

‘The effect of relative age in the first grade of primary school on long-term scholastic results: international comparative evidence using PISA 2003’.

Maresa Sprietsma\*

March 2006

*Preliminary version*

*Comments are welcome*

***Abstract***

In this paper, we estimate the effect of pupil’s relative age within the first grade of primary school on math, reading, and science test scores at age 15. The main objective is to evaluate the causal effect of relative age in the first grades of primary school on pupil’s test scores towards the end of compulsory schooling in 16 different countries. We use the national rule for admission to primary school to construct the predicted relative age of each pupil. We find that relative age at the start of primary school has a significant positive effect on test scores at the age of 15 in most countries. Moreover, we identify some of the channels through which the effect occurs.

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\* Université Catholique de Louvain, Place Montesquieu 3, 1348 Louvain-la-Neuve.

Tel : + 3210473461, Fax : +321047

Email address : [sprietsma@ires.ucl.ac.be](mailto:sprietsma@ires.ucl.ac.be)

*Acknowledgements:* I am grateful for the financial support from the Belgian French Community's program 'Action de Recherches Concertée' ARC 03/08-302. Thanks are due also to Prof. H.Oosterbeek for inspiring discussion and ideas.

## 1. Introduction

In this paper, we estimate the effect of pupil's relative age within the first grade of primary school on math and reading test scores at 15 years of age, that is, towards the end of compulsory schooling. In most countries there exists a compulsory school starting age, which obliges cohorts of e.g. 6 year old pupils to start school each year. If the admission rule into the first grade of primary school is respected, then pupils that end up in the same class should vary in age up to twelve months. Relative age is thus measured here on a scale of 1 to 12 months, and corresponds to the relative age of the pupil inside a given age cohort. The youngest pupil in the age cohort will have a relative age equal to 1, the eldest a relative age of 12. We will come back to the construction of our measure of relative age in detail in the second section.

There are several reasons why relative age may affect pupils performance. In the first grade of primary school, relative immaturity of the youngest pupils within their age cohort may disadvantage them in testing, because they are less able to perform the same tasks. But they may also benefit from positive peer effects from their older classmates. Furthermore, relative age in the first grade of primary school may also have a more lasting effect on performance. Indeed, recent research argues that early childhood maturity can influence long-term cognitive performance through ability grouping. According to Allen and Barnsley (1993), it might be difficult for teachers to differentiate between ability and maturity for children at school starting age. As a result, if children have different levels of maturity at the start of primary school and they are grouped according to observed ability, the youngest ones may be more likely to end up in low-performing peer groups. Over time, they risk learning less<sup>1</sup> and they may face a higher risk of repeating a grade in primary school (see e.g. Corman 2003). Moreover, even if there is no ability grouping, we think that teachers may hold lower expectations from the youngest pupils which may lower their self esteem and effort in the long run.

In this paper, we provide causal evidence on the long-term effect of relative age in the first grade of primary school on test scores at age 15. We also try to assess through which particular channels, for instance an increased risk of grade repetition, or selection into the vocational track, it occurs.

In educational research, causal analysis has been expanding only recently, triggering a vital concern for reliable identification strategies (Angrist, 2004). This paper fits into this development as we construct our measure of relative age using arguably exogenous national rules for admission to primary school and birth month. Several recent papers use the exogenous variation provided by administrative cut-off dates for starting school to instrument school starting age and time in school in education production functions (e.g. Angrist and Krueger 1992, Leuven et al., 2004, Bedard and Dhuey 2005).

All the papers that investigate pupil's test scores as their outcome variable find a positive effect of being older when you start school. This seems to be true whether

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<sup>1</sup> See Bedard and Dhuey (2005) for a the model of human capital accumulation that corresponds to the intuition developed here.

test scores are measured at kindergarten, the end of primary school or at the end of compulsory schooling. The 'age premium' seems to be decreasing over time, from  $\frac{3}{4}$  of a standard deviation in kindergarten (Datar 2004, Leuven et al. 2004), to about half a standard deviation at the end of primary school (Puhani and Weber 2005) and 20% of a standard deviation in secondary school (Frederiksson and Ökert 2005, Strom 2004). Fertig and Kluge (2005) on the other hand find no effect of age at school entry on grade retention or degree attainment in Germany. These studies respectively use data from the United States, the Netherlands, Germany, Sweden and Norway. Bedard and Dhuey (2006), using TIMSS<sup>2</sup> data from eleven OECD countries, find that there exist long lasting positive effects on student test scores in grade eight of being relatively older in grade one. These results seem to be more pronounced in countries with early tracking<sup>3</sup>. Furthermore, they find that relatively older children are more likely to attend college in New Zealand and the UK.

It is not possible to distinguish absolute versus relative age effects in this literature as children born e.g. just before the cut-off date to enter school will not only start school one year earlier than children born just after, but will also be the youngest of their class. There is no agreement as yet as to which effect dominates. Two of the mentioned papers comment on this issue, one concluding to the dominance of differences in maturity based on 25 questionnaires sent to primary school directors (Puhani and Weber), the other concluding to the dominance of absolute age effects based on peer age composition in grade 9 (Frederiksson and Ökert 2005).

We contribute to existing research by providing causal evidence on the effect of relative age in the first grade of primary school on test scores at age 15 in 16 different countries. Causality comes from the use of an exogenous measure of relative age. The objective is to evaluate whether relative age in the first grade of primary school has a significant long-term effect on reading and maths test scores. Indeed, the causal literature on the effect of school starting age, summarised above, studies only a limited number of northern European countries, and does not proceed to inter-country comparisons (with the exception of Bedard and Dhuey 2005). Furthermore, we investigate through which channels the effect of relative age occurs in each country.

We find that relative age does have a significant long-term impact on test scores, the eldest pupils having about twenty percent of a standard deviation higher test scores than the relatively youngest. Furthermore, according to our estimations, most of the effect of relative age comes from grade repetition and selection of pupils into different grades, as well as other forms of ability grouping. For instance, relatively younger pupils are more likely to repeat a grade, and grade repetition has a significant negative impact on test scores. Grade repetition is therefore considered a channel through which relative age affects test scores. We cannot distinguish relative from absolute age effects as such but will comment on the issue in the interpretation part.

In the following section, we explain how our measure of relative age was constructed and why is it exogenous. In the third section, we present the dataset and clarify the choice of included countries. Our empirical strategy is the topic of section 4. In Section 5 we present the results of our estimations.

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<sup>2</sup> Trends in International Mathematics and Science Study.

<sup>3</sup> Tracking refers to pupils being selected into different tracks at high school. Often there will be for instance a general versus a vocational track.

## **2. Relative age**

Our measure of relative age in the first grade of primary school is the relative age of a pupil inside his/her age cohort based on the national rule for admission to primary school (RELAGE). In most countries with high scolarisation rates, there exists a legal cut-off date for admission to primary school. For instance, children may enter grade one if they have reached a certain age before the start of the school year. We construct the relative age of the pupils using these national cut-off dates and information on pupil's birth month. If e.g. the cut-off date is January 1<sup>st</sup>, the pupil should have reached the age of 6 (the compulsory school starting age) before the 1<sup>st</sup> of January in order to start school that calendar year. Therefore, pupils born in December are the youngest in class, because their birthday is just before the cut-off date. When the school year starts in e.g. September, they still have three months before they turn 6. Pupils born in January on the other hand have to wait 9 months after their birthday before they can start school: they are the eldest in class.

The key advantage of using the national cut-off month for admission to primary school to construct RELAGE is that it provides us with an exogenous measure of a pupil's relative age inside his age cohort. Indeed, as explained above, it is the birth month of a pupil combined with the national cut-off month for starting school that determines its age position in class. We identify only two possible causes for endogeneity: birth month targeting by parents (in order to give their child a relative age advantage when starting school), or disease associated to birth months.

Bedard and Dhuey (2006) use the US Natality Files<sup>4</sup> and verify that parents do not target birth dates in a significant manner.

Mixed evidence exists in the psychological literature about seasonal variation in births of mentally ill children, the most important amount of research concerning autism and schizophrenia<sup>5</sup>. Such birth patterns would be related to different rates of occurrence of epidemics and differences in sunlight intensity and temperature between summer and winter (See e.g. Yeates-Frederikx et al. 2000). Leaving aside a discussion on the quality and abundance of evidence as to the existence of such birth patterns in this mainly qualitative branch of research, we do not think this represents a major issue for the identification of the effect of relative age. First of all, we use birth months relatively to legal cut-off dates as a variable, not season of birth. Moreover, we found no evidence as to any statistically significant association with cognitive skills of these mental problems. And, for many diseases, there is one specific month that appears to jump out as concentrating the highest frequency of patient birth dates. For unclear reasons, this is not the same month for all diseases and we found no evidence of systematic differences of IQ between birth months. Finally, we should keep in mind the low occurrence rates of autism and schizophrenia (around 0.1 % of adults are affected by autism<sup>6</sup>, 1% of adults by schizophrenia<sup>7</sup>).

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<sup>4</sup> These data contain the birth month and maternal socio-economic information for all US births.

<sup>5</sup> See e.g. Yeates-Frederikx et al.(2000), Landau et al.(1999) and Torrey et al.(1997) for reviews.

<sup>6</sup> Dougherty (2000)

<sup>7</sup> Gottesman (1991)

### **3. The data and choice of included countries**

#### **1) The data**

This paper uses cross-section data from the OECD survey conducted in 2003 as part of the Program for International Student Assessment (PISA). The PISA 2003 database contains comparable math, science and reading test scores of a sample of 15 years-old pupils coming from 40 countries. All pupil and school related variables are measured in the same way in all countries. Most importantly, the mathematics and reading test scores, which we use as our dependant variables, are also comparable across countries. Indeed, all pupils in the sample have been submitted to the same standard tests at age 15, taking into account possible school program and cultural differences. The scores are not comparable across disciplines. The sampling procedure ensures that schools and pupils inside them are randomly selected and that the sampled set of students is representative of the country's student population. Please refer to the PISA manual (2003) for more details as to how the data were constructed.

We have standardised test scores to mean 500 and standard deviation 100 by country in order to ease interpretation and comparisons of estimates.

#### **2) Choice of included countries**

In our paper, we only use 16 countries of the 40 countries in the dataset because we need the PISA sampling period to correspond to the legal cut-off date for entering primary school. In this section we explain how and why we proceeded to this restriction.

PISA is not sampled by grade. To the contrary, the sample of pupils in the dataset was created based on age. We have a representative sample of fifteen year olds, whichever grade they happen to attend at the time of the test. This means that we only observe all pupils that should be in the same grade if by chance the sampling was done from the legal cut-off date. Figure 1 illustrates why this is the case. If the cut-off date to enter primary school is for instance January and pupils were selected from January to January then we observe a complete grade. If on the other hand the cut-off date is September but sampling was done from January then we do not observe pupils born from September to December in grade x. These attend a grade lower for which we do not observe kids born from January to August. We can make similar reasoning for all cut-off birth month – sample birth month combinations that do not match.

In order to find out what the PISA sampling period was in each country it suffices to look at the birth month of the eldest pupils in the sample. If sampling was done from January to December then the oldest pupils in the country sample should be from January, the youngest from December.

We also need to verify what the legal cut-off date for starting primary school is in each country. We use the frequency patterns of pupils born in certain months per grade to check the legal cut-off date in each country. In the previous example, we for instance expect a frequency pattern of pupils per grade such that pupils born between September 1<sup>st</sup> and January 1<sup>st</sup> are mostly in grade 9, whereas those born before September should be mostly in grade 10. We thus compare the frequency pattern of pupils from each birth month per grade to infer the national cut-off rule for admission

to first grade. The cut-off dates deduced from the data are next verified to be correct using the information we found on the websites of the countries' Ministries of Education and on that of Eurydice<sup>8</sup>.

Finally, we compare the sampling birth month with the double-checked cut-off date. If they are the same we include the country in our dataset. This yields 16 useable countries. Most countries included have sampling cut-off dates in January. Exceptions are Korea and New Zealand where these are March and May respectively.

### **3) Spreading across grades**

As mentioned before, the PISA sample contains only 15 year-olds. But pupils may start the first grade of primary school at a higher(lower) age than the national compulsory school starting age. Some of these are put in the grade they should have started in at the legal age but some are placed a grade lower(higher). Therefore, when they reach the age of 15 they are one grade 'behind' ('in advance') with respect to the other 15 year olds. These pupils may attend other grades than the one they should attend according to the admission rule to primary school and their birth date. Moreover, pupils may repeat or skip a grade. They then end up in a lower (higher) grade than other pupils that have the same age.

The considered countries present different degrees of spread in the distribution of pupils across grades. Figure 2 presents the proportions of pupils in each grade, by country. In some countries (Iceland, Japan, Norway, Korea, Poland, Sweden, and former Yugoslavia<sup>9</sup>) nearly all 15 year olds are in the same grade. There are less than 3 percent of the pupils that repeated a grade and most pupils respect the entry rule. In other countries (such as Belgium, France, Portugal and Spain), 20 up to 30% of 15 year olds are in a lower grade than the one they should be in according to their birth-date. In the remaining countries (Canada, Denmark, Italy, Latvia, and New Zealand) between 5 and 15 % of pupils are in a too low grade according the admission to primary school rule.

### **4) Control variables**

We estimate a classical education production function where test scores are a function of a selection of explanatory variables, one of which is relative age. Because it is not possible to control for all possible country differences, we estimate this education production function for all included countries separately. We include a large set of pupil and school control variables taken from the three most important set of factors that are commonly thought to influence test scores: individual, family and school characteristics.

The only individual variable we include is the pupil's gender (dummy BOY). From the set of family variables, we firstly chose to use the socio-economic status of the family as measured by HISEI (an index based on the professional status of the

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<sup>8</sup> The information network on Education in Europe, European Commission (2006) . We also check our dates are the same as those used by Bedard and Dhuey (2005) for countries present in both our studies.

<sup>9</sup> Actual Serbia and Montenegro. We refer to the country by former Yugoslavia throughout the paper, as this was its name at the time of the survey.

parents<sup>10</sup>), and HOMEPOS (a PISA index based on ownership of several domestic equipments, cars , televisions etc.), an indicator of material wealth. The level of education and attitude to learning of the parents are approximated by HISCED (highest level of education reached by the parents), NBBOOKS, the number of books in the home, and HEDRES, the amount of educational resources in the home (based on the availability of a dictionary, calculator, quiet place to study etc.). TIMEMATERNELLE is a discrete variable indicating the time spent in kindergarten by the pupil. Possible values are 0 (did not attend kindergarten), 1 (one year of kindergarten) and 2 (more than one year of kindergarten). This variable is relevant as in some countries children may already acquire certain skills in kindergarten, before they start school.

At the level of the school, we include the degree of school autonomy of decision making (SCHAUTON), school size (SCHLSIZE), the amount of educational material (SCMATEDU, based on the availability of a library, computers etc.) and the degree of private financing of the school (PRIVATESCHOOL). Note that for France, all school variables are missing because the French schools did not answer the PISA questionnaires.

We do not include national institutional characteristics as we estimate countries separately. Note that our results should not be sensitive to the inclusion of these control variables as RELAGE is exogenous.

#### **4. Estimation strategy**

The aim of our estimation will be to assess first of all whether relative age in the first years of primary school has a significant effect or not on test scores at 15 years old. To reach this first objective we start by estimating what we call the aggregate effect of relative age. Its coefficient will capture variation in test scores that may be due to intra-grade relative age effects but also to misplacement<sup>11</sup>, grade retention and selection into the vocational track or other forms of ability grouping.

We take into account the hierarchical structure of educational data: pupils in a same class are thought to have similar unobserved characteristics, as do pupils from a same school or country. As a result, we know that pupil's test scores may be correlated if they are in the same school. We therefore estimate our equation as a mixed model that allows correlation of test scores inside the same school. This appears as the inclusion of a school 'random' effect in the estimated equation.

This yields specification (1) :

$$TestScore_{is} = C + \alpha.RELAGE_i + \Sigma\beta X_{is} + \gamma_s + \varepsilon_{is}$$

Where  $X_{is}$  stands for the mentioned control variables and  $\gamma_s$  for the school 'random' effect.

We expect the effect of relative age within the first grades of primary school on test scores at age 15 to work through several channels. As mentioned in the introduction, the theoretical literature points to ability grouping, grade retention and early selection

<sup>10</sup> PISA International Socio-Economic Index of Occupational Status

<sup>11</sup> By misplacement we refer to children starting school too early or too late, that therefore attend a different grade than the one predicted by their birth month.

into vocational tracks. We do not have information as to the degree of ability grouping that goes on at the start of primary school. However, we do know whether the school groups pupils by ability at age 15. We also know whether a pupil repeated a grade, attends a vocational school and we can deduce from the cut-off date, birth month and grade currently attended whether the pupil started on time. All these variables relate to spreading pupils across grades in a way that could be correlated with relative age.

In order to assess whether these channels might play a role and to what extent, we firstly estimate the effect of relative age on the probability of grade retention (dummy RETENTION equal to one if the pupil was retained one or more times) and of ending up in a vocational track (dummy VOCED) for each country. Relative age is exogenous with respect to these dependant variables. The included control variables are the previously presented individual and family characteristics. School characteristics at the time of the test cannot be deemed determinants for the probability of retention and tracking choices that took place earlier in time, and are therefore not included.

We would also like to know whether the probability of starting primary school too early or too late is related to RELAGE. Because very few pupils are concerned in some countries, we merely compute descriptive statistics of the relation between relative age and late or early start (dummies LATESTART and EARLYSTART equal to one if the pupil started one year too late or too early) of primary school. Secondly, we would like to know whether relative age has an effect on long term test scores only through these channels or whether there is a 'residual' effect that exists independently of 'spreading' the pupils across grades.

We thus estimate a second specification of the test scores production function (1) where we include RELAGE as well as potential 'channel' variables RETENTION, VOCED, LATESTART, and EARLYSTART as well as dummy variables for each grade (except the one most pupils attend). Pupils that attend different grades are indeed expected to obtain different test scores at the same age of 15 as they have not received the same amount of teaching<sup>12</sup>.

We are interested in the significance of the effect of the 'channel' variables but foremost in the residual effect of RELAGE. Indeed, the 'channel' variables are most likely endogenous. The choice of entry into primary school as well as grade retention can be expected to depend on unobserved pupil and parent characteristics such as ability or parent motivations. However, RELAGE remains exogenous. Correlations between RELAGE and the channel variables can therefore be interpreted as an effect of RELAGE on them, not the reverse. As a result, we can interpret changes in the coefficient of RELAGE on test scores as meaning that the channel variables are the channels through which RELAGE affects test scores.

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<sup>12</sup> It is possible to include both grade dummies and RETENTION because RETENTION is a dummy variable equal to one if the pupil repeated one or more grades, it does not specify the number of grade repetitions.

This yields specification (2):

$$TestScore_{is} = C + \alpha.RELAGE_i + \beta.X_{is} + \beta_r.RETENTION_i + \beta_v.VOCED_i + \beta_E.EARLYSTART_i + \beta_L.LATESTART_i + Dummies_{GRADE} + \gamma_s + \varepsilon_{is}$$

For each country, we adjust this specification according to the existence of grade retention, a vocational track, and the number of grades pupils are spread over.

## 5. Results

### 1) The baseline model

Table 1 shows the estimates of the aggregate effect of relative age on test scores for each country (specification (1)). This effect is significantly positive in 10(9)<sup>13</sup> out of 16 countries for Reading and Maths test scores respectively. Being one month older in your age-cohort in the first years of schooling increases reading (maths) test scores by between 0.80 (0.40) and 2,21(2.68) points (one point is one percent of a standard deviation). Though highly statistically significant, this effect may seem negligible in size. We should not forget however that pupils differ in age up to 11 months in the same grade. The obtained coefficients imply that the oldest pupil in a cohort (RELAGE = 12) on average obtains between 9(4.4) and 24(30) points higher test scores than the youngest cohort member (RELAGE = 1). Relative age hence has an important and significant positive effect on long-term test scores in most considered countries.

The magnitude of the coefficient is in line with previous results in the literature. Indeed, the effect of being the oldest pupil in a cohort is generally estimated to yield around 0.2 (between 10 and 27% ) standard deviations higher test scores than the youngest pupils in the cohort (Strom, 2004, Frederiksson and Ökert, 2005, Bedart and Dhuey, 2005). Our standard deviation is 100 test score points therefore these effects correspond to about 20 points. We do not interpret our results as being related to maturity differences at age 15. First of all, such differences seem unlikely to exist at age 15. Moreover, if there were significant maturity differences between children at age 15 we would expect the effect of RELAGE to be significant in all countries, which is not the case.

The control variables yield the expected sign and significance. Boys have significantly higher test scores in maths and significantly lower test scores in reading. All the family characteristics except parents' education have a very significant positive impact on test scores. To the contrary, school variables such as decision making autonomy and availability of educational materials are barely significant. Only school size seems to have a small positive effect on test scores. Our results are not significantly different when we exclude the control variables. The average difference between the coefficient of relative age in specification 1 and the specification with no control variables is of 0.01<sup>14</sup>. This indicates that RELAGE is indeed exogenous with respect to the other explanatory variables.

<sup>13</sup> 10(9) stands for coefficient in Portuguese (coefficient in Maths )

<sup>14</sup> Detailed results of the specification with no controls are available upon request.

## 2) What channels could relative age work through?

The results of the estimations of the impact of relative age on the probability of repeating a grade and of ending up in a vocational track are reported in Tables 2 and 3, for the countries where grade retention and a vocational track exist. The relative age of a pupil when he enters primary school has a significant effect on the probability of grade retention in six out of eleven countries. In these countries, being the eldest rather than the youngest in your age cohort when starting school reduces the probability to repeat a grade by between 2 and 15 percent.

Other factors that seem to play an important role<sup>15</sup> in determining grade retention are gender, parents' material wealth (HOMEPOS) and the number of books at home (NBBOOKS). Boys are more likely to repeat a grade than girls. Noticeably, not living with both parents has a very strong effect on the probability of repeating a grade. We could interpret this as being (indirectly) related to the temporary problems at home related to e.g. parents divorce.

The probability of reaching a vocational track does not seem related to relative age at school entry in most countries. The coefficient is significant in only two out of seven countries: Italy and former Yugoslavia. Strong determinants of the probability of being tracked into vocational education are parent education (HISCED), material wealth and the number of books at home. Not living with both parents has no significant effect on the probability of following the vocational track. Entering a vocational track seems to be much more related to the 'structural' family background than to more punctual events such as parents divorce or relative age at school entry.

In Table 4, we observe the percentage of pupils that attend a too high grade that are relatively eldest of their cohort according to their birth date. In all countries where there are 15-year-old pupils in too high grades, these are mostly (more than 50%) the eldest pupils of their cohort (relative age between 9 and 12). This points to a relation between skipping a grade and relative age. Relatively older pupils seem to be more likely to skip a grade.

Table 5 presents the correlation between relative age and early or late start of primary school in each country. In 9 out of 15 countries more than 2% of pupils start school a year or more too early. In 11 out of 15 more than 2% start school a year or more too late. In most concerned countries, starting too early is significantly positively correlated with relative age. In these countries, it is more likely to start school too early when you are the eldest of your age cohort. This is in line with the intuition that school entry decisions are probably based on observed pupil maturity. Starting school too late is significantly correlated with relative age in 11 out of 15 countries, but the sign of the correlation is not clear.

We conclude from this section that there is a strong relation between relative age and the variables we expect to be channels through which age affects long-term test scores. An exception is selection into the vocational track which does not seem related to relative age but rather to more 'structural' family characteristics (based on the restricted set of control variables included). We shall now look at potential residual

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<sup>15</sup> Presented results are conform to specification 3. Complete results are available from author upon request.

effects of relative age (effects that would not go through these channels) and at the significance of the different channels in each country.

### 3) What channels does relative age work through?

Table 1 presents the estimates of specification 2 in the right column (residual effect). It appears that most of the effect of relative age goes through the included channels as in eight out of ten (nine) countries where RELAGE has an effect the coefficient becomes non significant or is importantly reduced when including the ‘channel’ control variables. Exceptions to this conclusion are Japan and Korea (only for reading test scores) where the effect of relative age seems unrelated to the mentioned channels, and remains significantly positive in both specifications. In Denmark and in Sweden, a significant residual effect remains though the coefficients are significantly reduced.

The coefficients of the additional control variables of specification (2) are reported in Tables 6 and 7 for reading and maths test scores respectively. Recall that the coefficients of these channel variables are not to be interpreted as causal effects because of an endogeneity problem. We present them merely to give an idea of their overall significance by country. Note that in some countries one or more of the additional variables may be missing. This is because these variables do not exist for the concerned country. For instance there is no vocational track in many of the considered countries. Overall, many of the additional variables have a large significant effect on test scores. Attending a vocational rather than a general secondary track, repeating a grade, as well as starting primary school either too late or too early all have a major negative impact on test scores in most countries considered. Attending lower or higher grades than the majority of the age cohort also very significantly affects test scores as expected.

Combining the significance of these coefficients in each country with the information about the relation between relative age and the additional variables (presented in Tables 2 to 5) we come to the following conclusions.

In five countries, relative age has no effect on test scores in either discipline or specification. These are France, Iceland, Poland, Portugal and former Yugoslavia. In Iceland all pupils are in the same grade, there is no tracking into vocational education or grade retention, and pupils even all start at the right age. It is therefore not surprising that relative age in the first years of primary school has no effect on test scores in Iceland, as none of the proposed channels through which it may work exist. In Poland 96 percent of 15 year olds attend the same grade and LATESTART and grade retention are not significantly correlated with relative age. In former Yugoslavia, the vast majority of pupils (97%) also attend the same grade, start on time and grade retention is exceptional. 77% of pupils attend vocational education but the probability of following that track is only weakly correlated with relative age.

France and Portugal do not fit into this frame. They present a lot of grade retention, late and early start that is related to RELAGE, but the aggregate effect of RELAGE on test scores is not significantly different from zero. We note that in both countries, the probability of repeating a grade strongly decreases with RELAGE. This means a lot of relatively younger pupils are repeating grades (Table 3). However, the younger

children tend to perform better on average than the older (the residual coefficient of RELAGE is negative). A possible explanation of this result could be the expectations held by teachers as to the performance of relatively younger pupils. If teachers do not have lower expectations from younger pupils, then the effect of relative age would disappear. However, we do not have evidence on the type of expectations held by teachers in France and Portugal compared to other countries and thus we cannot test the validity of this explanation here.

To the opposite, in Denmark, Sweden, Japan, and Korea, relative age has a significant positive effect on test scores in both specifications and disciplines. This implies there is an effect of relative age at the start of primary schools on test scores in these countries, but it does not work through the proposed channels. In effect, in Japan, all 15 year olds are in the same grade and there is no grade retention. There is a vocational track in Japan but the probability of following it is not related to relative age. In Sweden, the only channel that is at work is older pupils skipping a grade. The coefficient of RELAGE is slightly reduced by including GRADE11 but remains significantly positive and of the same magnitude. In Korea, relative age in the first years of primary school has a significant positive effect on Reading test scores, none on Maths test scores. None of the channel variables have a significant effect on test scores in Korea and 99 percent of 15 year olds attend the same grade. The residual effect of relative age in Japan, Korea (Reading) and Sweden, as well as that found in the countries where including the channel variables only reduces the coefficient of relative age, point to the existence of other explanations for an effect of relative age.

As mentioned before, we only have information on ability grouping at age 15 therefore we cannot test the relevance ability grouping as a channel for RELAGE. Nevertheless, it is interesting to see that in these countries there does seem to occur some ability grouping though no clear streaming of pupils exists. Figure 3 shows that there is a considerable amount of intra-class ability grouping and that many pupils attend different levels of maths classes in countries with a comprehensive schooling system such as Iceland, Norway and Sweden. Moreover, there exists a positive correlation between the estimated residual effect of RELAGE and the degree of ability grouping at age 15 as can be seen in Table 8.

This does not hold for Japan and for Korea (Reading test scores). However, these two countries have another particularity in that the schooling system is extremely competitive from the very start. Pupils have to go through admittance examinations from an early age. This may imply that the eldest pupils, being more mature at time of first entry exam) are selected into the better schools. Table 9 shows the percentage of pupils that attend schools where academic record is a prerequisite for admittance in each country. Noticeably, Korea, Japan and Yugoslavia are very selective schooling systems as compared to the rest of the countries at age 15. Moreover, there is a positive correlation between attending a school where academic record is a prerequisite and RELAGE in Japan. Further research using primary school data on these countries is required to check these hypotheses.

In the majority of countries (Belgium, Canada, Denmark, Italy, Latvia, New Zealand, Norway, Spain) the positive effect of relative age on test scores seems to happen mainly through the proposed channels. When we include the channel variables, the coefficient of relative age either becomes non significant or is strongly reduced in

size. Which channels are active in each country is reported in Table 10. Overall, the most important channels for RELAGE appear to be LATESTART, RETENTION and GRADE11. Starting school too late, repeating a grade or skipping a grade are strongly correlated with RELAGE, and all have a significant effect on test scores. This reinforces our conclusion that in those countries where RELAGE has a lasting effect on student performance, this mainly occurs through the pupils being spread across grades according to their relative age. Selection into the vocational track on the other hand is not related with RELAGE, except in Italy.

## ***Conclusion***

Relative age when entering primary school has a positive effect on long-term test scores in 10 out of 16 countries. The size of the effect is around 0.2 standard deviations in test scores, which is quite large in the estimation of education production functions. According to our results, in most countries the effect of relative age on long term test scores works through the spreading of pupils across grades and other types of ability grouping. In particular, the effect seems to occur through increased grade retention and/or late start among the youngest pupils in an age cohort. An exception to this rule are Japan, Korea and Sweden, where other types of ability grouping and very early pupil competition could play a role. Moreover, there are also some countries where relative age has no effect on test scores in spite of the existence of ability grouping. We think these differences may be due to the role played by teacher expectations towards the performance of younger children. In the psychological literature there is qualitative evidence that teacher expectations and objectivity are important determinants of test scores<sup>16</sup>. Our results are compatible with the existence of large scale effects of teacher expectations on pupil performance. Further research could justify policy intervention to improve teacher awareness and training as to the impact of their expectations.

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<sup>16</sup> See e.g. Bulterman-Bos et al. (2002) for a recent paper

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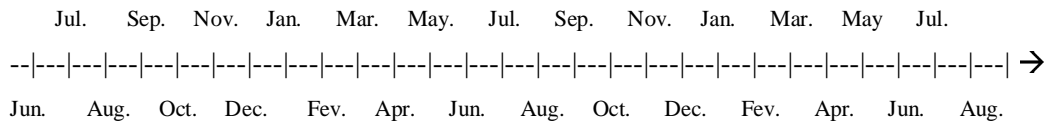
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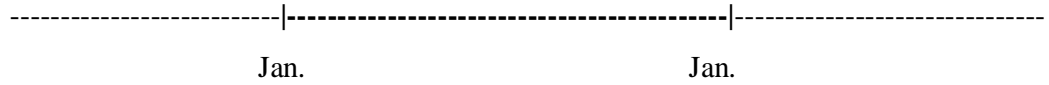
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**Figure 1: Matching PISA sampling dates and national cut-off dates.**

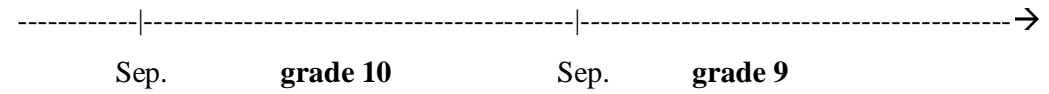
Birth months



PISA sampling rule (example)



Cut-off date for admission into primary school (example)



Missing observations if PISA sampling rule is different from admission rule.

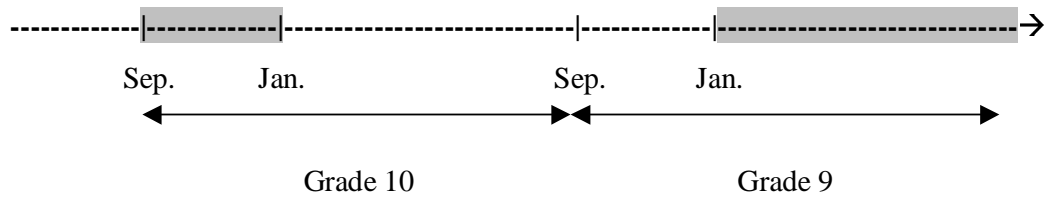
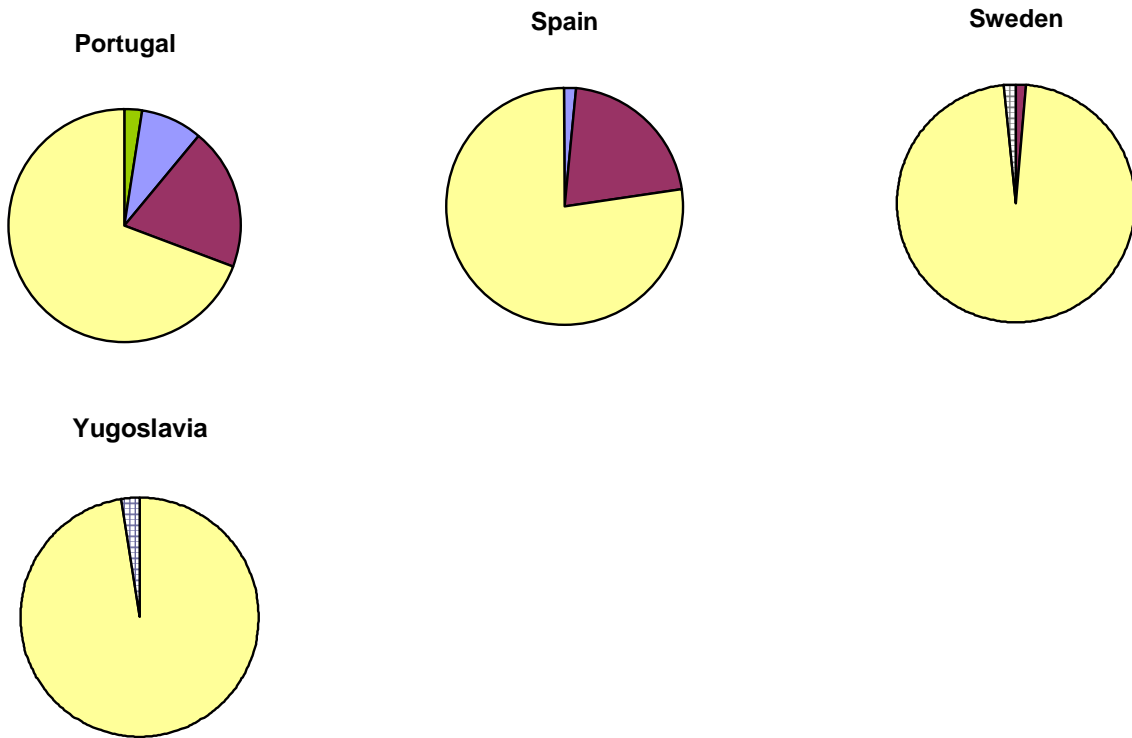


Figure 2: How are 15 year olds spread over grades in each country ?





Percentage of pupils in each grade, by country.

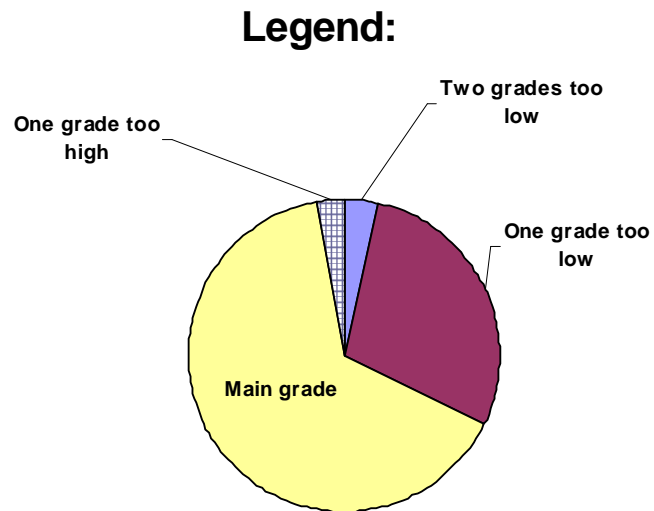
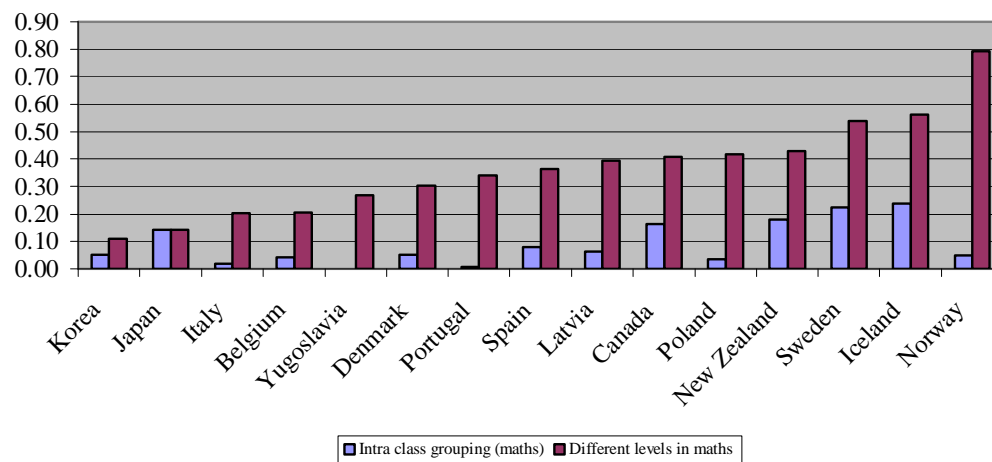


Figure 3:

### Percentage of schools that practice ability grouping



**Table 1: Aggregate and residual effect of being one month older at the start of primary school  
Dependant variables: reading and maths test scores at age 15**

	READ				MATHS			
	Aggregate effect		Residual effect		Aggregate effect		Residual effect	
	Coeff.	Std Dev.	Coeff.	Std Dev.	Coeff.	Std Dev.	Coeff.	Std Dev.
Belgium	1.29 ***	0.32	0.39	0.31	1.00 ***	0.31	-0.08	0.30
Canada	0.80 ***	0.26	0.28	0.26	1.30 ***	0.27	0.54 **	0.27
Denmark	2.21 ***	0.53	1.01 *	0.57	2.68 ***	0.54	1.54 ***	0.57
France	-0.19	0.46	-0.73	0.45	-0.49	0.46	-1.11 **	0.44
Iceland	0.00	0.54	0.01	0.54	0.78	0.56	0.79	0.56
Italy	1.05 ***	0.24	0.82 ***	0.24	0.40 *	0.24	0.12	0.24
Japan	1.06 ***	0.37	1.06 ***	0.37	0.95 ***	0.34	0.96 ***	0.34
Korea	1.06 ***	0.35	1.04 ***	0.35	0.29	0.33	0.32	0.33
Latvia	0.89	0.57	-0.74	0.61	1.45 **	0.57	-0.60	0.61
New Zealand	1.83 ***	0.52	0.23	0.57	1.75 ***	0.53	0.01	0.57
Norway	1.09 **	0.53	0.82	0.54	0.53	0.53	0.326	0.542
Poland	0.48	0.40	0.30	0.39	0.45	0.41	0.24	0.40
Portugal	-0.21	0.44	-0.97 ***	0.43	0.58	0.44	0.01	0.43
Spain	0.80 ***	0.30	0.16	0.28	1.19 ***	0.30	0.46 *	0.28
Sweden	1.21 **	0.47	0.87 *	0.47	1.70 ***	0.47	1.35 ***	0.47
Yugoslavia	-0.13	0.47	-0.34	0.47	0.72	0.46	0.36	0.47

\*, \*\* and \*\*\* respectively stand for significance at a 10, 5 and 1% level of confidence

Aggregate effect : Coefficient of RELAGE : specification 1.

Residual effect: Coefficient of RELAGE when we include grade retention, VOCED, late and early entry and grade attended by pupil: specification 2.

**Table 2: Probit estimation, coefficients of relative age (effect of being one year older). Dependant variable: Being in a vocational track. Specification (2)**

Country	Coeff.
Belgium	-0.03
France	0.01
Italy	-0.11 ***
Japan	0.01
Korea	0.01
Portugal	0.02
Yugoslavia	-0.04 *

<sup>1</sup> \*, \*\* and \*\*\* respectively stand for significance at a 10, 5 and 1% level of confidence

Only countries where there exists a vocational track are included.

**Table 3: Probit estimation, coefficients of relative age (effect of being one year older). Dependant variable: Pupil has repeated a grade at least once. Specification (3)**

Country	Coeff.
Belgium	-0.09 ***
Canada	-0.03 ***
Denmark	0.00
France	-0.12 ***
Italy	-0.02 ***
Latvia	-0.01
New Zealand	0.00
Poland	-0.01
Portugal	-0.15 ***
Spain	-0.07 ***
Sweden	0.00

<sup>1</sup> \*, \*\* and \*\*\* respectively stand for significance at a 10, 5 and 1% level of confidence

Only countries where there exists significant grade retention are included.

**Table 4: Percentage of pupils attending a grade higher than their age cohort, that are the eldest of their age cohort (relative age between 9 and 12).**

	Percentage	Nb. of pupils
Belgium	77%	51
Canada	70%	178
Denmark	90%	56
France	64%	65
Iceland	NA <sup>17</sup>	0
Italy	70%	196
Japan	NA	0
Korea	100%	4
Latvia	66%	156
Norway	NA	0
New Zealand	93%	81
Poland	58%	19
Portugal	80%	21
Spain	NA	0
Sweden	63%	54
Yugoslavia	90%	67

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<sup>17</sup> Not applicable

**Table 5: Correlation coefficient between RELAGE and early and late start of primary school.**

Country	Corr. Coeff. Early start		Corr. Coeff. Late start		Percentage Early start	Percentage Late start	Nb. of Obs
Belgium	0.02	*	-0.03	**	2.0	10.6	5093
Canada	0.04	***	0.09	***	0.3	24.7	10465
Denmark	0.02		-0.16	***	4.5	7.0	2599
France	0.04	*	0.06	***	6.6	10.0	2380
Iceland	No obs.		0.00		0.0	1.7	2473
Italy	0.12	***	0.05	***	1.9	3.3	8717
Japan	No obs.		No Obs.		No obs.	No obs.	3709
Korea	-0.11	***	-0.01		1.2	7.0	4352
Latvia	0.19	***	-0.26	***	7.2	10.6	2160
New Zealand	0.03		-0.18	***	4.4	5.8	2429
Norway	-0.14	***	-0.06	***	5.2	0.4	2572
Poland	0.04	***	-0.01		0.6	5.5	3998
Portugal	0.04	**	0.04	**	0.8	11.4	2997
Spain	0.19	***	0.02	**	9.5	2.5	7449
Sweden	-0.02		0.035	**	3.6	0.7	3244
Yugoslavia	0.17	***	0		2.4	1.5	2577

\*, \*\* and \*\*\* respectively stand for significance at a 10, 5 and 1% level of confidence.

Early start is a dummy equal to one if the pupil started school one or more years too soon compared to the compulsory school starting age.

Late start is a dummy equal to one if the pupil started school one or more years too late compared to the compulsory school starting age.

**Table 8: Correlation coefficient of the estimated effect of RELAGE (in maths) and the percentage of schools that group pupils by ability in each country.**

	Aggregate effect	Residual effect
Intra-class ability grouping at age 15	0.35	0.43
Different level courses in maths	0.09	0.08

**Table 9 : Percentage of pupils that attend a school where Academic record is a prerequisite for admittance, by country.**

	Academic record prerequisite
Average other countries	0.02
Belgium	0.15
Korea	0.41
Japan	0.67
Yugoslavia	0.78

**Table 10: Summary of effect of RELAGE and channels through which it occurs**

Country	Total effect of RELAGE	Channels through which it occurs					Remarks
		Voced	Earlystart	Latestart	Retention	Grade too high	
Belgium	Positive	x	x	x	x		
Canada	Positive				x	x	
Denmark	Positive			x		x	
France	None	x				x	No school info
Iceland	None						
Italy	Positive	x		x	x	x	
Japan	Positive						
Korea	Positive						Only for reading
Latvia	Positive	x				x	
New Zealand	Positive			x		x	
Norway	Positive	x					Only for reading
Poland	None	x					
Portugal	None			x	x	x	
Spain	Positive			x	x		
Sweden	Positive			x		x	
Yugoslavia	None					x	

READ	VOCED		RETENTION		EARLYSTART		LATESTART		GRADE -3		GRADE -2		GRADE -1		GRADE +1	
	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.
Belgium	-59.80***	3.39	-14.71**	7.00	-11.44	8.00	-8.38**	3.63	-218.13***	46.13	-107.90***	12.16	-37.46***	7.00	32.99***	11.29
Canada			-62.97***	4.82	8.36	19.82	0.90	2.15			-73.43***	21.22	-28.02***	5.00	17.64*	10.50
Denmark			-37.90***	12.14	0.37	10.83	-7.26	8.90					-30.06***	10.40	54.66***	16.06
France	-64.25***	7.69	-7.05	17.67	-17.16***	6.54	5.63	5.14			-143.17***	20.02	-90.27***	18.15	49.55***	10.07
Iceland							-20.76	14.52								
Italy	-49.38***	5.08	-21.90**	8.95	-16.08	11.56	-8.42*	4.83	-234.43***	85.33	-112.18***	22.12	-20.86**	8.70	30.21***	10.72
Japan	-28.08**	11.79														
Korea	-80.76***	6.33	-53.62**	26.37	-11.24	11.61	-14.06**	5.76					-15.36	16.67	81.11**	40.61
Latvia			-44.31***	11.05	28.44*	15.79	-2.19	9.23			-62.58***	23.25	-28.89***	9.73	3.16	16.00
New Zealand			-50.58***	9.77	-11.49	8.53	-25.95***	9.55			-82.61	85.40	-31.19***	11.05	50.28***	9.94
Norway					-18.65**	8.11	16.52	44.18					-63.96	42.04		
Poland			-123.49***	13.46	-15.45	26.03	-26.67***	6.03					3.25	14.39	51.64	32.89
Portugal	-24.23***	5.03	-85.80***	4.38	-20.69	20.70	-7.32	5.00	-99.54***	9.62	-49.14***	5.82			75.33***	21.24
Spain			-19.37*	11.42	2.96	3.33	-12.45**	6.14			-100.65***	13.46	-60.24***	11.42		
Sweden	-11.87	27.72	3.47	13.20	-5.41	10.33	-53.11***	18.90					-103.32***	18.39	41.65**	17.81
Yugoslavia	-68.79***	8.53	-49.24**	24.51			-12.38	13.22							19.81*	10.43

<sup>1</sup> \*, \*\* and \*\*\* respectively stand for significance at a 10, 5 and 1% level of confidence

**Table 6: Coefficients of control variables through which relative age may affect test scores.  
Dependant variable: Reading test scores.  
Mixed model with specification (2).**

MATH	VOCED		RETENTION		EARLYSTART		LATESTART		GRADE -3		GRADE -2		GRADE -1		GRADE +1	
	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.	Coeff.	Std Err.
Belgium	-71.24 ***	3.27	-26.66 ***	6.61	-17.25 **	7.55	-10.11 ***	3.43	-189.93 ***	44.08	-111.43 ***	11.52	-32.18 ***	6.61	55.72 ***	10.65
Canada			-66.53 ***	4.86	29.11	19.98	3.71 *	2.16			-84.34 ***	21.40	-39.06 ***	5.04	39.17 ***	10.58
Denmark			-58.89 ***	12.14	-7.42	10.84	-17.82 *	8.91					-24.21 **	10.37	47.59 ***	15.80
France	-63.49 ***	7.04	-13.00	17.22	-19.33 ***	6.38	-0.70	4.93			-152.38 ***	19.58	-94.34 ***	17.56	63.49 ***	9.84
Iceland							-14.04	14.86								
Italy	-44.49 ***	5.31	-18.95 **	8.66	-5.84	11.20	-23.59 ***	4.68	-307.54 ***	84.42	-123.69 ***	22.56	-34.61 ***	8.43	38.21 ***	10.39
Japan	-39.31 ***	13.37														
Korea	-82.58 ***	6.70	-4.25	24.85	-1.99	10.91	-18.83 ***	5.48					-13.70	17.21	45.80	38.08
Latvia			-61.95 ***	10.91	8.92	15.58	-6.82	9.11			-101.34 ***	22.95	-31.29 ***	9.61	30.39 *	15.80
New Zealand					-15.96 *	8.19	2.87	44.67					-32.56	42.51		
Norway			-53.77 ***	9.89	-11.38	8.63	-34.92 ***	9.67			-42.90	86.40	-32.72 ***	11.18	46.84 ***	10.06
Poland			-119.01 ***	13.81	-47.00 *	26.65	-34.00 ***	6.18					6.74	14.77	113.39 ***	32.45
Portugal	-35.30 ***	4.89	-97.13 ***	4.15	-24.03	20.32	-5.23	4.89	-98.84 ***	9.40	-42.93 ***	5.70			101.06 ***	21.36
Spain			-19.42 *	11.22	4.71	3.27	-18.21 ***	6.04			-125.40 ***	13.23	-69.44 ***	11.23		
Sweden	1.49	27.95	-1.02	13.40	-2.48	10.37	-31.06	18.93					-81.05 ***	18.50		
Yugoslavia	-85.68 ***	9.03	-29.83	24.32			-34.75 ***	13.12							40.30 ***	10.35

<sup>1</sup> \*, \*\* and \*\*\* respectively stand for significance at a 10, 5 and 1% level of confidence

**Table 7: Coefficients of control variables through which relative age may affect test scores.  
Dependant variable: Maths test scores.  
Mixed model with specification (2).**