-0.0042	(0.44)	
Armenia	-0.0283†	(1.91)	0.1814	(1.30)	-0.0025	(1.34)	
Bahrain	-0.0058	(0.56)	0.2730	(1.14)	-0.0047	(1.22)	
Botswana	-0.0374*	(2.76)	-2.3064	(1.14)	0.0319	(1.13)	
Chile	-0.0217*	(6.27)	-0.5443*	(4.97)	0.0083*	(4.99)	32.8
Egypt	-0.0175*	(3.23)	-0.1609*	(2.82)	0.0018*	(2.77)	45.0
Ghana	0.0164	(1.40)	0.1712	(0.61)	-0.0022	(0.58)	
Indonesia	-0.0463*	(3.25)	1.0878*	(3.17)	-0.0147*	(3.15)	37.1
Iran	-0.0066	(1.41)	-0.0667	(1.38)	0.0010	(1.37)	
Jordan	-0.0120*	(3.60)	-0.0325	(0.91)	0.0003	(0.63)	
Lebanon	-0.0504*	(2.39)	13.444	(0.08)	-0.2044	(0.08)	
Malaysia	-0.0564*	(5.87)	-1.1750*	(5.77)	0.0160*	(5.77)	36.7
Morocco	-0.0253†	(2.53)	-0.1347†	(2.49)	0.0015†	(2.44)	45.6
Palestine	-0.0236*	(4.15)	-1.9402	(0.69)	0.0279	(0.69)	
Philippines	-0.0753*	(10.67)	-0.8113*	(5.74)	0.0072*	(5.73)	56.4
Saudi Arabia	-0.0186‡	(1.81)	-0.0559	(1.29)	0.0008	(1.13)	
South Africa	-0.0163*	(6.94)	-0.0727*	(4.82)	0.0006*	(4.42)	61.6
Syrian A.R.	0.0109	(1.31)	0.1173†	(1.76)	-0.0016†	(1.81)	37.1
Tunisia	0.0188	(0.52)	-0.1003	(0.46)	0.0016	(0.47)	

Significance levels (based on clustering-robust standard errors): *1%. †5%. ‡10%. Dependent variable: achievement z-score. Controlling for student, school and family background variables. Numbers in parenthesis represents robust absolute t statistics. In order to simplify the presentation, coefficients significant at the 10 percent level or better are in bold. Instrument used for class-size is the teacher's response to the question 'to what extent does a high student/teacher ratio limit how you teach'? First stage estimates are not reported but are available.

Appendix Table A5: Background data for FE and FE IV regressions

Country	(1) <i>Number of similar Class size</i>	(2) <i>Total Number of Classes</i>	(3) <i>% Similar Class sizes</i>	(4) <i>Value of Class size variation between subjects</i>	(5) <i>Class size Mean (SD)</i>	(6) <i>Observations in table 3</i>	(7) <i>Pupils in table 3</i>
Australia	198	414	48	1.14 (2.13)	26.16 (4.66)	6668	3334
Belgium (Flemish)	384	506	76	0.28 (1.00)	20.77 (4.15)	12095	4314
England	124	260	48	1.84 (3.05)	27.30 (5.55)	2080	1040
Hong-Kong	194	262	74	0.46 (1.47)	39.59 (4.06)	9314	4657
Italy(a)	432	432	100	0.00 (0.00)			
Japan ^(a)	247	292	85	0.15 (0.53)			
Korea	160	298	54	0.76 (2.14)	37.16 (5.26)	4510	2255
Netherlands	179	309	58	0.44 (1.06)	26.13 (3.66)		
New Zealand	151	354	43	0.87 (1.55)	25.88 (4.90)	6142	3071
Norway ^(a)	276	358	77	0.16 (0.81)			
Scotland	107	310	35	2.99 (3.56)	24.08 (7.31)	2982	1491
Sweden	164	328	50	2.19 (3.50)	21.13 (7.19)	4660	2330
Taiwan ^(a)	296	300	99	0.00 (0.04)			
USA	332	914	36	2.28 (3.09)	24.76 (6.81)	11808	5904
Bulgaria	518	724	72	0.40 (1.49)	23.15 (5.20)	14374	3102
Cyprus	196	572	34	1.55 (2.56)	25.65 (4.17)	11598	3294
Estonia	429	604	71	0.62 (2.64)	29.22 (6.92)	15687	3855
Hungary	519	696	75	0.38 (1.58)	23.74 (5.56)	14626	3214
Latvia	225	531	42	3.67 (5.82)	29.40 (12.29)	10778	3318
Lithuania	625	1071	58	0.53 (1.70)	26.39 (4.52)	20765	4877
Macedonia	459	672	68	0.30 (0.86)	28.14 (5.19)	17580	3861
Moldova	234	398	59	1.18 (3.30)	25.78 (6.48)	7172	2961
Romania	533	812	66	0.43 (1.50)	25.06 (5.45)	18797	4104
Russian Federation ^(a)	938	1068	88	0.05 (0.22)			
Serbia	617	772	80	0.57 (2.87)	26.66 (6.29)	18547	4072
Slovakia ^(a)	604	715	84	0.17 (0.72)			
Slovenia	348	651	53	1.02 (2.10)	22.05 (4.10)	13080	3497
Armenia	206	856	24	4.85 (5.62)	31.79 (11.66)	19100	4992
Bahrain	130	294	44	0.62 (1.33)	30.70 (5.16)	7798	3899
Botswana ^(a)	193	292	66	0.26 (0.56)			
Chile	310	390	79	0.19 (1.06)	35.46 (7.32)	10090	5045
Egypt	325	434	75	0.85 (2.10)	37.48 (7.84)		
Ghana ^(a)	258	300	86	0.33 (1.38)			
Indonesia ^(a)	355	423	84	0.22 (1.22)			
Iran	283	362	78	0.62 (2.26)	29.56 (6.67)		
Jordan	201	280	72	0.37 (1.25)	36.39 (8.23)		
Lebanon	203	423	48	1.23 (2.76)	29.20 (7.88)		
Malaysia ^(a)	232	300	77	0.21 (0.50)			
Morocco ^(a)	128	150	85	1.84 (3.75)			
Palestine	209	290	72	0.94 (3.16)	39.43 (8.62)		
Philippines	205	274	75	0.73 (2.19)	54.82 (10.68)		
Saudi Arabia ^(a)	281	344	82	0.37 (1.40)			
South Africa	302	510	59	2.08 (5.16)	45.79 (15.46)		
Syrian A.R.	210	268	78	0.81 (2.89)	38.45 (9.16)		
Tunisia	149	300	50	0.85 (1.61)	34.09 (3.92)		

(a) Countries not included in the Pupil Fixed Effect Estimation, see text for more details.

Appendix Table A6: First stage estimates for FE IV regression

Country	(1)	(2)	(3)	(4)	(5)	(6)
	Linear Form			Quadratic Form		
	Coef.	t Stat	Instrument statistic	Coef.	t Stat	Instrument statistic
Australia	1.0039*	(58.66)	3440	0.0717‡	(1.74)	3
Belgium (Fl.)	0.7605*	(31.49)	992	0.0552*	(8.00)	64
England	0.9435*	(15.79)	249	0.0702*	(3.26)	11
Hong-Kong	1.0008*	(30.37)	922	-0.0579*	(4.04)	16
Italy	n	n	n	n	n	n
Japan	n	n	n	n	n	n
Korea	1.0618*	(46.42)	2155	-0.0697*	(2.62)	7
Netherlands	na	na	na	na	na	na
New Zealand	0.9985*	(25.77)	664	0.0582*	(3.19)	10
Norway	n	n	n	n	n	n
Scotland	1.0041*	(25.87)	669	0.1947*	(7.88)	62
Sweden	0.9770*	(62.32)	3883	0.3458*	(15.17)	230
Taiwan	n	n	n	n	n	n
USA	0.9824*	(121.9)	>10000	0.1149*	(8.22)	67
Bulgaria	0.9945*	(468.9)	>10000	0.6608*	(24.37)	594
Cyprus	0.9871*	(171.9)	>10000	0.0793*	(18.41)	339
Estonia	0.9911*	(113.6)	>10000	0.8470*	(30.52)	931
Hungary	0.9998*	(61.00)	3720	0.1253*	(9.17)	84
Latvia	0.9992*	(473.1)	>10000	0.7827*	(27.98)	783
Lithuania	0.9804*	(45.83)	2101	0.0912*	(5.11)	26
Macedonia	0.9906*	(920.2)	>10000	0.7493*	(33.12)	1097
Moldova	0.9910*	(572.8)	>10000	0.8134*	(51.62)	2665
Romania	0.9258*	(21.19)	449	0.1712*	(12.59)	159
Russian Fed.	n	n	n	n	n	n
Serbia	0.9693*	(105.8)	>10000	0.3573*	(9.25)	86
Slovakia	n	n	n	n	n	n
Slovenia	1.0000*	(164.5)	>10000	0.6748*	(23.16)	536
Armenia	1.0048*	(357.2)	>10000	0.1597*	(28.45)	809
Bahrain	0.9974*	(70.10)	4926	0.0156*	(2.52)	6
Botswana	n	n	n	n	n	n
Chile	1.0000*	(25.53)	652	-0.2437*	(3.27)	11
Egypt	na	na	na	na	na	na
Ghana	n	n	n	n	n	n
Indonesia	n	n	n	n	n	n
Iran	na	na	na	na	na	na
Jordan	na	na	na	na	na	na
Lebanon	na	na	na	na	na	na
Malaysia	n	n	n	n	n	n
Morocco	n	n	n	n	n	n
Palestine	na	na	na	na	na	na
Philippines	na	na	na	na	na	na
Saudi Arabia	n	n	n	n	n	n
South Africa	na	na	na	na	na	na
Syrian A.R.	na	na	na	na	na	na
Tunisia	na	na	na	na	na	na