On average, students attending selective schools outperform their non-selective counterparts in national exams. These differences are often attributed to value added by the school, as well as factors schools use to select pupils, including ability, achievement and, in cases where schools charge tuition fees or are located in affluent areas, socioeconomic status. However, the possible role of DNA differences between students of different school types has not yet been considered. We used a UK-representative sample of 4814 genotyped students to investigate exam performance at age 16 and genetic differences between students in three school types: state-funded, non-selective schools (‘non-selective’), state-funded, selective schools (‘grammar’) and private schools, which are selective (‘private’). We created a genome-wide polygenic score (GPS) derived from a genome-wide association study of years of education (EduYears). We found substantial mean genetic differences between students of different school types: students in non-selective schools had lower EduYears GPS compared to those in grammar (d = 0.41) and private schools (d = 0.37). Three times as many students in the top EduYears GPS decile went to a selective school compared to the bottom decile. These results were mirrored in the exam differences between school types. However, once we controlled for factors involved in pupil selection, there were no significant genetic differences between school types, and the variance in exam scores at age 16 explained by school type dropped from 7% to <1%. These results show that genetic and exam differences between school types are primarily due to the heritable characteristics involved in pupil admission.

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INTRODUCTION
Achievement at the end of full-time compulsory education represents a major tipping point in life, opening up avenues for higher education, including university and beyond. Therefore, understanding the potential predictors of academic achievement at this juncture is of great importance. One such predictor that has been hotly debated is school type. In England, when students transition from primary to secondary school at age 11, they have the option of attending one of three school types. Ninety-three percent of children start state-funded schools, the majority of which are non-selective (state non-selective). A small proportion of state-funded schools (163 schools out of 3113 schools in England) are academically selective ‘grammar’ schools. These schools select their intake based on achievement and ability, assessed by an entrance exam. The remainder of students (approximately 7%), are private educated. As well as being fee-paying, private schools are often also academically selective. These school types are assumed to set children on different trajectories, with research linking selective schools (grammar and private schools) to later success, including higher levels of academic achievement, acceptance at university, and even higher earning potential compared to pupils educated in non-selective schools.2-4

However, by design, selective schools are able to choose their student intake based on certain pupil characteristics. This can include selection on ability or achievement on an entrance test; both of which have been shown to correlate positively with life outcomes, including later academic achievement.5,6 Furthermore, by virtue of being fee-paying, entrance into private schools is usually dependent on whether the family can afford it (their socioeconomic status (SES)), which also correlates with future outcomes.7-10 Even for state schools, family SES may play a role in what school type a student attends, with grammar schools typically located in more affluent areas and attracting higher SES students on average.11 It is, therefore, possible that improved outcomes for pupils in selective schools do not necessarily reflect a higher quality of education, but may simply be the consequence of selection—either active, as in the case of ability or achievement, or passive, as in the case of family SES.

Given the considerable fees charged by private schools, in addition to the potential stress of selective school entrance exams,
why do families choose these schools? Among the many possible reasons is superior academic achievement. The finding that pupils at selective schools outperform their non-selective school counterparts in exams has been frequently reported. At age 16, students in the UK typically take the General Certificate of Secondary Education (GCSE) exams. The UK Department for Education shows that 99% of grammar school students obtain top GCSE grades (A*-C grade) in English and mathematics, compared to 64% for all state-funded mainstream school students. However, academic achievement at age 16 is positively correlated with the factors involved in pupil selection, such as prior achievement, ability and SES. Therefore, this raises the question—are selective schools adding anything over and above these factors in the prediction of academic achievement?

Several studies have attempted to elucidate the effect of school type on achievement over and above factors on which schools can select (for example, review, see Coe et al.). However, many of these have not been published in peer-reviewed journals—for example—and we are not aware of a recent peer-reviewed study looking at all three school types: state non-selective, grammar and private schools in the UK. However, the non-peer-reviewed reports support the conclusion that there are only small academic advantages to attending a selective school, after student factors such as achievement, ability and family SES have been taken into account.

Traditionally, the relationship between the factors involved in school admission and later achievement have been thought to operate environmentally. For example, parents with higher SES may invest more time in their children’s education and can afford more resources (e.g., more books or private tuition), which in turn may lead to better opportunities and improved achievement. However, a less frequently investigated factor influencing both selection factors, as well as achievement, is genetics. In the example above, parents with higher SES are not only passing on educationally relevant environments, but they are also passing on educationally relevant genes, a concept referred to as gene-environment correlation (rGE).

A vast literature from quantitative genetics has shown that genetic factors explain a substantial amount of variance in selection factors, including ability and achievement. Heritability estimates of general cognitive ability (g) from twin studies range from around 30% in childhood, to 40–50% in adolescence and approximately 60% in adulthood. Twin studies also show that much of the relationship between selection factors, such as g, and later achievement, are substantially influenced by genetics.

Because twins typically grow up in the same family, the aetiology of traits such as family SES, which do not vary between twins, cannot be estimated in this way. However, heritability can be estimated by genome-wide complex trait analysis (GCTA), which uses DNA from unrelated individuals to estimate the proportion of phenotypic variance explained by hundreds of thousands of single-nucleotide polymorphisms (SNPs) genotyped on DNA arrays. This method has also shown that genetics accounts for a significant amount of individual differences in family SES, as well as g and achievement.

School type, like SES, does not tend to vary within twin pairs. However, because GCTA requires large sample sizes, it has so far not been possible to look at the genetic differences between students of different school types. However, powerful genome-wide association (GWA) studies of behavioural traits, which test associations between specific SNPs and traits are starting to make this possible. Although individually these SNPs, identified through GWA studies, are of small effect, by summing their effects together it is possible to create a genetic score for each individual in an independent sample, which explains a substantial proportion of the genetic variation. These scores, dubbed ‘genomewide polygenic scores’ (GPS) are a game-changer for genetic research and have already proved insightful within the area of educational achievement. For example, a recent study using a GPS derived from a 2016 GWA study of years of education (EduYears) has shown educational achievement scores at age 16 differ as a function of GPS. There was approximately one standard deviation difference between those in the highest GPS septile and those in the lowest; representing almost a whole school grade difference. Furthermore, while 65% of students in the highest GPS septile went on to university, only 37% in the lowest septile progressed to university-level education.

For the first time, we assess differences in a polygenic score for years of education (EduYears) between students from three school types: non-selective, grammar and private schools. We predict that selection involving heritable traits such as achievement, ability and family SES will be reflected in the genetic differences between students of different school types. Furthermore, in line with previous literature, we expect that selection will also create large achievement differences between students attending the three school types, which will reduce substantially once controlling for the selection factors.

RESULTS

Polygenic score differences between school types

Students attending different school types (state non-selective, grammar and private schools) differed genetically, as shown by their mean EduYears GPS (see Fig. 1, analysis of variance (ANOVA) details in Table S1). Non-selective state school students had significantly lower EduYears GPS scores compared to grammar school students (t = 4.87, p < 0.0001) and private school students (t = 7.17, p < 0.0001). These differences translate to more than a third of a standard deviation difference (d = 0.41 and 0.37, respectively).

There were no significant mean differences in EduYears GPS scores between grammar and private school students (t = 0.44, p = 0.66). There were also no significant mean differences between state non-selective schools in varying selectivity areas (see Table S2 and Supplementary Fig. S1).

Associations between EduYears GPS and selection factors

EduYears GPS was positively correlated with each of the selection factors (see Supplementary Table S3), explaining 2.1% of the variance in ability, 5.2% in achievement and 6.6% in family SES. EduYears GPS was also positively correlated with GCSE, explaining 7.6% of the variance in GCSE scores, similar to previous analysis of these data. Because selective schools actively select for achievement and ability and passively select for SES, all of which correlate with EduYears GPS, we tested whether mean differences in EduYears GPS remained once controlling for these factors.

We found that, after accounting for the variance explained by heritable selection factors, there were no significant EduYears GPS differences between students of the three school types: state non-selective, grammar and private (see Supplementary Fig. S2 and Supplementary Table S4). Similar results also emerged when we looked at differences between state non-selective schools in varying selectivity areas (see Supplementary Table S5 and Supplementary Fig. S3), showing small differences in EduYears between school types.

GCSE differences

Supplementary Table S6 and Fig. 2 show unadjusted average GCSE grades for state non-selective, grammar and private school students, as well as average GCSE score adjusting separately for EduYears GPS, family SES, prior ability and prior achievement, and for all variables together. Unadjusted GCSEs between school types mirrored unadjusted EduYears GPS results, with large differences between non-selective and selective schools (see ‘Unadjusted GCSE’ in Fig. 2, details in Supplementary Table S6). Indeed, the
mean GCSE score of students attending state non-selective schools was approximately 1 SD below the mean GCSE score of those attending grammar schools ($d = 1.05, 95\% \text{ CIs} = 0.83–1.28$) and private school students ($d = 0.92, 95\% \text{ CIs} = 0.75–1.09$). This translates to a whole grade difference between average GCSE scores for state non-selective school students and selective school students. There was no difference between grammar and private school students ($t = 1.00, p = 0.32$). There were also no significant differences between non-selective schools in areas that varied in the selectivity of their schools (see Supplementary Table S7 and Supplementary Fig. S4).

Controlling for selection factors

Controlling for EduYears GPS had a small effect on average GCSE grades, with the GCSE variance explained by school type dropping slightly from $R^2 = 0.07$ to 0.06, see Fig. 2, details in Supplementary Table S6). This relatively small effect is to be expected given that EduYears GPS accounts for only 8\% of the variance in GCSE (see Supplementary Table S3). Controlling for family SES and prior ability had a slightly larger effect on GCSE, in line with the GCSE variance they account for ($R^2 = 24\%$ and 27\%, respectively). Out of all of the selection factors, prior achievement had the biggest impact on GCSE grades between school type, with average GCSE for grammar schools falling from 10.12 (grade A) to 9.21 (grade B). After controlling for prior achievement, the variance in GCSE explained by school type dropped from 7.1 to 1.3\%.

Controlling for all of the selection factors and EduYears GPS together saw a further reduction in average GCSE between school types, with average GCSE score for grammar ($M = 9.14; t = 2.35, p < 0.019$) and private ($M = 9.32, t = 6.16, p < 0.001$) similar to that of state non-selective school students’ average grade ($M = 8.96$). Although these mean differences between school types remained significant, they were greatly reduced. Standardised betas...
indicated that attending a grammar school compared to a non-selective state school was associated with an increase of just 0.03 of a standard deviation in GCSE, and for private schools, the increase was 0.07. In addition, no significant differences emerged between non-selective schools in varying selectivity areas (see Supplementary Table S7 and Supplementary Fig. S4).

One of our main findings was that after accounting for the variance explained by the selection factors and EduYears GPS, the variance in GCSE explained by school type dropped from 7.1% to only 0.5% (see Supplementary Table S6 for regression results).

**DISCUSSION**

We report genetic mean differences between students attending three different types of school: state non-selective, grammar and private schools. We find that, on average, students in state non-selective schools have lower polygenic scores for years of education (EduYears) compared to their peers in selective schools. Furthermore, following the same pattern of results as EduYears, there are also substantial mean differences in GCSE performance between pupils in selective and non-selective school types. However, almost all of these differences are explained by heritable, individual-level factors, which schools actively or passively use in the pupil selection process.

Although finding genetic differences between state non-selective, grammar and private school students may initially seem surprising, when we consider the heritable traits that selection is based on, this difference is less unexpected. Put another way, students with higher polygenic score for years of education have, on average, higher cognitive ability, better grades and come from families with higher SES, and these students are subsequently more likely to be accepted into selective schools. This results in a system in which children are intentionally phenotypically selected, but unintentionally genetically selected.

However, despite finding mean genetic differences between students of different school types, it should be noted that the majority of the variation in EduYears GPS occurs within the school type, not between the school types. For example, a Cohen's d of 0.41, (the difference between mean EduYears scores for state non-selective school students and grammar school students), which is classed as a small-medium effect size, translates to an overlap of approximately 83% between the two distributions.40

Nevertheless, finding an association between genotype and school type suggests that genetic factors are contributing to variation in educational environments, a concept known as gene-environment correlation (rGE). This occurs when individuals select, modify and ‘inherit’ their environment, in part based on their genotype.70,41 Putting our research within the context of rGE, we suggest that in addition to students being selected into schools based on their genetically influenced traits (evocative rGE), children themselves also actively select educational environments that correlate with their genotype (active rGE). In the case of high achieving students, these environments might be challenging or competitive academic institutions, which grammar and private schools are often reputed to be. Finally, because we know that the factors used in school selection are substantially heritable, it is likely that academically gifted children will come from academically gifted parents. These parents not only provide the genes but also the environments to help them progress academically.

As well as having a higher average EduYears polygenic score, students attending selective schools also achieve better GCSE results on average.2,3,12–14,17 There has been some debate in the literature as to the size of this achievement gap, with studies accounting for different background characteristics in their analysis. We find that almost all of the selective school advantage in GCSE can be explained by family SES, achievement, ability and EduYears GPS. After controlling for these factors, going to a grammar vs. a state non-selective school is associated with a mean GCSE grade increase of just 0.026 of a standard deviation and for private schools, 0.070 of a standard deviation. Furthermore, the variance in GCSE that school type explains falls from 7% to <1%

Controlling for EduYears alone had a fairly small effect on average GCSE grades between school types. However, this is to be expected considering that EduYears GPS currently predicts approximately 8% of the variance in GCSE—15% of the heritability estimated by the twin design22 and approximately one-third of the heritable variance from SNP-based studies of GCSE at age 16.30 The predictive nature of EduYears is likely to increase with more powerful GWA studies. For example, there was a threefold increase in prediction of educational achievement at age 16 from the 2016 EduYears GPS (based on a GWA study with N = 293,723) as compared to the 2013 EduYears GPS (N = 136,559).52

Although there were only small mean differences between school types once selection factors and EduYears were controlled for, this does not mean that other factors are not important for achievement at age 16. Altogether, these factors do not predict all of the variance in GCSE ($R^2 = 0.69$). As shown previously, achievement is the result of many genetically influenced traits, including behaviour, personality, home environment and health.22 Furthermore, by finding a small effect of school type, we are not saying schools are unimportant, or that teaching does not work. Without schools, it is hard to imagine a successful education system that allows children to reach their academic potential. However, while schools themselves are important for academic achievement, the type of school appears less so. Educational achievement is not necessarily the only reason parents opt to send their children to selective schools. A recent report on private schools found that these students earned about £200,000 more in their early career (between ages 26 and 42) as compared to state school students.2 However, this report did not distinguish between non-selective and selective state schools. More research is needed to see whether differences in university attendance, career choice and earnings are still predicted by school type once individual student factors have been accounted for. In addition to differences in university and career outcomes, it would also be of interest to identify potential differences between school types in terms of non-cognitive traits as outcomes, with one survey finding 66% of parents believing that private schools ‘instil a sense of confidence in pupils’.2

There are several limitations to our study. First, we recognise that there is considerable variation in schools within our three school types—within each of the school types, there will be examples of exceptional and under-performing schools. In particular, there is more variance in the state non-selective schools category as it includes most of the schools. It also includes a wide variety of other categories, such as schools that are allowed to select for religion and schools that are allowed to select up to 10% of their pupils for talent in specialist subjects, such as sport, performing or visual arts, and languages. These schools are not allowed to select directly on academic grounds. However, there is some evidence that they do in fact select more able students.42 Nonetheless, accounting for prior achievement and ability at age 11, before most children enter secondary school, adjusts for this.

Another limitation of the present study is access to school type. Grammar and private schools are not evenly distributed around the country. Therefore, in local authority areas where there are no selective schools, the average GCSE grade of pupils in non-selective schools may be higher and in areas where there are a greater number of selective schools, the average GCSE grade of non-selective schools may be lower. Because there are far fewer selective schools, this geographical effect may potentially inflate the average non-selective school GCSE grade. To see whether this had an impact on GCSE differences, we split the non-selective school group into three further groups: non-selective schools in selective areas, partially selective areas and non-selective areas. Once we controlled for all of the selection factors, we found that there were
no differences between non-selective schools in areas of varying selectivity (see Supplementary Table S7 and Supplementary Fig. S4).

A final limitation to note is that the GCSE variable we used in the analysis is a composite of only the three core subjects taken at age 16—English, science and mathematics. For other subjects, such as languages, art and social sciences, school type may have a greater influence. However, because different school types prioritise different subjects, it is difficult to untangle the effect of school type on optional rather than core subjects, although this would be a useful direction for future research.

In the current study, we find genetic differences between students attending three school types: state non-selective schools, grammar schools and private schools. We find that selective school students have higher polygenic scores for years of education on average compared to students attending non-selective schools. Furthermore, we find substantial mean differences in GCSE between school types. However, once student and family factors have been accounted for, as well as EduYears GPS, the type of school that a child attends explains less than one percent of the individual differences in educational achievement (GCSE mean grade) at age 16.

METHODS

Sample
This study included unrelated individuals from the Twins Early Development Study (TEDS). TEDS is a large, representative sample of 16,000 twin pairs born in England and Wales between 1994–1996 and followed from birth to the present day. Ethical approval for this study was received from King’s College London Ethics Committee. Although there has been some attrition throughout the years, approximately 10,000 twin pairs are still actively involved in the study and provide rich behavioural and cognitive data. Importantly, TEDS was and still is a representative sample of England and Wales, as described in detail elsewhere. In the present study, we included 4814 unrelated individuals (one twin randomly in a pair) who had data present for the selection factors: ability, achievement and SES, females (54%) and 2217 males (46%). Of this sample, 2533 individuals also included a series of questions asking what type of school they attended.

Measures
School type. When TEDS twins were 18, they received a questionnaire that included a series of questions asking what type of school they attended when they took exams at age 16—the GCSEs. Respondents were asked to indicate either ‘Yes’ or ‘No’ for different school types. We classified all respondents who reported attending either a state non-selective school as ‘State non-selective’, all those who indicated that they went to a grammar school as ‘Grammar’ and all those indicating that they went to a private school as ‘Private’. In addition to TEDS data, we also accessed school type information through the National Pupil Database (NPD, https://www.gov.uk/government/collections/national-pupil-database). By supplementing TEDS data with that from NPD, our final school type numbers were: state non-selective: n = 4263, grammar: n = 143, private: n = 408. We also further split state non-selective schools into three categories for follow-up analysis: non-selective in fully selective areas (selective schools in fully selective areas (state non-selective schools into three categories for follow-up analysis: non-selective schools, grammar schools and private schools. We find that selective school students have higher polygenic scores for years of education on average compared to students attending non-selective schools. Furthermore, we find substantial mean differences in GCSE between school types. However, once student and family factors have been accounted for, as well as EduYears GPS, the type of school that a child attends explains less than one percent of the individual differences in educational achievement (GCSE mean grade) at age 16.

English, mathematics and science. In addition, students are allowed to choose a range of other subjects such as geography, history and art. These subjects were graded from 4 (G, the minimum pass grade) to 11 (A*, the best possible grade). In the current sample, GCSE results were obtained from questionnaires sent via mail, in addition to telephone interviews with twins and their parents. We further supplemented this with data from NPD. Our analyses focused on the three core subjects: English, mathematics and science taken by all students. Students taking science GCSE are either awarded separate GCSEs for physics, chemistry and biology (‘triple science’) or as one course, which is double weighted (‘double science’). Therefore, we browsed the mean grade of the science GCSEs. Because English, mathematics and science grades correlated highly (r = 0.70–0.82), we created a GCSE composite. There were 3920 individuals for whom we had both self-reported GCSE and NPD data, this dataset correlated at r = 0.99 between both data sources, which supported the high accuracy of TEDS data.

Selection factors
Socioeconomic status. Family SES was measured by taking the arithmetic mean of five measures: maternal and paternal education (measured on a scale from 1–8, where 1 = no education and 8 = postgraduate qualifications), occupation (indexed by the Standard Occupational Classification (2000) on a scale from 1–9, where 1 = elementary administration and service occupations and 9 = managers, directors and senior officials) and maternal age at birth of first child. All measures were standardised to have a mean of 0 and a SD of 1 and at least three measures were required to calculate the arithmetic mean.

Achievement tests at age 11. We did not have access to selective school entrance exams, however, before children transition to secondary school, they are usually required to take exams, which include English and mathematics tests. In our sample these tests comprise two English tests (reading and writing) and three maths tests (calculator and non-calculator test as well as a mental arithmetic test). Due to the high correlation between maternal and English scores (r = 0.67), we created a composite of these test scores requiring both to be present.

Ability (general cognitive ability, g). To measure general cognitive ability, participants were asked to complete an online battery of cognitive tests administered as part of TEDS testing at age 11. These tests included verbal and non-verbal abilities (M = 11.2, SD = 0.69). A mean score was derived from four tests, two verbal tests (the Wechsler Intelligence Scale for Children (WISC) Vocabulary Multiple-Choice and the WISC General Knowledge test) and two non-verbal tests (Raven’s Progressive Matrices and the WISC Picture Completion task).

Data availability
For information on data availability, please see the Twins Early Development Study data access policy. This can be found at: http://www.teds.ac.uk/research/collaborators-and-data/teds-data-access-policy.

Analyses
Genome-wide polygenic scores. We calculated polygenic scores that were based on the summary statistics of the largest GWA study for years of education (N = 293,723 individuals). A GPS is calculated by using information from GWA study summary statistics about the strength of association between a genetic variant and a trait, to score individuals’ genotypes in independent samples. For each genotype in the independent sample, all trait-associated alleles are counted and multiplied by their effect size (i.e., their strength of association with a trait as reported in GWA summary statistics). The sum of these weighted and counted alleles forms a polygenic score for each individual. We used the software PRSice to calculate peak GPS. The individual GPS quality was also checked for linkage disequilibrium by applying an R² = 0.1 cutoff within a 250-kb window. It is possible to calculate various GPS based on different GWA study significance thresholds for genetic variants, with less stringent p-value thresholds resulting in GPS that include more SNPs. Here, we calculated GPS for seven p-value thresholds (0.001, 0.05, 0.1, 0.2, 0.3, 0.4, 0.5). We report analyses for the p-value threshold of 0.05 in the main text; however, the analyses for the other p-value thresholds are reported in Supplementary Fig. S5. We regressed all GPS on the first ten principal components and used these standardised residuals in our analyses to account for population stratification.
Mean differences. To estimate differences between the three school types: state non-selective, grammar and private schools, we used a one-wayANOVAwith planned contrasts. In addition to the three-level school type analysis, we also conducted follow-up analysis looking at differences between state non-selective schools in areas with and without grammar schools: non-selective schools in fully selective areas, non-selective schools in partially selective areas and non-selective schools in non-selective areas. As the sample sizes varied between groups, we used adjusted Cohen’s d to estimate effect size. This test adjusts the calculation of the pooled standard deviation with weights for the sample sizes.

To test the effect of school type after controlling for selection factors (SES, prior achievement and prior ability) and EduYears GPS, we conducted hierarchical linear regression with dummy coding. See Supplementary Methods S3 for further information on analysis.

All methods were performed in accordance with relevant regulations and guidelines.

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AUTHOR CONTRIBUTIONS

R.P. directs and received funding for the Twins Early Development Study (TEDS). R.P. and E.S.W. conceived the present study. E.S.W. analysed and interpreted the data with advice from all co-authors. R.P. supervised the project and interpreted the data. R.P. and E.S.W. wrote the manuscript with help from all authors (J.B.P., S.S., K.R., E.K., S.v.S., K.A., P.S.D., T.Y., R.A., and Y.K.).

ADDITIONAL INFORMATION

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REFERENCES

1. Department for Education UK Government. National Statistics: Schools, Pupils and Their Characteristics. National Tables, available at: https://www.gov.uk/government/uploads/system/uploads/attachment_data/file/552342/SFR20_2016_Main_Text.pdf (2016).
2. Broughton, N., Ezeyi, O., Hupkau, C., Keohane, N. & Shorthouse, R. Open Access: An Independent Evaluation, available at: http://www.smc.co.uk/wp-content/uploads/2014/07/Open-Access-an-independent-evaluation-Embargoed-00-01-030714.pdf (2014).
3. Coe, R. et al. Evidence on the Effects of Selective Educational Systems, available at: https://www.gov.gg/CHTHPHandler.aspx?id=974858p&=0 (2008).
4. Deary, I. J., Ferron, J. M. & Megnoli, C. The effect of educational attainment and wages. Rev. Econ. Stat. 84, 1-20 (2002).
5. Schmidt, F. L. & Hunter, J. General mental ability in the world of work: occupational attainment and job performance. J. Pers. Soc. Psychol. 86, 162 (2004).
6. Strenze, T. Intelligence and socioeconomic success: A meta-analytic review of longitudinal research. Intelligence 35, 401-426 (2007).
7. Sirin, S. R. Socioeconomic status and academic achievement: A meta-analytic review of research. Rev. Educ. Res. 75, 417-453 (2005).
8. Bradley, R. H. & Corwyn, R. F. Socioeconomic status and child development. Annu. Rev. Psychol. 53, 371-399 (2002).
9. Andrews, J., Hutchinson, J. & Johnes, R. Grammar Schools and Social Mobility, (Education Policy Institute, London, 2016). Available at: http://epi.org.uk/wpcontent/uploads/2016/09/grammar-schools-and-social-mobility_.pdf (2017).
10. Sullivan, A. & Heath, A. State and Private Schools in England and Wales. (University of Oxford, 2002).
11. Department for Education UK Government. Revised GCSE and Equivalent Results in England, available at: https://www.gov.uk/government/statistics/revised-gcse-and-equivalent-results-in-england-2015-to-2016 (2016).
12. Goldstein, H. & Sammons, P. The influence of secondary and junior schools on sixteen year examination performance: A cross-classified multilevel analysis. Sch. Eff. Sch. Improv. 8, 219-230 (1997).
13. Clark, D. Selective schools and academic achievement. BE J Econ Anal Policy 10, 20-24 (2010).
14. Anderson, K., Gong, X., Hong, K. & Zhang, X. Do selective high schools improve student achievement? Effects of school exams in China. China Econ. Rev. 40, 121–134 (2016).
15. Atkinson, A., Gregg, P. & McConnell, B. The Result of 11 plus Selection: An Investigation into Opportunities and Outcomes for Pupils in Selective Lea. (Centre for Market and Public Organisation Working Paper, UK, Bristol, 2006).
16. Waldofjel, J. & Washbrook, E. On Your Marks: Measuring the School Readiness of Children in Low-to-middle Income Families (Resolution Foundation, London, 2011).
17. Knopik, V. S., Neiderheiser, J. D., DeFries, J. C. & Plomin, R. Behavioral Genetics 7th edn (Worth Publishers, New York, 2017).
18. Plomin, R. & Deary, I. J. Genetics and intelligence differences: five special findings. Mol. Psychol. 20, 98–108 (2015).
19. Krapohl, E. et al. The high heritability of educational achievement reflects many genetically influenced traits, not just intelligence. Proc. Natl Acad. Sci. 111, 15273–15278 (2014).
20. Wainwright, M. A., Wright, M. J., Luciano, M., Geffen, G. M. & Martin, N. G. Multivariate genetic analysis of academic skills of the Queensland core skills test and IQ highlight the importance of genetic. Twin Res. Human. Genet. 8, 602–608 (2005).
21. Bartels, M., Rietveld, M. J., Van Baal, G. C. M. & Boomsma, D. I. Heritability of educational achievement in 12-year-olds and the overlap with cognitive ability. Twin Res. 5, 544–553 (2002).
22. Thompson, L. A., Dettman, D. K. & Plomin, R. Associations between cognitive abilities and scholastic achievement: Genetic overlap but environmental differences. Psychol. Sci. 2, 158–165 (1991).
23. Calvin, C. M. et al. Multivariate genetic analyses of cognition and academic achievement from two population samples of 174,000 and 166,000 school children. Behav. Genet. 42, 699–710 (2012).
24. Yang, J., Lee, S. H., Goddard, M. E. & Visscher, P. M. Genome-wide Association Studies and Genomic Prediction (GCTA): Methods, Data Analyses, and Interpretations. In: Genome-Wide Association Studies and Genomic Prediction. Methods in Molecular Biology (Methods and Protocols), vol 1019. (eds Gondro C., van der Werf J., Hayes B.) 215–236 (Humana Press, Totowa, NJ, 2013).
25. Yang, J., Lee, S. H., Goddard, M. E. & Visscher, P. M. GCTA: a tool for genome-wide complex trait analysis. Ann. J. Human. Genet. 88, 76–82 (2011).
26. Trzaskowski, M. et al. Genetic influence on family socioeconomic status and children’s intelligence. Intelligence 42, 83–88 (2014).
27. Krapohl, E. & Plomin, R. Genetic link between family socioeconomic status and children’s educational achievement estimated from genome-wide SNPs. Mol. Psychiatry 21, 437 (2016).
28. Trzaskowski, M., Shakeshaft, N. G. & Plomin, R. Intelligence indexes generalist genes for cognitive abilities. Intelligence 41, 560–565 (2013).
29. Trzaskowski, M. et al. DNA evidence for strong genome-wide pleiotropy of cognitive and learning abilities. Behav. Genet. 43, 267–273 (2013).
30. Plomin, R. et al. Common DNA markers can account for more than half of the genetic influence on cognitive abilities. Psychol. Sci. 24, 562–568 (2013).
31. Harlaar, N. et al. A behavioural genetic analysis of DNA markers associated with general cognitive ability in 7-year-olds. J. Child Psychol. Psychiatry 46, 1097–1107 (2005).
32. Wray, N. R. et al. Research review: polygenic methods and their application to psychiatric traits. J. Child Psychol. Psychiatry 55, 1068–1087 (2014).

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37. Selzam, S. et al. Predicting educational achievement from DNA. Mol. Psychiatry 22, 267–272 (2017).
38. Rietveld, C. A. et al. GWAS of 126,559 individuals identifies genetic variants associated with educational attainment. Science 340, 1467–1471 (2013).
39. Okbay, A. et al. Genome-wide association study identifies 74 loci associated with educational attainment. Nature 533, 539–542 (2016).
40. Magnusson, K. Interpreting Cohen’s d Effect Size: An interactive Visualization, available at: http://rpsychologist.com/d3/cohend/ (2014).
41. Rutter, M. et al. Integrating nature and nurture: Implications of person–environment correlations and interactions for developmental psychopathology. Dev. Psychopathol. 9, 335–364 (1997).
42. Gibbons, S. & Silva, O. Faith primary schools: better schools or better pupils? J. Labor. Econ. 29, 589–635 (2011).
43. Iannelli, C. The role of the school curriculum in social mobility. Br. J. Sociol. Educ. 34, 907–928 (2013).
44. Haworth, C. M., Davis, O. S. & Plomin, R. Twins Early Development Study (TEDS): a genetically sensitive investigation of cognitive and behavioral development from childhood to young adulthood. Twin Res. Human. Genet. 16, 117–125 (2013).
45. Kovas, Y. et al. The genetic and environmental origins of learning abilities and disabilities in the early school years. Monogr. Society Res. Child Dev. 72, vii–144 (2007).
46. Kaplan, E., Fein, D., Kramer, J., Delis, D. & Morris, R. The WISC-III as A Process Instrument (The Psychological Corporation, New York, 1999).
47. Raven, J. Raven Progressive Matrices. In: Handbook of Nonverbal Assessment (ed. McCallum R.S.) (Springer, Boston, MA, 2003).
48. Wechsler, D. Wechsler Intelligence Scale for Children — Third Edition UK (WISC-III UK) Manual (Psychological Corporation, London, 1949).

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