

1 **Supplementary Information (SI)**

2 For:

3 **Gender differences in individual variation in academic grades fail to fit expected**
4 **patterns for STEM**

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Supplementary Methods

Literature search

We collected data in two rounds. First, in March 2015, we collated the list of 309 studies and 502 effect sizes used in Voyer's systematic meta-analysis on gender differences in scholastic achievement¹. We obtained copies of all but 14 studies from databases and libraries, and selected those with variance data for males and females (see below for eligibility criteria). Second, in June 2015, we searched for studies published since August 2011 (the time when Voyer's original search ended¹) using three online literature databases. The exact search strings used were:

ERIC:

"school grade*" OR "school achievement*" OR "school mark*" OR "grade point average*"

Last 5 years, Peer reviewed only

SCOPUS:

TITLE-ABS-KEY ("school grade*" OR "school achievement*" OR "school mark*" OR "grade point average*") AND (LIMIT-TO(PUBYEAR, 2015) OR LIMIT-TO(PUBYEAR, 2014) OR LIMIT-TO(PUBYEAR, 2013) OR LIMIT-TO(PUBYEAR, 2012) OR LIMIT-TO(PUBYEAR, 2011))

ISI Web of Science:

TOPIC: (("school grade*" OR "school achievement*" OR "school mark*" OR "grade point average*")) Timespan: 2011-2015. Search language=Auto

Supplementary Fig. 3 shows the number of studies obtained by our search. Two investigators independently screened the titles and abstracts of these studies using *abstrackr* software², based on the detailed exclusion/inclusion criteria shown in Supplementary Fig. 3. Disagreements (17.7% of abstracts) were discussed and resolved. We then briefly screened the included studies for available data (presence of mean scores for males and females), before undergoing full-text screening and data extraction (described in the main text).

52

53 *Mean-variance relationship*

54 We observed strong relationships between mean grades and their variance estimates,
55 for both genders, in the raw data (Supplementary Figs. 1 – 2), justifying the use of
56 *lnCVR* as an effect size for statistically comparing grade variability between the sexes.
57 However, when we restricted our plots to the data that could be considered most
58 comparable due to using similar grading scales (0 – 5, or 0 – 100, North American
59 schools), the relationship between means and SD disappeared (Supplementary Fig. 5).

60

61 *Multiple imputations*

62 We estimated missing values of students' mean age (37% missing) and study year
63 (50% missing) using multiple imputation methods, implemented with the *mice* R
64 package³. We used the following variables as predictors of missing values:
65 publication year, school type, subject, minimum and maximum age of students in the
66 cohort, and sex ratio of the sample cohort. We performed 100 imputation rounds ($m =$
67 100) for each data subset (full dataset, school subset, university subset, and data
68 stacked by gender that was used for *lnCV* analyses). We ran meta-regression models
69 that included study year, or mean student age, using the imputed datasets, and then
70 pooled the results to average the estimates and compute the variances.

71

72 *Data heterogeneity*

73 We estimated total heterogeneity in the collected effect sizes as I^2 total values. I^2
74 statistics represent the proportion of the total variation that is due to the between-
75 study variation (i.e. the true heterogeneity) rather than due to sampling errors
76 associated with each effect size. Total heterogeneity values in the whole dataset and

both school and university data subsets were very high ($I^2_{\text{Total}} > 98\%$; Supplementary Table 19), warranting looking for predictor variables (moderators) to explain variation among effect sizes.

University data subset and full data set

Meta-analytical and meta-regression models reported for the school data subset in the main text were also run on the university subset and the full data set (school and university data pooled). We found no effect of study year (Supplementary Tables 5-6). The gender difference in mean grades decreased significantly as students grew older, but this difference was only significant in the full data set (Supplementary Table 6). The gender difference in grade variability did not change with student age (Supplementary Table 6), but total variability decreased with age among both girls and boys at university, and only among boys in the full data set (Supplementary Table 7). When study subject was considered, the gender gap in mean grades was observed in the university subset and the full data set for every subject except “Other/NR” (Supplementary Table 8). Lower female variability was significant in every subject for the full data set, but only significant in STEM and Global subjects in the university subset (Supplementary Table 9). Additionally, data from the university was predominantly for “Global” grades; there were only 13 data points for science, 10 for maths and 4 for language. These low sample sizes resulted in broad confidence intervals for moderator analyses on the effect of subject (e.g., Supplementary Fig. 10E), and sensitivity to outliers. In particular, one influential study for language grade variability in the university subset strongly influenced the difference in variability between STEM and non-STEM subjects, causing there to be no significant differences in variability in non-STEM (Supplementary Table 9).

Robustness of results

1. We considered an alternative way to classify school/study subjects, following the approach in¹, with six subject categories: global, science, math, language, social sciences and other. We re-ran appropriate meta-regression models using this moderator. The results are qualitatively similar to these obtained with our original subject classification (STEM, non-STEM, Global, Other) and are presented in Supplementary Fig. 10 and Supplementary Tables 16 – 19.
2. We tested the robustness of our meta-analytical and meta-regression results using the *robumeta* R package⁴ for robust variance estimation (RVE). The results of these analyses are very similar to those obtained using the *robust* function in the *metafor* R package, and they are available in this online repository: <https://osf.io/ejqm4/>.
3. We constructed multivariate (full) meta-regression models to simultaneously test the effects of three moderators: study year, mean student age, and subject type (classified as in Voyer & Voyer 2014). The outcomes from these models (Supplementary Tables 21 – 23) match the patterns observed in the univariate models (no effect of study year or age, higher female mean grades, and lower female grade variability, especially for Global and Language grades in the school subset).
4. We also used Bayesian meta-regression models, with all three moderators (study year, mean student age, and subject type) included. These models were run using *MCMCglmm* package⁵ with default (non-informative) prior, except for university subset where a inverse-Gamma prior was used for the variance component for comparison ID, and parameter expanded prior for the variance component for cohort ID. For each model we ran 3 chains, with 110,000 iterations, 10,000 burn-in periods, and thinning by every 1000. The results obtained were qualitatively similar to the

results from the maximum-likelihood (*metafor*) univariate and multivariate (full) models (Supplementary Tables 24 – 25).

Publication bias analyses

Publication bias can be present in a meta-analytic data set when experimental results that show strong effects are preferentially published. To test for publication bias in our data set we first visually inspected funnel plots for asymmetry in the distribution of the collected effect sizes, and then performed Egger's regression on residuals of effect-sizes from meta-analytic models and sampling errors^{6,7}. Residuals were calculated from full Bayesian models including study year, student mean age and subject as moderators. Visual inspection of funnel plots (Supplementary Figs. 6 - 7) and results of Egger regression do not indicate publication bias in the school subset, but there is evidence of publication bias in the university subset (Supplementary Table 25). We found no effect of publication year on our *lnRR* and *lnCVR* values in univariate models (Supplementary Table 12).

Inferring grade distributions overlaps

We used simulations to translate the estimated effect sizes for the ratios between mean grades and grade variabilities between the sexes into predicted distributions of student's academic abilities. We assumed that school grades represent student's ability for future success in the academic discipline⁸, and we assumed normal distributions of grades within populations. From the North American school data with grading scale 0-5 we calculated mean and SD of boys' grades for STEM and non-STEM subjects. We then obtained respective mean and SD of girls' grades by utilising mean effect size estimates, expressed as percentages. Using means and SD

values of boys' and girls' grades we generated a sample of 1,000,000 boys and 1,000,000 girls. The sample was used to plot distributions of individual grades (Fig. 3A and 3B), and to estimate the proportion of girls to boys in the top portion of the distribution, for divisions of 0.1% (i.e. 1000 divisions, from 0.001 to 1) (Fig. 3C).

Comparison of STEM and non-STEM grades within the same cohort

In our school data subset, we found girls had a smaller mean advantage in STEM compared to non-STEM subjects. Does this imply that girls are receiving lower grades in STEM than non-STEM, or are boys receiving higher grades in STEM than non-STEM? We cannot test this question directly, because our data does not contain individual-level data, but we can test it at the level of a cohort. To test whether, on averages, students in a class or school are receiving higher grades in STEM compared to non-STEM, we reduced our data to studies that presented mean school grades in STEM and non-STEM subjects for boys and girls from the same cohort (30 cohorts from 25 studies). We then computed the log response ratio ($\ln RR$) as the logged ratio between mean STEM and mean non-STEM grades.

We performed meta-analysis on the within-cohort difference between mean STEM and non-STEM grades, and a meta-regression with sex as a moderator. The results of these analyses are shown in Supplementary Table 11. Overall, there was a non-significant trend for STEM grades to be 3.9% lower than non-STEM grades. This trend was entirely driven by girls, whose STEM grades were significantly 6.7% lower than their non-STEM grades. Boys' grades showed no significant difference between STEM and non-STEM.

Direct comparison to Voyer & Voyer 2014: Standardised Mean Difference

In addition to analysing gender differences in mean grades using $\ln RR$, we also ran analyses using the standardised mean difference (SMD). This allows more direct comparison with the results of Voyer's¹ original meta-analysis, which used Cohen's d as the effect size. We have used a modified version of Cohen's d , called Hedge's d or g^{10} , which has a bias correction for small sample sizes. We calculated SMD and its sampling variance, s_{SMD}^2 , as:

$$SMD = \frac{\bar{x}_f - \bar{x}_m}{s_{\text{pooled}}} J \text{ (Supplementary Equation 1)}$$

$$J = 1 - \frac{3}{4(n_f + n_m - 2) - 1} \text{ (Supplementary Equation 2)}$$

$$s_{\text{pooled}} = \sqrt{\frac{(n_f - 1)s_f^2 + (n_m - 1)s_m^2}{n_f + n_m - 2}} \text{ (Supplementary Equation 3)}$$

$$s_{SMD}^2 = \frac{n_C + n_E}{n_C n_E} + \frac{SMD^2}{2(n_C + n_E)} \text{ (Supplementary Equation 4)}$$

Where:

$\bar{x}_f$ and $\bar{x}_m$ = the mean grade of female and male students, respectively

s_m^2 and s_f^2 = the variance in grades of female and male students, respectively

n_m and n_f = the number of male and female students in each sample, respectively

Our results are similar to Voyer's¹ in both magnitude and significance, with the exception that course material classified as "Other/NR" showed no significant mean difference (our results: Supplementary Table 16; Voyer's results: Table 2, p.1189¹).

Supplementary Note 1

Do gender differences vary with racial composition?

Following Voyer's¹ methods, for the samples from the United States we coded the racial composition of each cohort if the original study reported a >75% racial majority for the student population (for coding details, see Supplementary Table 1). Twelve studies reported a majority Black/African American racial composition ($n = 15$ effect sizes), and eighteen studies reported a White/Caucasian majority ($n = 36$ effect sizes). No studies reported an Asian American majority, and only two studies ($n = 2$ effect sizes) reported a Latino/Hispanic majority, providing insufficient data for a comparison. Therefore, to test for the effects of racial composition on gender differences in school grades, we used the racial composition category of White/Black as a moderator variable in univariate meta-regression models. We found no significant differences in gender differences in either mean or variance (Supplementary Table 13).

Supplementary Note 2

Analysis of test scores – 2015 PISA

To explore whether gender differences in variability across STEM and non-STEM are broadly applicable to school achievement, and not confined to school grades which tend to favour girls, we analysed data from the 2015 Programme for International Student Assessment (referred to as PISA hereafter). PISA is an international measurement of achievement on standardised tests by 15-year-old students.

We downloaded results tables for test performance in reading, mathematics, and science subjects, from the PISA 2015 Results (Volume I)¹¹, and extracted the means

220 and standard deviation in achievement for boys and girls in each jurisdiction
 221 (country). We obtained the corresponding sample sizes from separate tables, available
 222 within the “Questionnaire items” from the Compendia on the PISA 2015 Database
 223 webpage¹². To calculate sample sizes we used the number of “valid” students for each
 224 subject type, and the given percentage of boys and girls within each jurisdiction.

225 Overall, the PISA dataset summarised test scores in maths, reading and science for
 226 226,131 female and 226,480 male students. The students were tested from 64
 227 jurisdictions. The minimum and maximum number of students in a jurisdiction was
 228 3,371 and 23,141, respectively.

229 To test for gender differences in means we used the same metric as our main meta-
 230 analysis ($\ln RR$). To test for differences in variance we used a different metric, because
 231 there was no consistent mean-variance relationship in the PISA dataset (in contrast to
 232 our main dataset). We therefore used the log variability ratio ($\ln VR$) and its sampling
 233 variance ($s^2_{\ln VR}$) to test for gender differences in variance¹³, where:

$$234 \quad \ln VR = \ln \left(\frac{s_f}{s_m} \right) + \frac{1}{2(n_f-1)} - \frac{1}{2(n_m-1)} \text{ (Supplementary Equation 5)}$$

$$235 \quad s^2_{\ln VR} = \frac{1}{2(n_f-1)} + \frac{1}{2(n_m-1)} \text{ (Supplementary Equation 6)}$$

236 We also directly modelled between-subject differences in variability by testing for the
 237 moderating effects of sex and subject on the logged standard deviation ($\ln SD$) in test
 238 scores¹⁴:

$$239 \quad \ln SD = \ln s + \frac{1}{2(n-1)} \text{ (Supplementary Equation 7)}$$

$$240 \quad s^2_{\ln SD} = \frac{1}{2(n-1)} \text{ (Supplementary Equation 8)}$$

We fitted meta-analytic models to each effect size with the same approach as our main analysis (using the *rma.mv* and *robust* functions from the *metafor* package (v. 2.0.0) in *R* (v. 3.4.3)¹⁵). We accounted for non-independence arising from multiple effect sizes from the same study by fitting the ID of the jurisdiction and a comparison ID as random effects. We modelled sampling variances with a covariance matrix, assuming a 0.5 correlation between variances arising from the same jurisdiction. To test for differences between STEM and non-STEM subjects, we fitted univariate meta-regression models by including subject as a fixed effect, where subject was either STEM (maths and science) or non-STEM (reading).

PISA Results

Full results are shown in Supplementary Table 14. Overall, girls sitting the 2015 PISA received 2% higher scores than boys ($\ln RR_{\text{overall}(\text{mean})}$ CI: 1.4% to 2.7%; Supplementary Fig. 11A), with 6.4% less variation among girls than among boys ($\ln VR_{\text{overall}(\text{variance})}$ CI: 5.6% to 7.2%; Supplementary Fig. 11B).

Girls' small advantage in PISA tests scores was entirely driven by a 6.9% advantage in non-STEM ($\ln RR_{\text{STEM}(\text{mean})}$ CI: 6% to 7.7%. In contrast, girls' showed a non-significant 0.3% *disadvantage* in STEM ($\ln RR_{\text{non-STEM}(\text{mean})}$ CI: -0.9% to 0.2%; Supplementary Fig. 11A).

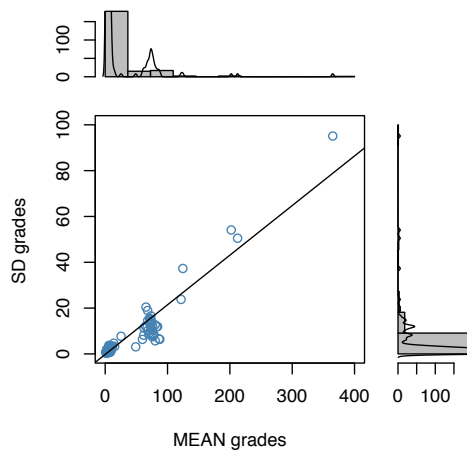
While girls' test scores were significantly more consistent across subjects, this variability gap was no different between non-STEM and STEM subjects ($\ln VR_{\text{non-STEM STEM diff}}$ CI: -0.7% to 0.3%). When looking at the variability for girls and boys separately, both girls' and boys' scores were more variable in non-STEM than STEM subjects, but the difference was smaller for girls (Supplementary Fig. 11C, Supplementary Table 14).

265 Together these results suggest that the shape of the achievement distributions was
266 consistent, but in STEM subjects boys' test scores shifted to the right.

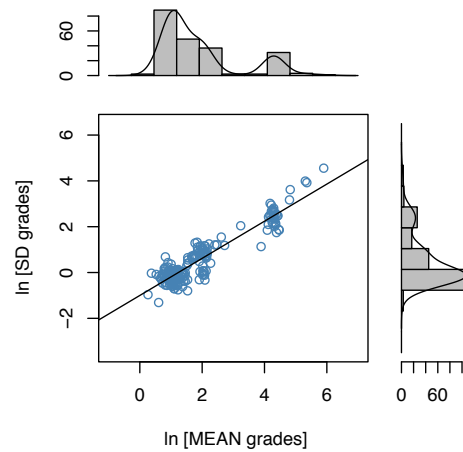
267 **Supplementary Figures**

school pupils

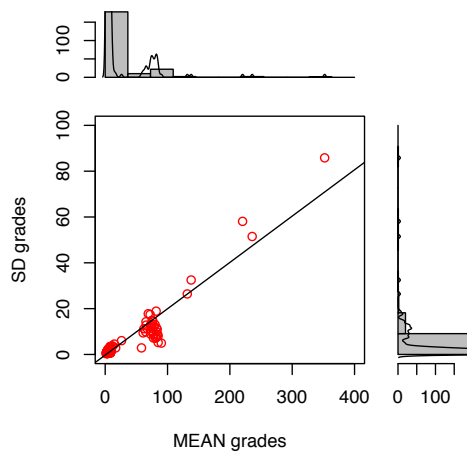
A boys grades - original scales



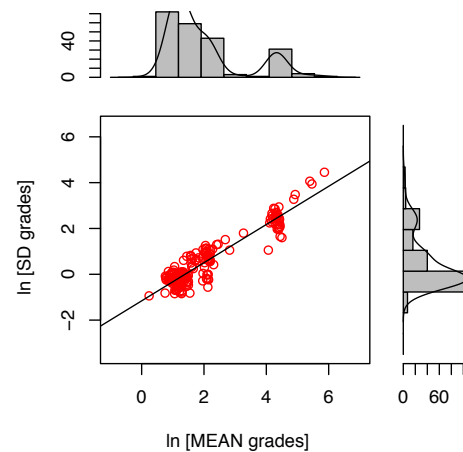
B boys grades - ln scales



C girls grades - original scales



D girls grades - ln scales



268

269 **Supplementary Figure 1**

270 School data subset – relationships between the means and their standard deviations

271 (SD) of grades of boys (**A, B** - blue) and girls (**C, D** - red), on original and log scales.

272 Distributions of means and standard deviations, presented as side histograms, indicate

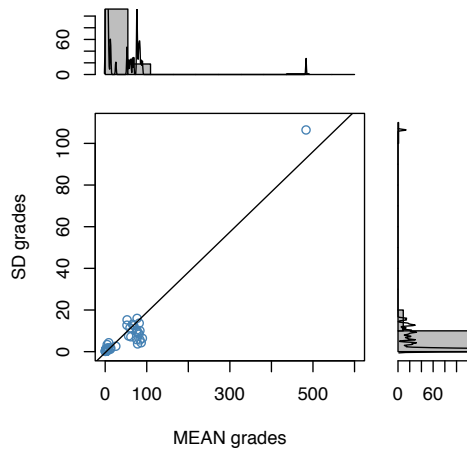
273 that a few different grading scales were used for grading students in the original

274 studies.

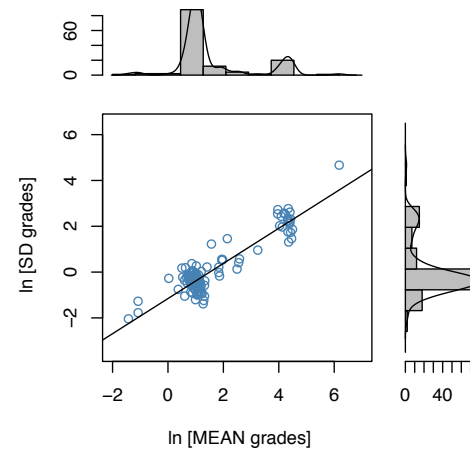
275

university students

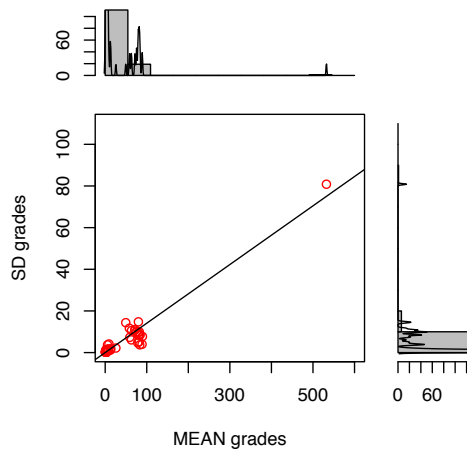
A boys grades - original scales



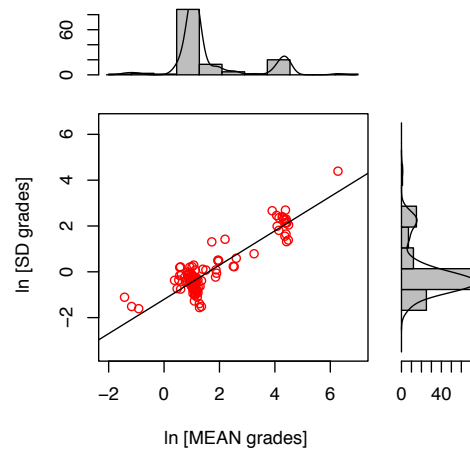
B boys grades - ln scales



C girls grades - original scales



D girls grades - ln scales



276

277 **Supplementary Figure 2**

278 University data subset – relationships between the means and their standard

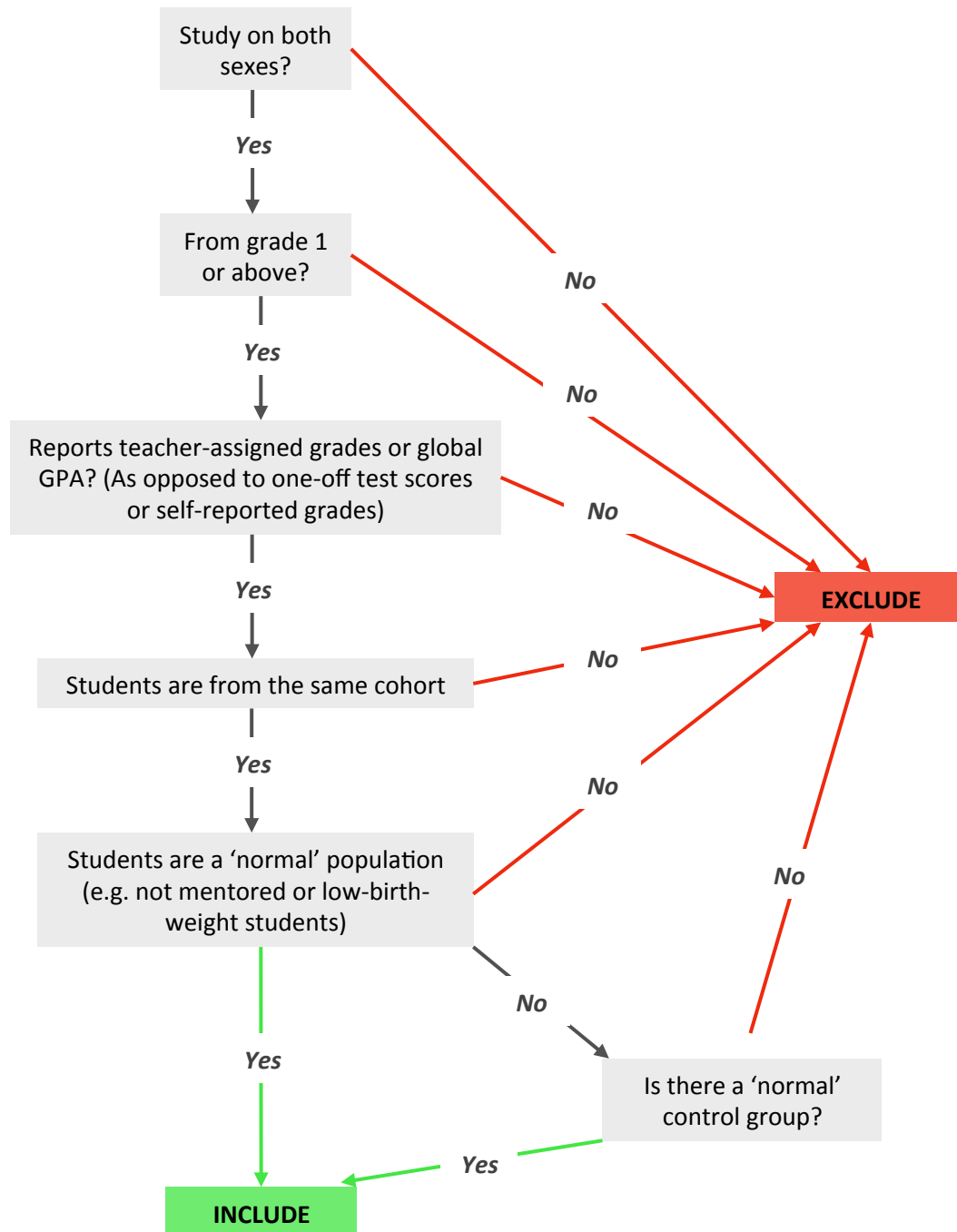
279 deviations (SD) of grades of boys (**A, B** - blue) and girls (**C, D** - red), on original and

280 log scales. Distributions of means and standard deviations, presented as side

281 histograms, indicate that a few different grading scales were used for grading students

282 in the original studies.

283



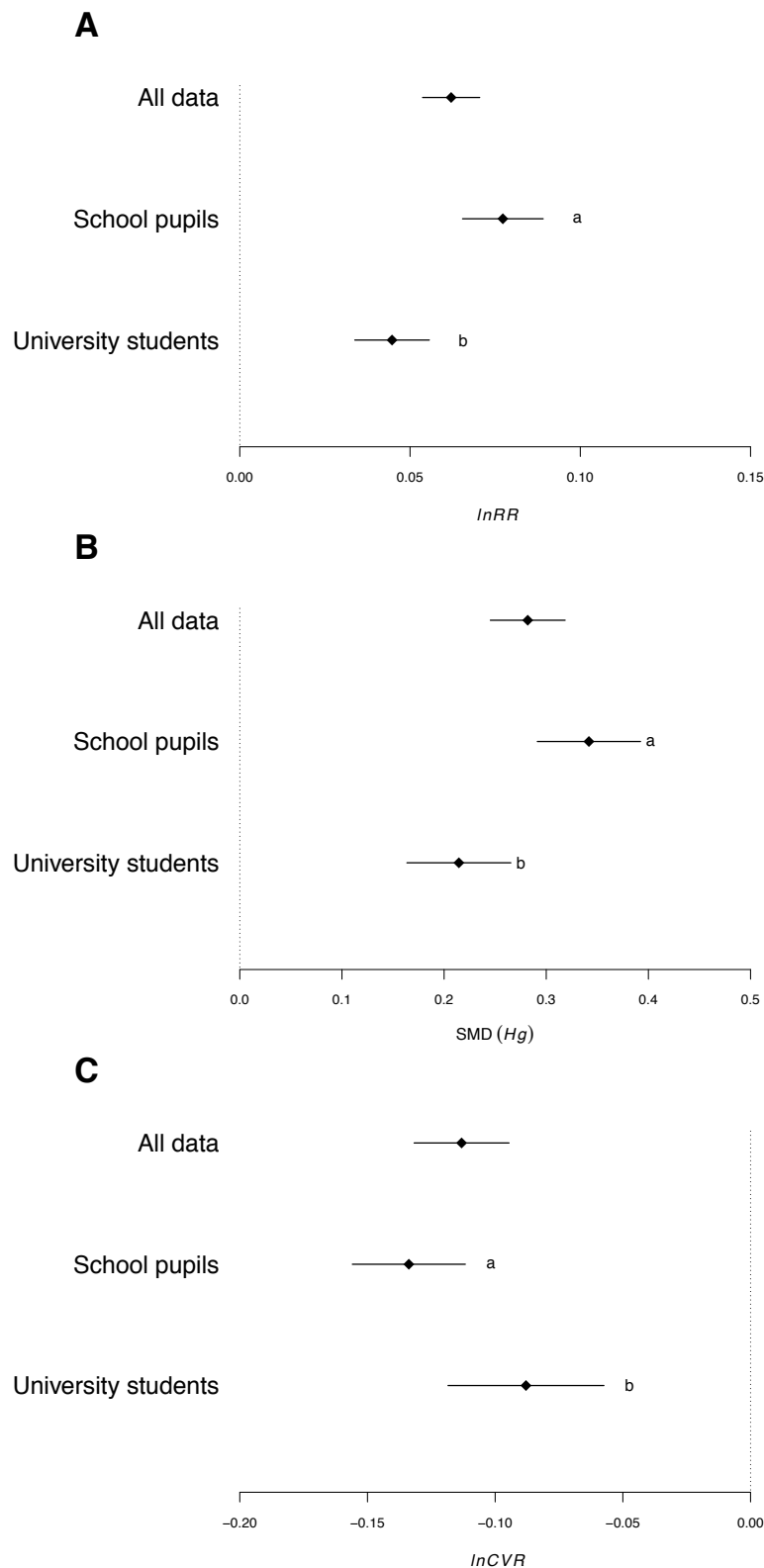
284

285 **Supplementary Figure 3**

286 Inclusion/exclusion decision tree with criteria used to evaluate found studies for
287 inclusion in the meta-analysis.

288

289



Supplementary Figure 4

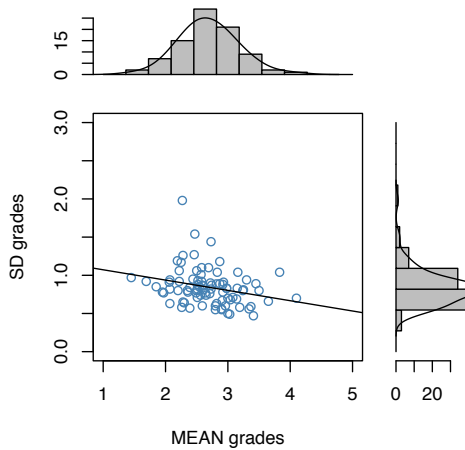
Meta-analytic estimates of mean effect sizes, and 95% confidence intervals, for gender differences in grades in the whole sample (All data), and in the school and

294 university data subsets. (A) and (B) Positive *lnRR* and *SMD* estimates indicate lower
295 boys' mean grades in comparison to girls' mean grades. (C) Negative *lnCVR*
296 estimates indicate greater grade variation for boys in comparison to girls, with the
297 gender gap being larger at school than at university. Lowercase 'a' and 'b' letters
298 indicate that school and university mean estimates are statistically different, with a
299 larger gender gap at school than at university.

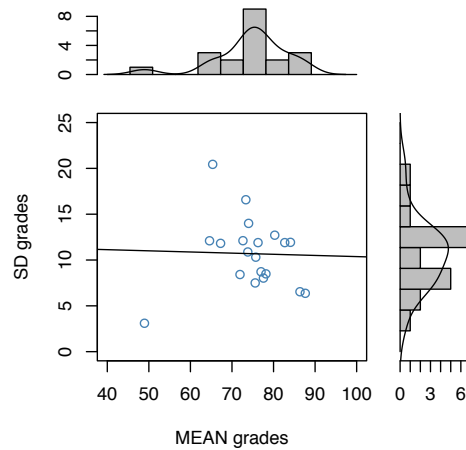
300

North American school pupils

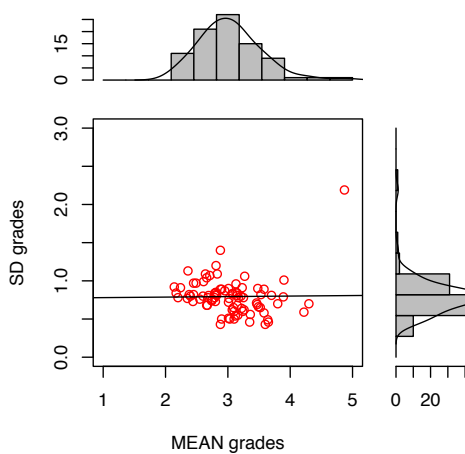
A boys grades - scale 0 to 5



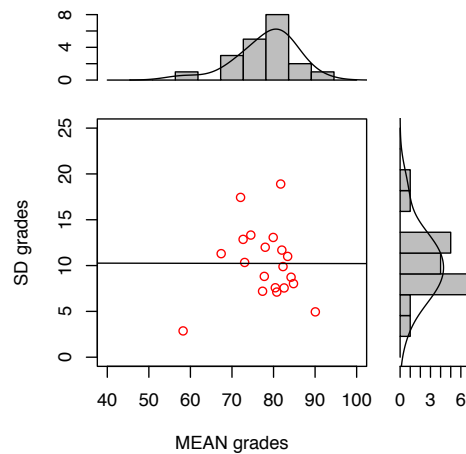
B boys grades - scale 0 to 100



C girls grades - scale 0 to 5



D girls grades - scale 0 to 100

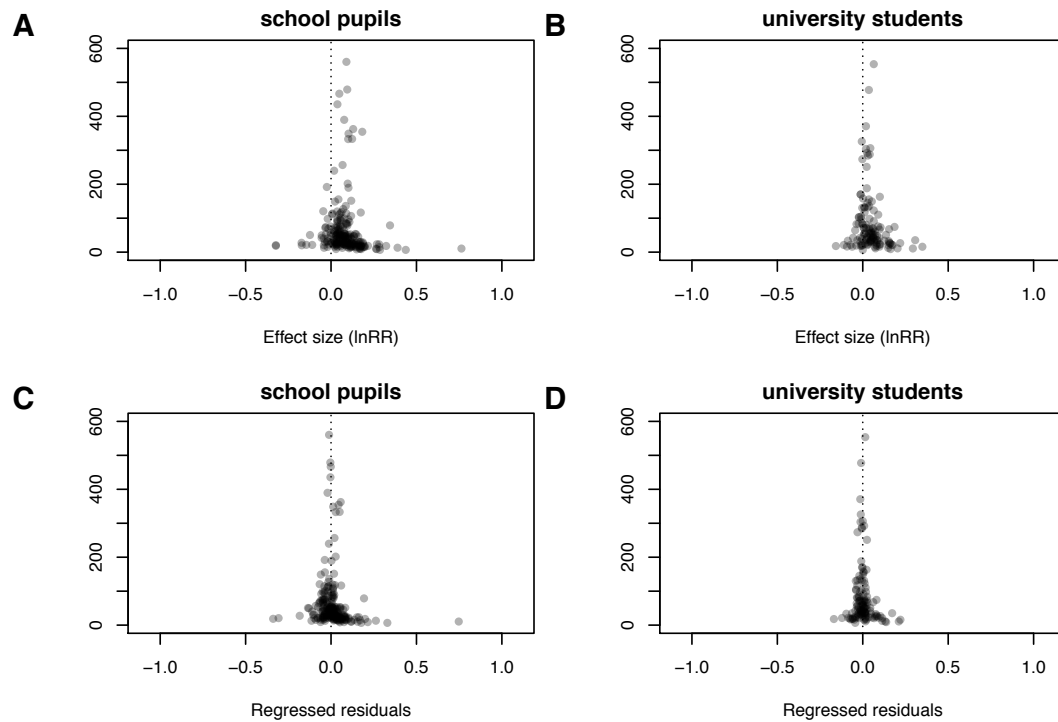


301

302 **Supplementary Figure 5**

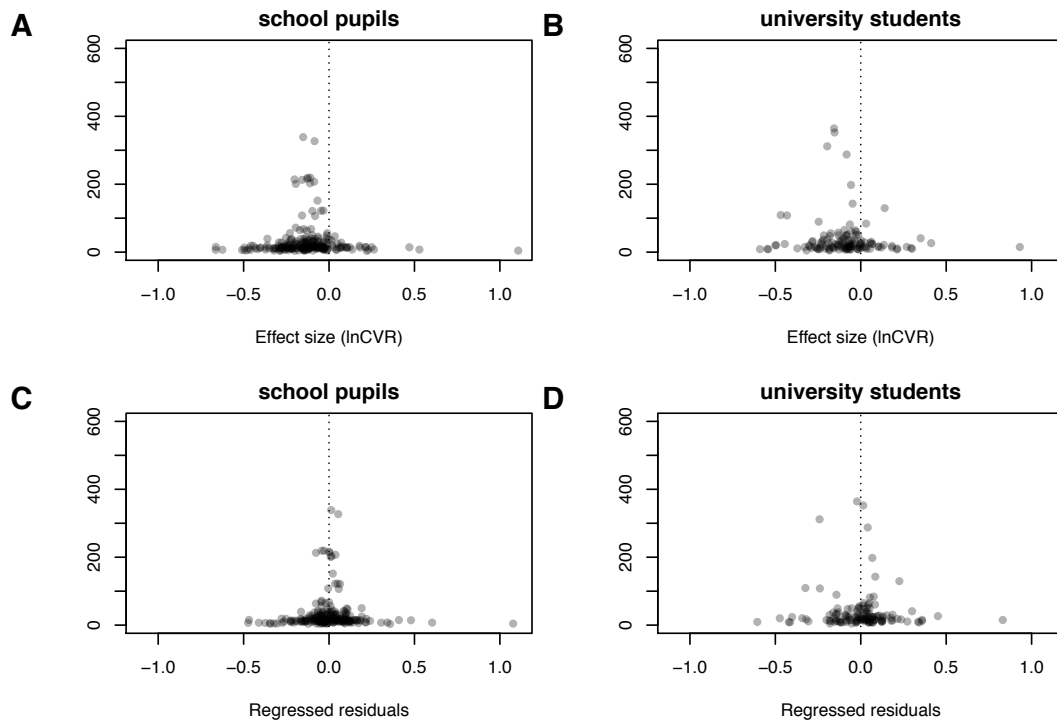
303 School data subsets for North American students graded on scales 0-5 (82 effect
 304 sizes) and 0-100 (23 effect sizes) – relationships between the means and their standard
 305 deviations (*SD*) of grades of boys (**A, B** - blue) and girls (**C, D** - red), on both scales.
 306 Distributions of means and standard deviations presented as side histograms do not
 307 indicate ceiling effect in these data subsets.

308



Supplementary Figure 6

Funnel plots for differences in mean grades of boys and girls ($\ln RR$). Panels **A** and **B** are based on raw effect sizes and their precision (inversed square root of standard errors). Panels **C** and **D** are based on the residual values from the full meta-regression models, containing all moderators, and their precision.

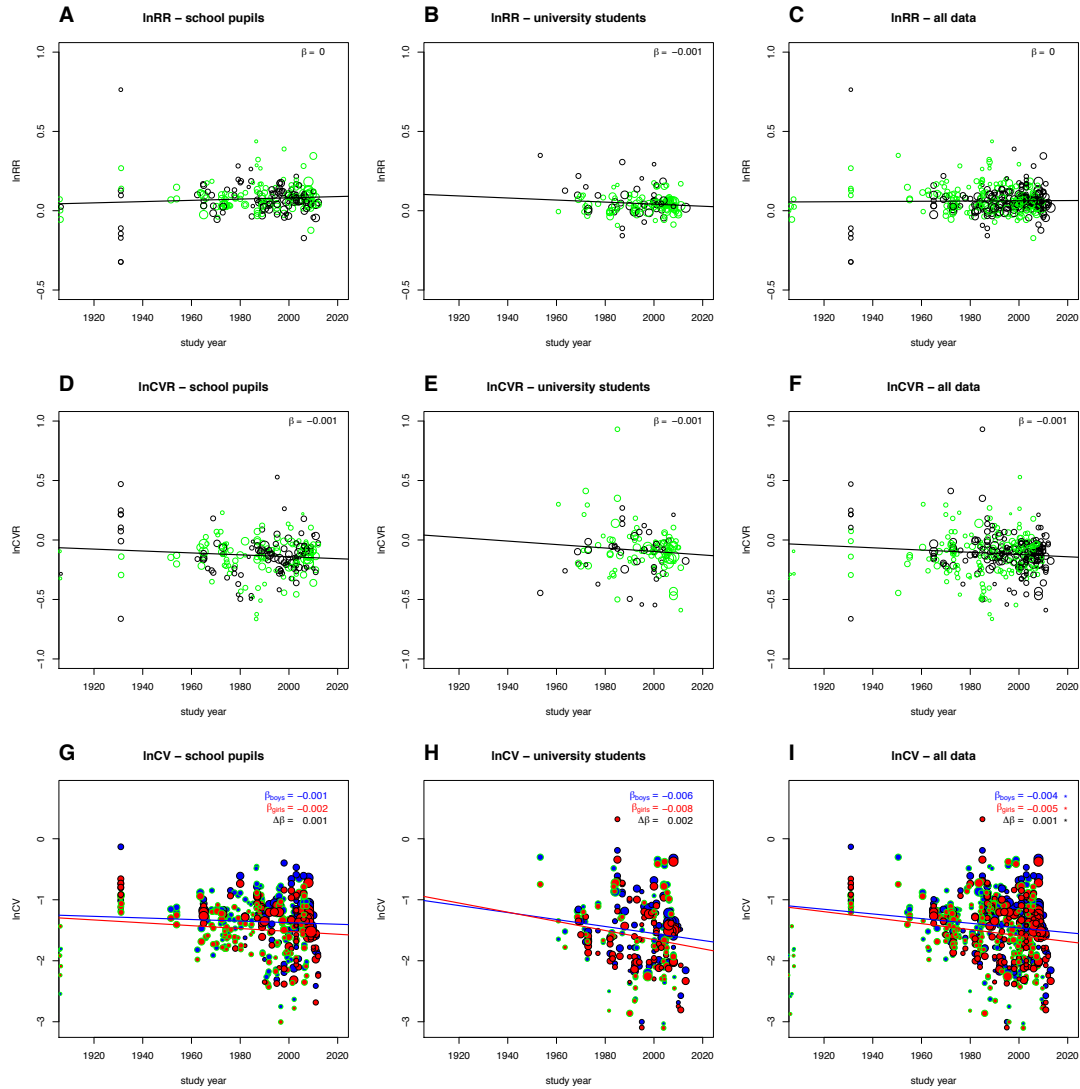


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317 **Supplementary Figure 7**

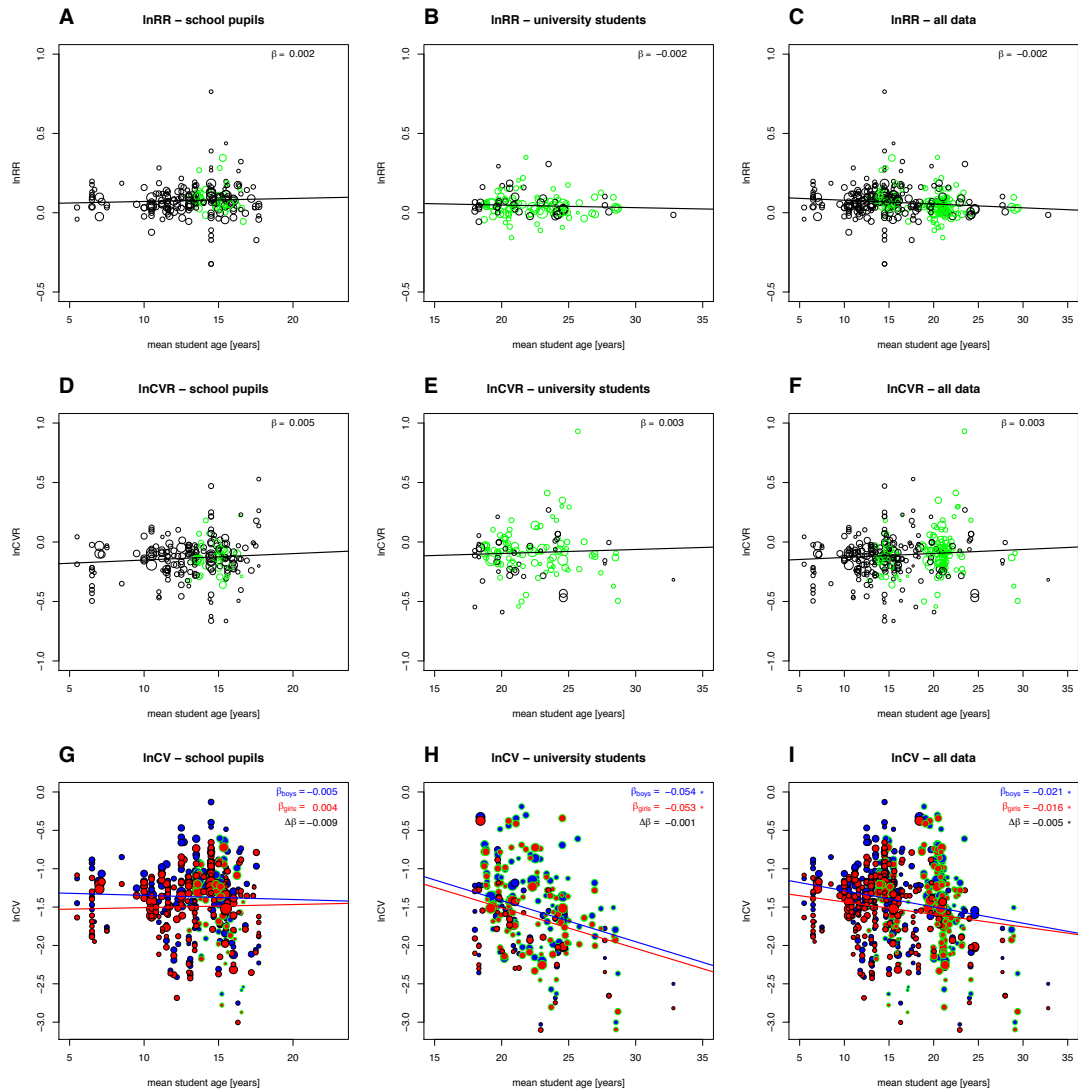
318 Funnel plots for differences in variation of grades of boys and girls ($\ln\text{CVR}$). Panels **A**
 319 and **B** are based on raw effect sizes and their precision (inversed square root of
 320 standard errors). Panels **C** and **D** are based on the residual values from the full meta-
 321 regression models, containing all moderators, and their precision.

322



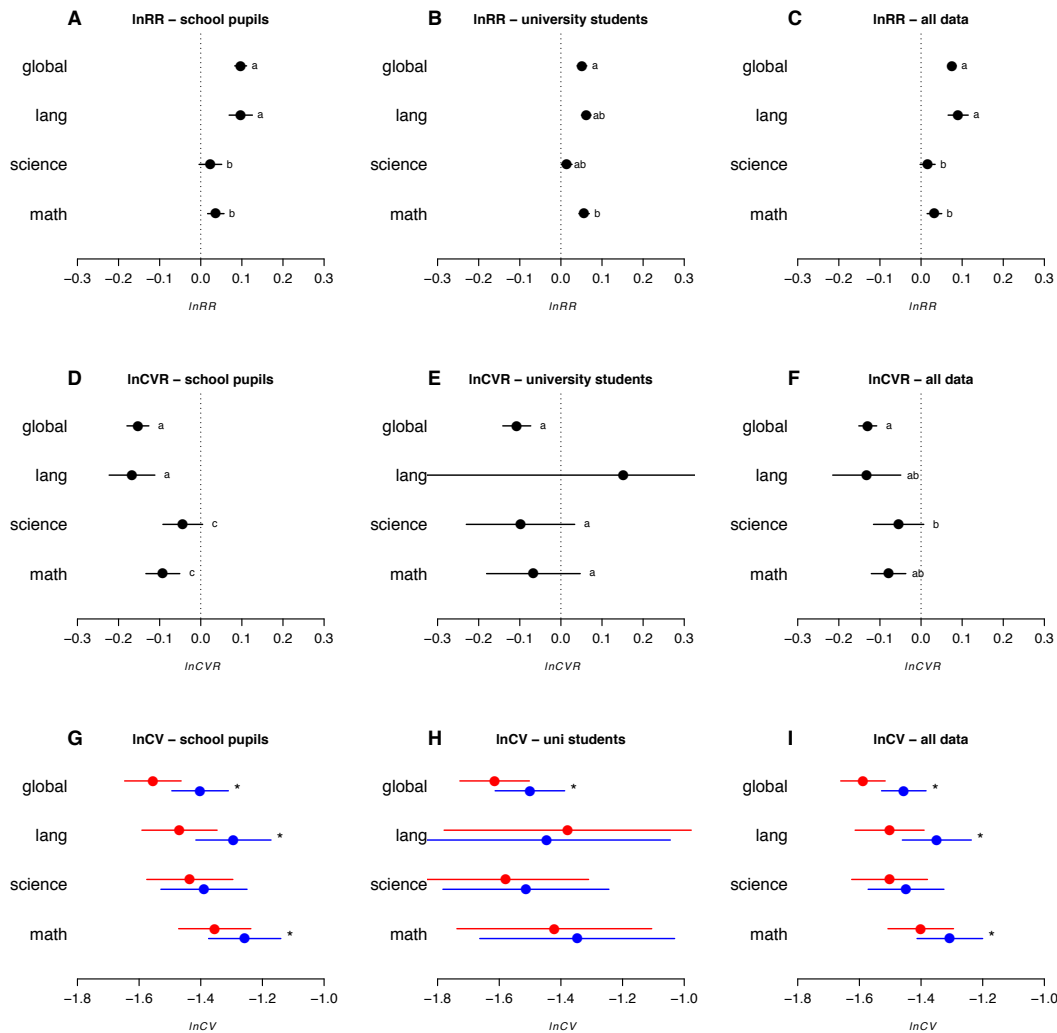
Supplementary Figure 8

Scatter plots for effect sizes against study year for the school subset (panels **A**, **D** and **G**), the university subset (**B**, **E** and **H**), and the whole dataset (**C**, **F** and **I**). Green circles represent imputed study year data. In panels **G**, **H** and **I**, blue points represent data for boys' grades, and red points represent data for girls' grades.



Supplementary Figure 9

Scatter plots for effect sizes against mean student age for the school subset (panels **A**, **D** and **G**), the university subset (**B**, **E** and **H**), and the whole dataset (**C**, **F** and **I**). Green circles represent imputed mean age data. In panels **G**, **H** and **I**, blue points represent data for boys' grades, and red points represent data for girls' grades.



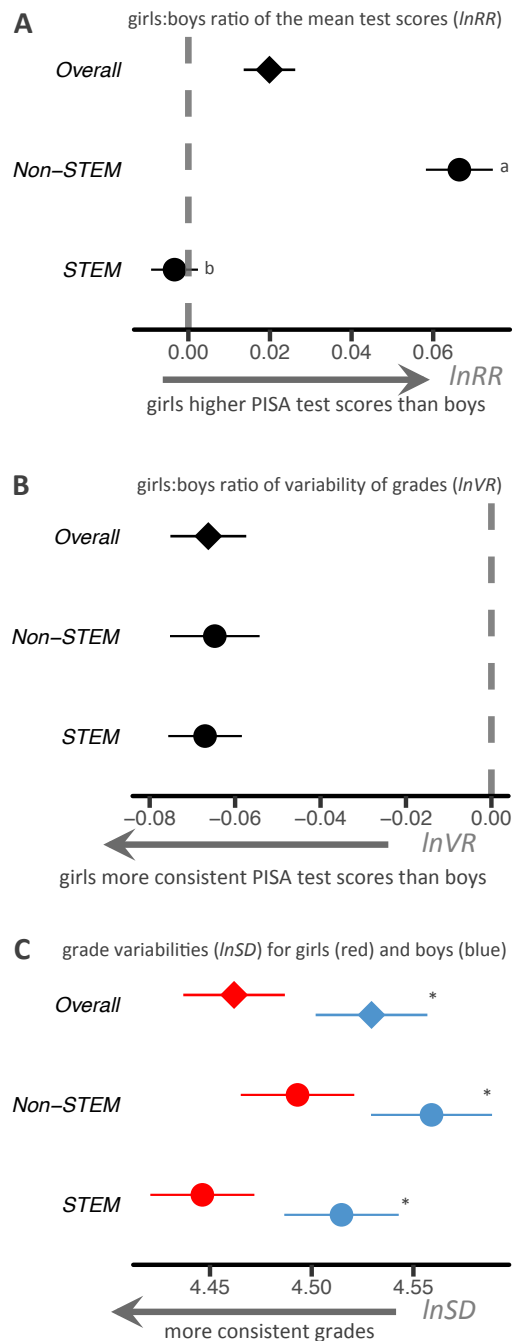
Supplementary Figure 10

Meta-analytic estimates of mean effect sizes (circles), and their 95% confidence intervals (whiskers), for gender differences in school grades according to subject type.

First column: panels **A**, **D** and **G** – the school data subset. Second column: panels **B**, **E** and **H** – university data subset. Third column: panels **C**, **F** and **I** – whole sample.

Blue represents data for boys' grades, and red represents data for girls' grades.

Positive values of $\ln RR$ indicate lower male mean grades. Negative values of $\ln CVR$ indicate greater male variability in grades. Decreasing (more negative) values of $\ln CV$ indicate reduced variance.



Supplementary Figure 11

Results of analyses on (A) ratios of the grade means, (B) ratios of grade variabilities, and (C) coefficients of variations for girls (red) and boys (blue) from the 2015 PISA, corresponding to Supplementary Table 14. Diamonds and circles represent meta-analytic estimates of mean effect sizes, and their 95% confidence intervals are drawn as whiskers. In panel A, natural logarithm of response ratio ($\ln RR$) represents the

354 average difference between girls' and boys' test scores; positive values of $\ln RR$
355 indicate lower boys' tests scores. In panel **B**, natural logarithm of variation ratio
356 ($\ln VR$) represents the average difference in test score variation between boys and
357 girls; negative values of $\ln VR$ indicate greater male variance. In panel **C**, natural
358 logarithms of standard deviation in test scores ($\ln SD$) are shown for girls and boys to
359 illustrate grade variation by gender; more negative values of $\ln SD$ indicate less
360 variation.
361

Supplementary Tables

Supplementary Table 1

Additional moderator variables in the dataset. A full list of extracted and coded variables can be downloaded from a dedicated repository on the Open Science Framework ¹⁶.

Variable	Extracted by Voyer & Voyer (2014)	Type	Categories	Description
Study_ID	No	Categorical	Unique ID for each study	Used to identify each study in the dataset
Cohort_ID	No	Categorical	Unique ID for each cohort	Used to identify each cohort to avoid pseudoreplication, e.g., 1 study reports scores for mathematics and language for the same group of students
Published	Yes	Categorical	Yes	Study is published
			No	Study is not published
Nationality	Yes	Categorical	North America	United States, Canada
			Other	Everywhere else
Year of publication	Yes	Continuous		Study publication year
Year of study	no	Continuous		Recorded when reported by the study, otherwise left blank. When multiple years are covered (such as a high school GPA), the graduation year was recorded for school samples, and the most recent enrolment year was recorded for university samples
Racial composition	yes	Categorical	White	>= 75% White US sample
			Black	>= 75% Black US sample
			Hispanic	>= 75% Hispanic US sample
			Asian	>= 75% Asian US sample
			Diverse	No racial majority US sample
			Not Reported	Race not reported for US

				sample
			non-U.S.	Sample not from US
Male to female ratio	yes	Categorical	m>f	More males than females in the cohort
			f=m	Equal number of males and females in the cohort
			f>m	More females than males in the cohort
			Estimated	Sample sizes not split by gender, so equal sex ratio is assumed in the cohort
Sex ratio	no	Continuous		number of females/number of males in the cohort
Source of data	yes	Categorical	Elementary	Grades 1-6, ages 6-12
			Junior/middle	Grades 7-8, ages 13-14
			High school	Grades 9-12, ages 15-18
			Undergraduate	University / college
			Graduate	Post-graduate university / college
Mean age	no	Continuous		Where the mean age is reported, that value was extracted. For school samples, the mean age was estimated based on the grade level (e.g. year 4 students are recorded as 9.5 years). Age is not estimated for university samples unless explicitly reported by the study.

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Supplementary Table 2

Estimated effect sizes for meta-analytic (intercept-only) models for grade difference between girls and boys, showing results for both variance (*lnCVR*) and mean differences (*lnRR* and *SMD*). Negative *lnCVR* values can be interpreted as girls having less variable grades than boys. Positive values of *lnRR* and *SMD* can be interpreted as girls having higher grades than boys. Effects with confidence intervals (CI) not crossing zero are indicated in bold. Sigma² for cohort_ID is the amount of variance attributed to cohort identity (N levels = number of cohorts), and Sigma² for comp_ID represents residual variance for each unit of analysis (N levels = number of comparisons).

Measure	Data	Fixed effects			Random effects		
		Mean	CI.lb	CI.ub		Sigma ²	N levels
School subset							
lnCVR	Intercept	0.131	0.154	0.109	cohort_ID	0.00	141
					comp_ID	0.02	215
lnRR	Intercept	0.074	0.061	0.086	cohort_ID	0.00	141
					comp_ID	0.01	215
SMD	Intercept	0.330	0.280	0.38	cohort_ID	0.01	141
					comp_ID	0.08	215
University subset							
lnCVR	Intercept	-0.09	0.122	0.059	cohort_ID	0.00	122
					comp_ID	0.03	131
lnRR	Intercept	0.044	0.034	0.055	cohort_ID	0.00	122
					comp_ID	0.00	131
SMD	Intercept	0.213	0.162	0.264	cohort_ID	0.07	122
					comp_ID	0.00	131
All data							
lnCVR	Intercept	0.114	0.133	0.095	cohort_ID	0.00	261

					comp_ID	0.02	346
<i>lnRR</i>	Intercept	0.061	0.052	0.070	cohort_ID	0.00	261
					comp_ID	0.01	346
<i>SMD</i>	Intercept	0.279	0.243	0.315	cohort_ID	0.01	261
					comp_ID	0.07	346

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Supplementary Table 3

Estimated effect sizes for univariate meta-regression models with student population type (school pupils vs. university students) as a moderator. *lnCVR* is a measure of grade consistency of girls and boys and *lnRR* and *SMD* are measures of mean grade difference between girls and boys. Negative *lnCVR* values can be interpreted as girls having less variable grades than boys and positive *lnRR* and *SMD* values can be interpreted as girls having higher grades than boys. Effects with confidence intervals (CI) not crossing zero are indicated in bold. Effects with confidence intervals (CI) not crossing zero are indicated in bold. Number of effect sizes within each subset: school = 215; university = 131.

Measure	Data	Fixed effects		
		Mean	CI.lb	CI.ub
<i>lnCVR</i>				
	School intercept	-0.134	-0.157	-0.112
	University intercept	-0.087	-0.117	-0.057
	School – University contrast	0.041	0.002	0.080
<i>lnRR</i>				
	School intercept	0.076	0.064	0.088
	University intercept	0.045	0.034	0.056
	School – University contrast	-0.028	-0.044	-0.011
<i>SMD</i>				
	School intercept	0.339	0.289	0.388
	University intercept	0.213	0.161	0.264
	School – University contrast	-0.118	-0.188	-0.047

Supplementary Table 4

Estimated effect sizes for univariate meta-regression models with study year as a moderator. *lnCVR* is a measure of grade consistency of girls and boys and *lnRR* and *SMD* are measures of mean grade difference between girls and boys. Positive slope values can be interpreted as the difference between mean grades of boys and girls increasing over historical time. Estimates with confidence intervals (CI) not crossing zero are indicated in bold. “N missing” indicates how many missing values of study year moderator had to be imputed.

Measure	Data		Fixed effects			
			Mean	CI.lb	CI.ub	N missing
<i>lnCVR</i>	School subset					
		Study year	-0.001	-0.003	0.001	121
		Study year scaled	-0.029	-0.083	0.025	121
<i>lnCVR</i>	University subset					
		Study year	-0.002	-0.005	0.001	52
		Study year scaled	-0.017	-0.053	0.019	52
<i>lnCVR</i>	All data					
		Study year	-0.001	-0.003	0.001	173
		Study year scaled	-0.020	-0.052	0.013	173
<i>lnRR</i>	School subset					
		Study year	0.000	0.000	0.001	121
		Study year scaled	0.019	-0.017	0.055	121
<i>lnRR</i>	University subset					
		Study year	-0.001	-0.002	0.000	52
		Study year scaled	-0.010	-0.028	0.008	52
<i>lnRR</i>	All data					
		Study year	0.000	-0.001	0.001	173
		Study year scaled	0.005	-0.015	0.026	173
<i>SMD</i>	School subset					
		Study year	0.002	-0.001	0.006	121
		Study year scaled	0.067	-0.018	0.152	121
<i>SMD</i>	University subset					

<i>SMD</i>	All data	Study year	-0.001	-0.005	0.004	52
		Study year scaled	-0.007	-0.065	0.050	52
		Study year	0.001	-0.001	0.003	173
		Study year scaled	0.031	-0.010	0.073	173

Supplementary Table 5

Estimated slopes from univariate meta-regression models with sex and study year as moderators (with the interaction) for variation in grades of girls and boys (*lnCV*), using imputed data. Estimates with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included in the analysis.

Measure	Data	Fixed effects			<i>k</i>
		Mean	CI.lb	CI.ub	
<i>lnCV</i>	School subset				430
		Study year: girls	-0.002	-0.006	0.002
		Study year scaled: girls	-0.064	-0.154	0.026
		Study year: boys	-0.001	-0.005	0.003
		Study year scaled: boys	-0.032	-0.122	0.058
		Study year: boys-girls	0.001	0.000	0.002
		Study year scaled: boys-girls	0.032	0.004	0.060
<i>lnCV</i>	University subset				262
		Study year: girls	-0.002	-0.006	0.002
		Study year scaled: girls	-0.064	-0.154	0.026
		Study year: boys	-0.001	-0.005	0.003
		Study year scaled: boys	-0.032	-0.122	0.058
		Study year: boys-girls	0.001	0.000	0.002
		Study year scaled: boys-girls	0.032	0.004	0.060
<i>lnCV</i>	All data				692
		Study year: girls	-0.002	-0.006	0.002
		Study year scaled: girls	-0.064	-0.154	0.026
		Study year: boys	-0.001	-0.005	0.003
		Study year scaled: boys	-0.032	-0.122	0.058
		Study year: boys-girls	0.001	0.000	0.002
		Study year scaled: boys-girls	0.032	0.004	0.060

Supplementary Table 6

Estimated effect sizes for univariate meta-regression models with mean age of students as a moderator. *lnCVR* is a measure of grade consistency of girls and boys and *lnRR* and *SMD* are measures of mean grade difference between girls and boys. Positive slope values can be interpreted as the difference between mean grades of boys and girls increasing over historical time. Estimates with confidence intervals (CI) not crossing zero are indicated in bold. “N missing” indicates how many missing values of study year moderator had to be imputed.

Measure	Data		Fixed effects			
			Mean	CI.lb	CI.ub	N missing
<i>lnCVR</i>	School subset	Age	0.010	-0.001	0.021	37
		Age scaled	0.028	-0.003	0.060	37
<i>lnCVR</i>	University subset	Age	0.000	-0.020	0.019	91
		Age scaled	-0.002	-0.065	0.061	91
<i>lnCVR</i>	All data	Age	0.004	0.000	0.009	128
		Age scaled	0.018	-0.005	0.041	128
<i>lnRR</i>	School subset	Age	0.002	-0.003	0.007	37
		Age scaled	0.006	-0.008	0.019	37
<i>lnRR</i>	University subset	Age	-0.002	-0.010	0.006	91
		Age scaled	-0.006	-0.034	0.022	91
<i>lnRR</i>	All data	Age	-0.002	-0.004	-0.001	128
		Age scaled	-0.011	-0.020	-0.003	128
<i>SMD</i>	School subset	Age	0.015	-0.006	0.036	37
		Age scaled	0.043	-0.019	0.105	37
<i>SMD</i>	University subset	Age	0.003	-0.025	0.031	91
		Age scaled	0.014	-0.079	0.107	91
<i>SMD</i>	All data	Age	-0.006	-0.014	0.002	128
		Age scaled	-0.020	-0.061	0.021	128

Supplementary Table 7

Estimated slopes from univariate meta-regression models with sex and student mean age as moderators (with the interaction) for variation in grades of girls and boys ($\ln CV$), using imputed data. Estimates with confidence intervals (CI) not crossing zero are indicated in bold. k – number of effect sizes included in the analysis.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	k
$\ln CV$	School subset				430
		Age: girls	0.004	-0.023	0.030
		Age scaled: girls	0.010	-0.067	0.087
		Age: boys	-0.009	-0.035	0.018
		Age scaled: boys	-0.025	-0.101	0.052
		Age: boys-girls	-0.012	-0.022	-0.002
		Age scaled: boys-girls	-0.035	-0.062	-0.007
$\ln CV$	University subset				262
		Age: girls	-0.054	-0.100	-0.008
		Age scaled: girls	-0.188	-0.327	-0.049
		Age: boys	-0.053	-0.092	-0.015
		Age scaled: boys	-0.185	-0.305	-0.066
		Age: boys-girls	0.000	-0.018	0.019
		Age scaled: boys-girls	0.003	-0.057	0.062
$\ln CV$	All data				692
		Age: girls	-0.017	-0.031	-0.004
		Age scaled: girls	-0.102	-0.169	-0.035
		Age: boys	-0.022	-0.035	-0.008
		Age scaled: boys	-0.122	-0.188	-0.056
		Age: boys-girls	-0.005	-0.009	0.000
		Age scaled: boys-girls	-0.020	-0.041	0.001

Supplementary Table 8

Estimated effect sizes for univariate meta-regression models with subjects (STEM, non-STEM, etc.) as a moderator for mean grade difference between girls and boys (*lnRR* and *SMD*). Positive values can be interpreted as girls having higher grades than boys. Effects with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	<i>k</i>
<i>lnRR</i>	School subset				215
		Subject: STEM	0.031	0.011	0.051
		Subject: Non-STEM	0.075	0.049	0.102
		Subject: Global	0.098	0.084	0.113
		Subject: Other/NR	0.115	0.012	0.218
		non-STEM - STEM difference	-0.044	-0.065	-0.024
<i>lnRR</i>	University subset				131
		Subject: STEM	0.024	0.002	0.045
		Subject: Non-STEM	0.038	0.012	0.064
		Subject: Global	0.054	0.042	0.066
		Subject: Other/NR	-0.001	-0.025	0.023
<i>lnRR</i>	All data				346
		Subject: STEM	0.025	0.009	0.041
		Subject: Non-STEM	0.065	0.041	0.089
		Subject: Global	0.077	0.067	0.087
		Subject: Other/NR	0.031	-0.015	0.077
<i>SMD</i>	School subset				215
		Subject: STEM	0.132	0.069	0.194
		Subject: Non-STEM	0.356	0.269	0.443
		Subject: Global	0.438	0.366	0.51

		Subject: Other/NR	0.491	0.196	0.785	
		non-STEM - STEM difference	-0.224	-0.303	-0.146	
<i>SMD</i>	University subset					131
		Subject: STEM	0.104	0.007	0.201	
		Subject: Non-STEM	0.119	-0.02	0.259	
		Subject: Global	0.261	0.204	0.318	
		Subject: Other/NR	0.031	-0.168	0.229	
<i>SMD</i>	All data					346
		Subject: STEM	0.109	0.055	0.163	
		Subject: Non-STEM	0.307	0.225	0.389	
		Subject: Global	0.349	0.302	0.396	
		Subject: Other/NR	0.163	-0.057	0.383	

Supplementary Table 9

Estimated effect sizes for univariate meta-regression models with subject (STEM, non-STEM, etc.) as a moderator for grade consistency of girls and boys (*lnCVR*). Negative *lnCVR* values can be interpreted as girls having less variable grades than boys. Effects with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	<i>k</i>
<i>lnCVR</i>	School subset				215
	Subject: STEM	-0.079	-0.115	-0.043	
	Subject: Non-STEM	-0.149	-0.199	-0.099	
	Subject: Global	-0.153	-0.179	-0.126	
	Subject: Other/NR	-0.343	-0.546	-0.141	
	non-STEM - STEM difference	0.070	0.028	0.111	
<i>lnCVR</i>	University subset				131
	Subject: STEM	-0.071	-0.144	0.001	
	Subject: Non-STEM	0.044	-0.259	0.347	
	Subject: Global	-0.114	-0.148	-0.079	
	Subject: Other/NR	-0.007	-0.115	0.101	
<i>lnCVR</i>	All data				346
	Subject: STEM	-0.075	-0.109	-0.042	
	Subject: Non-STEM	-0.121	-0.186	-0.055	
	Subject: Global	-0.132	-0.154	-0.110	
	Subject: Other/NR	-0.091	-0.218	0.036	

Supplementary Table 10

Estimated effect sizes for univariate meta-regression models with subject (STEM, non-STEM, etc.) and sex as moderators (with the interaction) for variation in grades of girls and boys (*lnCV*). Intercept values are presented for each sex-subject combination separately alongside with girls-boys contrast for each subject. Effects with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included in the analysis.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	<i>k</i>
<i>lnCV</i>	School subset: girls intercepts				432
		Subject: STEM	-1.371	-1.471	
		Subject: Non-STEM	-1.472	-1.591	
		Subject: Global	-1.556	-1.651	
	boys intercepts	Subject: Other/NR	-1.547	-2.210	
		Subject: STEM	-1.292	-1.393	
		Subject: Non-STEM	-1.322	-1.451	
		Subject: Global	-1.402	-1.500	
	girls-boys difference	Subject: Other/NR	-1.287	-1.939	
		Subject: STEM	0.079	0.040	
		Subject: Non-STEM	0.150	0.093	
		Subject: Global	0.153	0.126	
		Subject: Other/NR	0.260	0.130	
<i>lnCV</i>	University subset: girls intercepts				262
		Subject: STEM	-1.514	-1.713	
		Subject: Non-STEM	-1.358	-1.755	

	boys intercepts	Subject: Global	-1.615	-1.723	-1.507
		Subject: Other/NR	-2.071	-2.579	-1.563
		Subject: STEM	-1.443	-1.633	-1.253
		Subject: Non-STEM	-1.403	-1.562	-1.245
		Subject: Global	-1.501	-1.606	-1.397
	girls-boys difference	Subject: Other/NR	-2.063	-2.629	-1.498
		Subject: STEM	0.070	-0.003	0.144
		Subject: Non-STEM	-0.046	-0.356	0.265
		Subject: Global	0.114	0.079	0.149
		Subject: Other/NR	0.008	-0.104	0.119
<i>lnCV</i>	All data girls intercepts				694
		Subject: STEM	-1.433	-1.525	-1.341
		Subject: Non-STEM	-1.511	-1.628	-1.394
		Subject: Global	-1.588	-1.658	-1.518
		Subject: Other/NR	-1.689	-2.207	-1.172
	boys intercepts	Subject: STEM	-1.357	-1.452	-1.261
		Subject: Non-STEM	-1.388	-1.503	-1.274
		Subject: Global	-1.456	-1.526	-1.386
		Subject: Other/NR	-1.605	-2.185	-1.025
	girls-boys difference	Subject: STEM	0.076	0.042	0.111
		Subject: Non-STEM	0.123	0.054	0.191
		Subject: Global	0.132	0.110	0.155
		Subject: Other/NR	0.085	-0.034	0.204

Supplementary Table 11

Estimated effect sizes for meta-analytic (intercept-only) models for mean school grade difference between STEM and non-STEM subject grades (*lnRR*), for cohorts where we had grade data available for both STEM and non-STEM subjects. Negative mean *lnRR* estimates can be interpreted as STEM grades being on average lower than non-STEM. Effects with confidence intervals (CI) not crossing zero are indicated in bold. The data represents 30 cohorts of pupils from 25 studies.

Measure	Model		Fixed effects				Random effects	
			Mean	CI.lb	CI.ub		Sigma ²	N levels
<i>lnRR</i>	Meta-analysis	Intercept (girls + boys)	-0.040	-0.108	0.027	cohort_ID	0.033	30
						comp_ID	0.003	60
<i>lnRR</i>	Meta-regression	Girls intercept	-0.069	-0.137	-0.001	cohort_ID	0.034	30
		Boys intercept	-0.009	-0.077	0.060	comp_ID	0.001	60

Supplementary Table 12

Estimated effect sizes for univariate meta-regression models with publication year as a moderator. *lnCVR* is a measure of grade consistency of girls and boys and *lnRR* and *SMD* are measures of mean grade difference between girls and boys. Positive slope values can be interpreted as the difference between mean grades of boys and girls increasing over historical time. Estimates with confidence intervals (CI) not crossing zero are indicated in bold. “N missing” indicates how many missing values of study year moderator had to be imputed.

Measure	Data		Fixed effects			
			Mean	CI.lb	CI.ub	k
<i>lnCVR</i>	School subset					
		Publication year	-0.001	-0.002	0.001	215
<i>lnCVR</i>	University subset					
		Publication year	-0.002	-0.004	0.001	131
<i>lnCVR</i>	All data					
		Publication year	-0.001	-0.002	0.000	346
<i>lnRR</i>	School subset					
		Publication year	0.000	-0.001	0.002	215
<i>lnRR</i>	University subset					
		Publication year	0.000	-0.001	0.000	131
<i>lnRR</i>	All data					
		Publication year	0.000	-0.001	0.001	346
<i>SMD</i>	School subset					
		Publication year	0.002	-0.002	0.005	215
<i>SMD</i>	University subset					
		Publication year	0.000	-0.004	0.003	131
<i>SMD</i>	All data					
		Publication year	0.001	-0.001	0.004	346

Supplementary Table 13

Estimated effect sizes for univariate meta-regression models on gender differences in school grades among North Americans students, with race of students (>75% Black or White) as a moderator. *lnCVR* is a measure of grade consistency of girls and boys and *lnRR* and *SMD* are measures of mean grade difference between girls and boys. A positive contrast for *lnCVR* can be interpreted as White students having a smaller gender difference in grade variability than Black students, whereas a negative contrast for *lnRR* and *SMD* can be interpreted as White students having a smaller gender difference in mean grades than Black students. Estimates with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	<i>k</i>
<i>lnCVR</i>	North American School subset				51
	Black	-0.174	-0.269	-0.079	
	White	-0.151	-0.203	-0.099	
	Black-White Contrast	0.023	-0.085	0.132	
<i>lnRR</i>	North American School subset				51
	Black	0.129	0.074	0.184	
	White	0.079	0.055	0.103	
	Black-White Contrast	-0.049	-0.109	0.010	
<i>SMD</i>	North American School subset				51
	Black	0.492	0.384	0.600	
	White	0.385	0.154	0.615	
	Black-White Contrast	-0.108	-0.362	0.147	

Supplementary Table 14

Estimated effect sizes and heterogeneity estimates for meta-analytic (intercept-only) models, and meta-regression models with subject (STEM or Non-STEM) as moderators, for PISA test scores. $\ln RR$ is a measure of mean difference between test scores for girls and boys. $\ln VR$ is a measure of the difference in variability between girls and boys. $\ln SD$ is the total variability for each Jurisdiction. I^2_{Total} represents proportion of variance not attributed to standard error. $I^2_{\text{Jurisdiction}}$ represents proportion of variance attributed to the Jurisdiction where students were tested. $I^2_{\text{comp_ID}}$ represents residuals against sampling error.

Measure	Data	Fixed effects			Random effects			Heterogeneity		
		Mean	CI.lb	CI.ub		Sigma ²	N levels	I^2_{Total}	$I^2_{\text{Jurisdiction}}$	$I^2_{\text{comp_ID}}$
$\ln RR$	Intercept	0.020	0.014	0.026	Jurisdiction	0.000	64	99.4	1.7	97.7
	Subject: STEM	0.066	0.058	0.075	comp_ID	0.002	192			
	Subject: Non-STEM	-0.003	-0.009	0.002						
	non-STEM - STEM difference	-0.070	-0.074	-0.066						
$\ln VR$	Intercept	-0.066	-0.075	-0.057	Jurisdiction	0.001	64	79.3	75.3	4.0
	Subject: STEM	-0.065	-0.075	-0.054	comp_ID	0	192			
	Subject: Non-STEM	-0.067	-0.076	-0.058						
	non-STEM - STEM difference	-0.002	-0.008	0.003						
$\ln SD$ - overall	non-STEM - STEM difference	4.526	4.499	4.554	Jurisdiction	0.009	64	98.9	73.9	25.0
	girls-boys difference	0.068	0.057	0.078	comp_ID	0.003	192			

<i>lnSD - girls</i>	non- STEM - STEM difference	-0.047	-0.062	-0.032
<i>lnSD - boys</i>	non- STEM - STEM difference	-0.044	-0.058	-0.030

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Supplementary Table 15

Estimated effect sizes for univariate meta-regression models with subject (classified as in Voyer & Voyer 2014) as a moderator for mean grade difference between girls and boys, using $\ln RR$ as the effect size. Positive $\ln RR$ values can be interpreted as girls having higher grades than boys. Effects with confidence intervals (CI) not crossing zero are indicated in bold. k – number of effect sizes included.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	k
$\ln RR$	School subset				215
	Subject: Global	0.098	0.084	0.113	
	Subject: Language	0.091	0.059	0.123	
	Subject: Math	0.033	0.013	0.053	
	Subject: Science	0.024	-0.006	0.055	
	Subject: SocialSci	0.037	0.016	0.058	
	Subject: Other/NR	0.114	0.009	0.218	
$\ln RR$	University subset				131
	Subject: Global	0.055	0.043	0.067	
	Subject: Language	0.050	0.026	0.075	
	Subject: Math	0.039	-0.004	0.082	
	Subject: Science	0.009	-0.008	0.025	
	Subject: SocialSci	0.012	-0.034	0.057	
	Subject: Other/NR	-0.001	-0.025	0.022	
$\ln RR$	All data				346
	Subject: Global	0.076	0.066	0.086	
	Subject: Language	0.084	0.056	0.112	
	Subject: Math	0.029	0.011	0.048	
	Subject: Science	0.017	-0.003	0.038	
	Subject: SocialSci	0.024	0.004	0.044	

Subject: Other/NR 0.031 -0.015 0.078

Supplementary Table 16

Estimated effect sizes for univariate meta-regression models with subject as a moderator for mean grade difference between girls and boys, using *SMD* as the effect size (for comparison with Voyer & Voyer 2017). Positive estimates can be interpreted as girls having higher grades than boys. Effects with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	<i>k</i>
<i>SMD</i>	School subset				215
	Subject: Global	0.438	0.366	0.510	
	Subject: Language	0.415	0.314	0.515	
	Subject: Math	0.132	0.067	0.197	
	Subject: Science	0.120	0.012	0.228	
	Subject: SocialSci	0.201	0.126	0.276	
	Subject: Other/NR	0.496	0.192	0.801	
<i>SMD</i>	University subset				131
	Subject: Global	0.266	0.208	0.324	
	Subject: Language	0.160	-0.041	0.362	
	Subject: Math	0.160	-0.032	0.351	
	Subject: Science	0.041	-0.053	0.134	
	Subject: SocialSci	0.015	-0.156	0.186	
	Subject: Other/NR	0.028	-0.169	0.225	
<i>SMD</i>	All data				346
	Subject: Global	0.349	0.302	0.396	
	Subject: Language	0.381	0.287	0.474	
	Subject: Math	0.124	0.060	0.188	
	Subject: Science	0.080	0.002	0.159	
	Subject: SocialSci	0.140	0.062	0.219	
	Subject: Other/NR	0.166	-0.058	0.390	

Supplementary Table 17

Estimated effect sizes for univariate meta-regression models with subject (classified as in Voyer & Voyer 2014) as a moderator for grade consistency of girls and boys (*lnCVR*). Negative *lnCVR* values can be interpreted as girls having less variable grades than boys. Effects with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	<i>k</i>
<i>lnCVR</i>	School subset				215
	Subject: Global	-0.153	-0.180	-0.126	
	Subject: Language	-0.170	-0.226	-0.115	
	Subject: Math	-0.093	-0.134	-0.053	
	Subject: Science	-0.042	-0.091	0.007	
	Subject: SocialSci	-0.089	-0.150	-0.029	
	Subject: Other/NR	-0.339	-0.541	-0.137	
<i>lnCVR</i>	University subset				131
	Subject: Global	-0.114	-0.148	-0.079	
	Subject: Language	0.068	-0.504	0.640	
	Subject: Math	-0.076	-0.205	0.053	
	Subject: Science	-0.068	-0.151	0.015	
	Subject: SocialSci	0.020	-0.209	0.250	
	Subject: Other/NR	-0.007	-0.116	0.102	
<i>lnCVR</i>	All data				346
	Subject: Global	-0.132	-0.154	-0.110	
	Subject: Language	-0.149	-0.228	-0.071	
	Subject: Math	-0.090	-0.130	-0.049	
	Subject: Science	-0.052	-0.099	-0.006	
	Subject: SocialSci	-0.058	-0.138	0.023	
	Subject: Other/NR	-0.089	-0.214	0.037	

Supplementary Table 18

Estimated effect sizes for univariate meta-regression models with subject (classified as in Voyer & Voyer 2014) and sex as moderators (with interaction) for variation in grades of girls and boys (*lnCV*). Intercept values are presented for each sex-subject combination separately alongside girls-boys contrast for each subject. Effects with confidence intervals (CI) not crossing zero are indicated in bold. *k* – number of effect sizes included in the analysis.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	<i>k</i>
<i>lnCV</i>	School subset:				430
	girls intercepts				
	Subject: Global	-1.555	-1.651	-1.458	
	Subject: Language	-1.466	-1.585	-1.346	
	Subject: Math	-1.352	-1.453	-1.250	
	Subject: Science	-1.433	-1.556	-1.310	
	Subject: SocialSci	-1.490	-1.660	-1.320	
	Subject: Other/NR	-1.561	-2.223	-0.899	
	boys intercepts				
	Subject: Global	-1.402	-1.500	-1.304	
	Subject: Language	-1.290	-1.428	-1.153	
	Subject: Math	-1.257	-1.361	-1.153	
	Subject: Science	-1.390	-1.512	-1.268	
	Subject: SocialSci	-1.406	-1.558	-1.254	
	Subject: Other/NR	-1.305	-1.961	-0.648	
	girls-boys difference				
	Subject: Global	0.153	0.125	0.181	
	Subject: Language	0.175	0.112	0.239	

<i>lnCV</i>	University subset:	Subject: Math	0.095	0.052	0.137
		Subject: Science	0.044	-0.013	0.101
		Subject: SocialSci	0.084	0.011	0.157
		Subject: Other/NR	0.257	0.127	0.386
	girls intercepts				
		Subject: Global	-1.615	-1.725	-1.504
		Subject: Language	-1.375	-2.012	-0.739
		Subject: Math	-1.421	-1.757	-1.085
		Subject: Science	-1.581	-1.849	-1.313
		Subject: SocialSci	-1.382	-1.883	-0.881
		Subject: Other/NR	-2.071	-2.587	-1.555
	boys intercepts				
		Subject: Global	-1.501	-1.608	-1.394
		Subject: Language	-1.447	-1.641	-1.253
		Subject: Math	-1.347	-1.603	-1.090
		Subject: Science	-1.514	-1.805	-1.222
		Subject: SocialSci	-1.402	-1.692	-1.113
		Subject: Other/NR	-2.063	-2.638	-1.489
	girls-boys difference				
		Subject: Global	0.114	0.078	0.150
		Subject: Language	-0.071	-0.663	0.521
		Subject: Math	0.074	-0.058	0.207
		Subject: Science	0.067	-0.017	0.152
		Subject: SocialSci	-0.020	-0.254	0.214
		Subject: Other/NR	0.008	-0.106	0.121
<i>lnCV</i>	All data				
	girls intercepts				
		Subject: Global	-1.588	-1.659	-1.517

	Subject: Language	-1.498	-1.616	-1.380
	Subject: Math	-1.397	-1.495	-1.299
	Subject: Science	-1.500	-1.610	-1.391
	Subject: SocialSci	-1.522	-1.684	-1.359
	Subject: Other/NR	-1.691	-2.208	-1.175
boys intercepts				
	Subject: Global	-1.456	-1.526	-1.385
	Subject: Language	-1.346	-1.469	-1.222
	Subject: Math	-1.306	-1.403	-1.209
	Subject: Science	-1.449	-1.566	-1.331
	Subject: SocialSci	-1.466	-1.599	-1.332
	Subject: Other/NR	-1.607	-2.187	-1.027
girls-boys difference				
	Subject: Global	0.132	0.110	0.155
	Subject: Language	0.153	0.070	0.235
	Subject: Math	0.091	0.049	0.133
	Subject: Science	0.052	0.005	0.099
	Subject: SocialSci	0.056	-0.029	0.140
	Subject: Other/NR	0.084	-0.036	0.203

Supplementary Table 19

Heterogeneity estimates for meta-analytic (intercept-only) models for overall grade consistency of girls and boys (*lnCVR*) and for overall mean grade difference between girls and boys (*lnRR* and *SMD*). Analyses were run on all included data and on two data subsets consisting of school pupils and university students, respectively. I^2_{Total} represents proportion of variance not attributed to standard error. $I^2_{\text{cohort_ID}}$ represents proportion of variance attributed to cohort identity. $I^2_{\text{comp_ID}}$ represents residuals against sampling error.

Measure	Data	Heterogeneity		
		I^2_{Total}	$I^2_{\text{cohort_ID}}$	$I^2_{\text{comp_ID}}$
	School subset			
lnCVR		98.4%	10.1%	88.3%
lnRR		99.0%	9.8%	89.2%
SMD		60.0%	9.0%	51.1%
	University subset			
lnCVR		99.5%	0.0%	99.5%
lnRR		99.2%	97.2%	2.1%
SMD		60.2%	57.5%	2.7%
	All data			
lnCVR		99.2%	4.0%	95.2%
lnRR		99.3%	11.0%	88.3%
SMD		61.3%	10.4%	51.0%

519 **Supplementary Table 20**

520 Estimated moderator effects from multivariate meta-regression models with study
 521 year, age and with subject (classified as in Voyer & Voyer 2014) as moderators for
 522 grade consistency of girls and boys (*lnCVR*). Negative *lnCVR* intercept values for
 523 different subjects can be interpreted as girls having less variable grades than boys.
 524 “Study year” and “Age” variables are scaled (Z-transformed). Effects with confidence
 525 intervals (CI) not crossing zero are indicated in bold. “N missing” indicates how
 526 many missing values of a moderator had to be imputed.

Measure	Data		Fixed effects			
			Mean	CI.lb	CI.ub	N missing
<i>lnCVR</i>	School subset					
		Study year (slope)	-0.035	-0.087	0.018	121
		Age (slope)	0.033	0.000	0.067	37
		Subject: Global	-0.151	-0.184	-0.118	
		Subject: Language	-0.184	-0.247	-0.122	
		Subject: Maths	-0.113	-0.158	-0.067	
		Subject: Other/NR	-0.368	-0.765	0.029	
		Subject: Science	-0.048	-0.099	0.003	
		Subject: SocialSci	-0.065	-0.197	0.067	
<i>lnCVR</i>	University subset					
		Study year (slope)	-0.025	-0.072	0.023	52
		Age (slope)	-0.013	-0.089	0.062	91
		Subject: Global	-0.121	-0.155	-0.087	
		Subject: Language	-0.086	-0.541	0.369	
		Subject: Maths	-0.074	-0.196	0.049	
		Subject: Other/NR	-0.007	-0.129	0.115	
		Subject: Science	-0.058	-0.142	0.026	
		Subject: SocialSci	0.012	-0.197	0.220	
<i>lnCVR</i>	All data					

Study year (slope)	-0.019	-0.050	0.011	173
Age (slope)	0.023	-0.004	0.051	128
Subject: Global	-0.149	-0.173	-0.125	
Subject: Language	-0.159	-0.226	-0.093	
Subject: Maths	-0.107	-0.151	-0.063	
Subject: Other/NR	-0.136	-0.276	0.004	
Subject: Science	-0.059	-0.110	-0.009	
Subject: SocialSci	-0.058	-0.165	0.049	

Supplementary Table 21

Estimated moderator effects from multivariate meta-regression models with study year, age and with subject (classified as in Voyer & Voyer 2014) as moderators for mean grade difference between girls and boys (*lnRR*). Positive *lnRR* intercept values for different subjects can be interpreted as girls having higher grades than boys. “Study year” and “Age” variables are scaled (Z-transformed). Effects with confidence intervals (CI) not crossing zero are indicated in bold. “N missing” indicates how many missing values of a moderator had to be imputed.

Measure	Data		Fixed effects			
			Mean	CI.lb	CI.ub	N missing
<i>lnRR</i>	School subset					
		Study year (slope)	0.019	-0.009	0.046	121
		Age (slope)	0.001	-0.013	0.016	37
		Subject: Global	0.102	0.086	0.118	
		Subject: Language	0.102	0.065	0.139	
		Subject: Maths	0.050	0.024	0.075	
		Subject: Other/NR	0.187	-0.220	0.594	
		Subject: Science	0.032	0.008	0.055	
		Subject: SocialSci	0.055	0.025	0.084	
<i>lnRR</i>	University subset					
		Study year (slope)	-0.011	-0.033	0.010	52
		Age (slope)	-0.011	-0.041	0.020	91
		Subject: Global	0.066	0.051	0.081	
		Subject: Language	0.026	-0.018	0.071	
		Subject: Maths	0.026	-0.051	0.102	
		Subject: Other/NR	0.007	-0.031	0.046	
		Subject: Science	0.014	-0.019	0.048	
		Subject: SocialSci	-0.026	-0.078	0.027	
<i>lnRR</i>	All data					
		Study year (slope)	0.004	-0.015	0.023	173

Age (slope)	-0.017	-0.028	-0.005	128
Subject: Global	0.093	0.080	0.106	
Subject: Language	0.090	0.058	0.122	
Subject: Maths	0.043	0.018	0.068	
Subject: Other/NR	0.086	-0.050	0.223	
Subject: Science	0.029	0.007	0.052	
Subject: SocialSci	0.030	-0.001	0.061	

537 **Supplementary Table 22**

538 Estimated moderator effects from multivariate meta-regression models with study
 539 year, age and with subject (classified as in Voyer & Voyer 2014) as moderators for
 540 mean grade difference between girls and boys (*SMD*). Positive *SMD* intercept values
 541 for different subjects can be interpreted as girls having higher grades than boys.
 542 “Study year” and “Age” variables are scaled (Z-transformed). Effects with confidence
 543 intervals (CI) not crossing zero are indicated in bold. “N missing” indicates how
 544 many missing values of a moderator had to be imputed. Results are not shown for the
 545 university subset, as this model did not converge.

Measure	Data	Fixed effects			
		Mean	CI.lb	CI.ub	N missing
<i>SMD</i>	School subset				
	Study year (slope)	0.051	-0.024	0.127	121
	Age (slope)	0.032	-0.034	0.098	37
	Subject: Global	0.449	0.361	0.538	
	Subject: Language	0.425	0.321	0.529	
	Subject: Maths	0.160	0.088	0.232	
	Subject: Other/NR	0.548	-0.324	1.421	
	Subject: Science	0.119	0.005	0.233	
	Subject: SocialSci	0.211	0.084	0.337	
<i>SMD</i>	University subset				
	Study year (slope)	0.011	-0.056	0.078	52
	Age (slope)	0.010	-0.089	0.109	91
	Subject: Global	0.293	0.222	0.364	
	Subject: Language	0.213	-0.006	0.433	
	Subject: Maths	0.180	-0.054	0.414	
	Subject: Other/NR	0.008	-0.228	0.245	
	Subject: Science	0.028	-0.140	0.196	

<i>SMD</i>	All data	Subject: SocialSci	0.061	-0.209	0.331	
		Study year (slope)	0.029	-0.012	0.070	173
		Age (slope)	-0.034	-0.079	0.010	128
		Subject: Global	0.395	0.333	0.457	
		Subject: Language	0.394	0.296	0.492	
		Subject: Maths	0.140	0.064	0.215	
		Subject: Other/NR	0.284	-0.044	0.612	
		Subject: Science	0.091	-0.001	0.183	
		Subject: SocialSci	0.158	0.045	0.271	

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Supplementary Table 23

Estimated moderator effects from Bayesian multivariate meta-regression models with study year, age and subject (STEM, non-STEM, etc.) as moderators for grade consistency of girls and boys (*lnCVR*). Negative *lnCVR* intercept values for different subjects can be interpreted as girls having less variable grades than boys. “Study year” and “Age” variables are scaled (Z-transformed). Effects with Highest Posterior Density intervals (HPD) not crossing zero are indicated in bold.

Measure	Data		Fixed effects				
			Mode	Mean	SD	HPD.lb	HPD.ub
<i>lnCVR</i>	School subset	Study year (slope)	-0.033	-0.03	0.019	-0.064	0.007
		Age (slope)	-0.005	-	0.012	-0.057	0.039
		Subject: STEM	-0.023	-	0.007	-0.161	0.108
		Subject: Non-STEM	0.109	0.086	0.098	-0.089	0.289
		Subject: Global	-0.121	-	0.021	-0.156	-0.076
		Subject: Other/NR	-0.059	-0.07	0.059	-0.182	0.041
<i>lnCVR</i>	University subset	Study year (slope)	-0.023	-0.03	0.018	-0.065	0.008
		Age (slope)	-0.016	-	0.013	-0.061	0.034
		Subject: STEM	0.036	-	0.005	-0.139	0.133
		Subject: Non-STEM	0.084	0.079	0.097	-0.102	0.278
		Subject: Global	-0.113	-	0.020	-0.154	-0.077
		Subject: Other/NR	-0.046	-	0.061	-0.184	0.051
Measure	Data		Random effects				
			Mode	Mean	SD	HPD.lb	HPD.ub
<i>lnCVR</i>	School subset	cohort_ID	0.000	0.002	0.002	0.000	0.007
		residuals	0.029	0.031	0.005	0.021	0.041

<i>lnCVR</i>	University subset						
		cohort_ID	0.000	0.002	0.002	0.000	0.007
		residuals	0.028	0.032	0.005	0.022	0.041
Measure	Data	Heterogeneity			DIC		
		Mode	Mean	SD	Mean	Range	
<i>lnCVR</i>	School subset						
		$I^2_{\text{cohort_ID}}$	0.1	5.3	6.6	-55.08	(-55.19, -54.99)
		$I^2_{\text{comp_ID}}$	99.4	94.2	6.6		
		I^2_{total}	99.6	99.5	0.1		
<i>lnCVR</i>	University subset						
		$I^2_{\text{cohort_ID}}$	0.1	5.2	6.5	-55.08	(-55.19, -54.99)
		$I^2_{\text{comp_ID}}$	99.4	94.3	6.5		
		I^2_{total}	99.6	99.5	0.1		

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Supplementary Table 24

Estimated moderator effects from Bayesian multivariate meta-regression models with study year, age and subject (STEM, non-STEM, etc.) as moderators for mean grade difference between girls and boys ($\ln RR$). Positive $\ln RR$ intercept values for different subjects can be interpreted as girls having higher grades than boys. “Study year” and “Age” variables are scaled (Z-transformed). Effects with Highest Posterior Density intervals (HPD) not crossing zero are indicated in bold.

Measure	Data		Fixed effects				
			Mode	Mean	SD	HPD.lb	HPD.ub
$\ln RR$	School subset	Study year (slope)	-0.006	-0.009	0.006	-0.021	0.001
		Age (slope)	-0.012	-0.012	0.007	-0.025	0.001
		Subject: STEM	0.006	0.004	0.020	-0.035	0.042
		Subject: Non-STEM	0.037	0.042	0.029	-0.015	0.096
		Subject: Global	0.057	0.056	0.006	0.044	0.068
		Subject: Other/NR	0.033	0.041	0.018	0.006	0.077
$\ln RR$	University subset	Study year (slope)	-0.006	-0.009	0.006	-0.021	0.001
		Age (slope)	-0.012	-0.012	0.007	-0.025	0.001
		Subject: STEM	0.006	0.004	0.020	-0.035	0.042
		Subject: Non-STEM	0.037	0.042	0.029	-0.015	0.096
		Subject: Global	0.057	0.056	0.006	0.044	0.068
		Subject: Other/NR	0.033	0.041	0.018	0.006	0.077
Measure	Data		Random effects				
			Mode	Mean	SD	HPD.lb	HPD.ub
$\ln RR$	School subset	cohort_ID	0.000	0.001	0.001	0.000	0.003
		residuals	0.002	0.002	0.001	0.000	0.003

<i>lnRR</i>	University subset						
		cohort_ID	0.000	0.001	0.001	0.000	0.003
		residuals	0.002	0.002	0.001	0.000	0.003

Measure	Data	Heterogeneity			DIC	
		Mode	Mean	SD	Mean	Range
<i>lnRR</i>	School subset					
		$I^2_{\text{cohort_ID}}$	0.5	41.1	27.1	- (-445.66, -
		$I^2_{\text{comp_ID}}$	98.7	58.0	27.1	443.75 442.72)
		I^2_{total}	99.2	99.1	0.2	
<i>lnRR</i>	University subset					
		$I^2_{\text{cohort_ID}}$	0.5	41.1	27.1	- (-445.66, -
		$I^2_{\text{comp_ID}}$	98.7	58.0	27.1	443.75 442.72)
		I^2_{total}	99.2	99.1	0.2	

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Supplementary Table 25

Intercept estimates from Bayesian Egger’s regressions, performed on the full meta-regression model residuals and measurement errors from the school and university data subsets. Intercepts with Highest Posterior Density intervals (HPD) not crossing zero (in bold) indicate publication bias in the data.

Test	Effect type	Data	Mean Intercept	HPD.lb	HPD.ub
Egger's regression	lnCVR				
		School subset	-0.199	-0.664	0.275
		University subset	0.398	-1.173	2.169
Egger's regression	lnRR				
		School subset	-0.130	-0.693	0.438
		University subset	0.860	0.126	1.520
Egger's regression	SMD				
		School subset	0.029	-0.474	0.538
		University subset	0.294	-0.194	0.832

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