## Appendix A

## Appendix Figure A1: McCrary Test



Notes: The figure shows the fraction of study participants by day of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show nonparametric birth cohort trends. The estimated discontinuity of the density is -0.0201 with a standard error of $0.0174 . N=271,234$.

## Appendix Figure A2: Male



The figure shows the fraction of male study participants by quarter of birth. The dashed vertical line marks the first birth cohort affected 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure A3: White



Notes:
The figure shows the fraction of white study participants by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure A4: Mixed Ethnicity



Notes: The figure shows the fraction of study participants of mixed ethnicity by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure A5: Asian



Notes: The figure shows the fraction of Asian study participants by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure A6: Black



Notes: The figure shows the fraction of black study participants by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

Appendix Figure A7: Other Ethnicity


Notes: The figure shows the fraction of study participants of another ethnicity by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

Appendix Figure A8: Born in England


Notes: The figure shows the fraction of study participants born in England by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure A9: Born in Wales



Notes: The figure shows the fraction of study participants born in Wales by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

Appendix Figure A10: Born in Scotland


Notes: The figure shows the fraction of study participants born in Scotland by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure A11: Right Handed



Notes: The figure shows the fraction of right-handed study participants by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,023$.

## Appendix Figure A12: Left Handed



Notes: The figure shows the fraction of left-handed study participants by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,023$.

## Appendix Figure A13: Ambidextrous



Notes: The figure shows the fraction of ambidextrous study participants by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,023$.

## Appendix Figure A14: Adopted



Notes: The figure shows the fraction of study participants who were adopted by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,723$.

## Appendix Figure A15: Twin



Notes:
The figure shows the fraction of study participants who were twins by quarter of birth. This question was not asked to those who had been adopted. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=267,130$.

Appendix Table A1: Balance Control Test

|  | Male | White | Mixed <br> Ethnicity | Asian | Black | Other <br> Ethnicity | Born in England |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Post | $\begin{gathered} 0.008 \\ {[0.006]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.002]} \end{gathered}$ | $\begin{gathered} 0.001 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} 0.001 \\ {[0.000]} \end{gathered}$ | $\begin{gathered} -0.000 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} 0.000 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.004]} \end{gathered}$ |
| Mean of Y | $\begin{gathered} 271,082 \\ 0.436 \end{gathered}$ | $\begin{gathered} 271,082 \\ 0.983 \end{gathered}$ | $\begin{aligned} & 271,082 \\ & 0.00511 \end{aligned}$ | $\begin{aligned} & 271,082 \\ & 0.00130 \end{aligned}$ | $\begin{aligned} & 271,082 \\ & 0.00501 \end{aligned}$ | $\begin{aligned} & 271,082 \\ & 0.00274 \end{aligned}$ | $\begin{gathered} 271,082 \\ 0.862 \end{gathered}$ |
|  | Born in Wales | Born in Scotland | Right <br> Handed | Left Handed | Ambidextrous | Adopted | Twin |
| Post | $\begin{gathered} -0.000 \\ {[0.003]} \end{gathered}$ | $\begin{gathered} 0.001 \\ {[0.004]} \end{gathered}$ | $\begin{gathered} 0.003 \\ {[0.004]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.004]} \end{gathered}$ | $\begin{gathered} -0.002 \\ {[0.002]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} -0.006 \\ {[0.002]^{* * *}} \end{gathered}$ |
| $N$ | 271,082 | 271,082 | 271,023 | 271,023 | 271,023 | 270,723 | 267,130 |
| Mean of Y | 0.0466 | 0.0915 | 0.881 | 0.103 | 0.0158 | 0.0127 | 0.0252 |

Notes: The table investigates whether predetermined characteristics are smooth are around the September 1, 1957 cutoff. It reports the coefficient on an indicator for being born on or after September 1, 1957 (i.e., "Post") from regressions where the dependent variables is listed in the column. The regressions also included quadratic polynomials in date of birth, which were allowed to differ on either side of the cutoff. The mean of Y corresponds to the average of the dependent variable among those born in the 12 months before September 1, 1957.

## Appendix Figure A16: East Coordinate of Birth Place


shows the pre- and post-reform CDFs of east coordinate of place of birth. The pre-reform CDF is the CDF in the limit when date of birth is converging to September 1, 1957 from the left. The post-reform $C D F$ is the CDF in the limit when date of birth is converging to September 1, 1957 from the right. $N=266,883$.

Appendix Figure A17: North Coordinate of Birth Place


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Notes:
The figure shows the pre- and post-reform CDFs of subischial height. Subischial height is the difference between standing height and sitting height. The pre-reform $C D F$ is the CDF in the limit when date of birth is converging to September 1, 1957 from the left. The post-reform $C D F$ is the CDF in the limit when date of birth is converging to September 1,1957 from the right. $N=271,173$.

## Appendix Figure A19: Fraction Missing Genetic Data



Notes: The figure shows the fraction of study participants with genetic data available by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. The discontinuity is 0.0044 with a standard error of 0.0031 (p-value of 0.14 ). The mean among those born in the 12 months before the cutoff is $0.0591 . N=271,234$.

## Appendix Figure A20: Body Mass Index Polygenic Score



Notes: The figure shows the pre- and post-reform CDFs of the polygenic score for BMI. The pre-reform CDF is the CDF in the limit when date of birth is converging to September 1,1957 from the left. The post-reform $C D F$ is the CDF in the limit when date of birth is converging to September 1, 1957 from the right. $N=253,715$.

## Appendix Figure A21: Educational Achievement Polygenic Score



Notes: The figure shows the pre- and post-reform CDFs of the polygenic score for educational achievement. The pre-reform CDF is the CDF in the limit when date of birth is converging to September 1, 1957 from the left. The post-reform CDF is the CDF in the limit when date of birth is converging to September 1, 1957 from the right. $N=253,715$.

## Appendix Table A2: Distributional Test

| Coordinates of |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| Birth Place |  |  |  |  |
| East | North | Subischial <br> Height | Polygenic Scores |  |
| Educational |  |  |  |  |

Notes: The table show the p-values of tests of the equality of the pre- and post-reform CDFs. $\mathrm{N}=266,883$ (coordinates of place of birth); 269,173 (subischial height); and 253,715 (polygenic scores for BMI and educational achievement)

Appendix B

## Appendix Figure B1: Average of Body Size Index by Quarter of Birth



Notes: The figure assesses the sensititivity of the results for body size index to the choice of bandwidth and to the use of linear trends. It shows the average of body size index by quarter of birth. The left-hand side column uses quadratic trends in quarter of birth. The right-hand side column uses linear trends in quarter of birth. The top row uses a 10 -year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The circumference of each circle reflects the number of participants born in that quarter.

## Appendix Figure B2: Average of Lung Function Index by Quarter of Birth



Notes: The figure assesses the sensititivity of the results for lung function index to the choice of bandwidth and to the use of linear trends. It shows the average of lung function index by quarter of birth. The left-hand side column uses quadratic trends in quarter of birth. The righthand side column uses linear trends in quarter of birth. The top row uses a 10-year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3 -year bandwidth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The circumference of each circle reflects the number of participants born in that quarter.

## Appendix Figure B3: Average of Blood Pressure Index by Quarter of Birth



Notes: The figure assesses the sensititivity of the results for blood pressure index to the choice of bandwidth and to the use of linear trends. It shows the average of blood pressure index by quarter of birth. The left-hand side column uses quadratic trends in quarter of birth. The righthand side column uses linear trends in quarter of birth. The top row uses a 10 -year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3 -year bandwidth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The circumference of each circle reflects the number of participants born in that quarter.

## Appendix Figure B4: Average of Summary Index by Quarter of Birth



Notes: The figure assesses the sensititivity of the results for the summary index to the choice of bandwidth and to the use of linear trends. It shows the average of the summary index by quarter of birth. The left-hand side column uses quadratic trends in quarter of birth. The right-hand side column uses linear trends in quarter of birth. The top row uses a 10 -year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The circumference of each circle reflects the number of participants born in that quarter.

## Appendix Figure B5: Distributional Effects on Body Size (No Controls)



Notes: The figure assesses the sensititivity of the distributional effects on the body size index to the choice of bandwidth and to the use of linear trends. It shows the pre- and post-reform CDFs for complies of the body size index. The left-hand side column uses quadratic trends in date of birth. The right-hand side column uses linear trends in date of birth. The top row uses a 10 -year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. No controls.

## Appendix Figure B6: Distributional Effects on Body Size (With Controls)



Notes: The figure assesses the sensititivity of the distributional effects on the body size index to the choice of bandwidth and to the use of linear trends. It shows the pre- and post-reform CDFs for complies of the body size index. The left-hand side column uses quadratic trends in date of birth. The right-hand side column uses linear trends in date of birth. The top row uses a 10 -year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. The regressions include the following set of controls: gender, age in days (at the time of the baseline assessment) and age squared, dummies for ethnicity, dummies for country of birth, and dummies for calendar month of birth.

## Appendix Figure B7: Distributional Effects on Lung Function (No Controls)



Notes: The figure assesses the sensititivity of the distributional effects on the lung function index to the choice of bandwidth and to the use of linear trends. It shows the pre- and post-reform CDFs for complies of the lung function index. The left-hand side column uses quadratic trends in date of birth. The right-hand side column uses linear trends in date of birth. The top row uses a 10 -year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. No controls.

## Appendix Figure B8: Distributional Effects on Lung Function (With Controls)



Notes: The figure assesses the sensititivity of the distributional effects on the lung function index to the choice of bandwidth and to the use of linear trends. It shows the pre- and post-reform CDFs for complies of the lung function index. The left-hand side column uses quadratic trends in date of birth. The right-hand side column uses linear trends in date of birth. The top row uses a 10-year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. The regressions include the following set of controls: gender, age in days (at the time of the baseline assessment) and age squared, dummies for ethnicity, dummies for country of birth, and dummies for calendar month of birth.

## Appendix Figure B9: Distributional Effects on Blood Pressure (No Controls)



Notes: The figure assesses the sensititivity of the distributional effects on the blood pressure index to the choice of bandwidth and to the use of linear trends. It shows the pre- and post-reform CDFs for complies of the blood pressure index. The left-hand side column uses quadratic trends in date of birth. The right-hand side column uses linear trends in date of birth. The top row uses a 10-year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. No controls.

## Appendix Figure B10: Distributional Effects on Blood Pressure (With Controls)



Notes: The figure assesses the sensititivity of the distributional effects on the blood pressure index to the choice of bandwidth and to the use of linear trends. It shows the pre- and post-reform CDFs for complies of the blood pressure index. The left-hand side column uses quadratic trends in date of birth. The right-hand side column uses linear trends in date of birth. The top row uses a 10 -year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3-year bandwidth. The regressions include the following set of controls: gender, age in days (at the time of the baseline assessment) and age squared, dummies for ethnicity, dummies for country of birth, and dummies for calendar month of birth.

Appendix Table B1: P-values of Distributional Tests for Body Size

| 3 Years | $\mathbf{5}$ Years |  | $\mathbf{1 0}$ Years |  |
| :---: | :---: | :---: | :---: | :---: |
| Linear | Quad. | Linear | Quad. | Linear | Quad. C

Full Distribution

| No Controls | 0.2164 | 0.5310 | 0.0668 | 0.3070 | 0.0060 | 0.0896 |
| ---: | :--- | :--- | :--- | :--- | :--- | :--- |
| With Controls | 0.1656 | 0.3780 | 0.0674 | 0.2598 | 0.0050 | 0.0932 |

## Bottom Half

| No Controls | 0.9660 | 0.8154 | 0.9602 | 0.9730 | 0.4158 | 0.9526 |
| ---: | :--- | :--- | :--- | :--- | :--- | :--- |
| With Controls | 0.8186 | 0.4496 | 0.9716 | 0.8514 | 0.3812 | 0.9396 |


| Top Half |  |  |  |  |  |  |
| ---: | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| No Controls | 0.0554 | 0.3446 | 0.0094 | 0.1058 | 0.0002 | 0.0126 |
| With Controls | 0.0450 | 0.3266 | 0.0094 | 0.0950 | 0.0002 | 0.0138 |

## Appendix Table B2: P-values of Distributional Tests for Lung Function

| 3 Years |  | $\mathbf{5}$ Years |  | $\mathbf{1 0}$ Years |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Linear | Quad. | Linear | Quad. | Linear | Quad. |

Full Distribution

| No Controls | 0.1354 | 0.1234 | 0.2012 | 0.1964 | 0.0768 | 0.1712 |
| ---: | :---: | :---: | :---: | :---: | :---: | :---: |
| With Controls | 0.1578 | 0.1572 | 0.2620 | 0.2340 | 0.0706 | 0.2352 |

## Bottom Half

| No Controls | 0.3672 | 0.4248 | 0.4022 | 0.3768 | 0.0552 | 0.5962 |
| ---: | :--- | :--- | :--- | :--- | :--- | :--- |
| With Controls | 0.3744 | 0.6082 | 0.5066 | 0.4406 | 0.0626 | 0.7634 |

## Top Half

$\begin{array}{llllllll}\text { No Controls } & 0.0626 & 0.0438 & 0.1102 & 0.1168 & 0.1090 & 0.0618\end{array}$
$\begin{array}{llllllll}\text { With Controls } & 0.0778 & 0.0524 & 0.1402 & 0.1396 & 0.0754 & 0.0870\end{array}$
Notes: The table shows the p-values of tests of the equality of the full distribution, the bottom and top halves of the pre- and post-reform CDFs of the lung function index.

## Appendix Table B3: P-values of Distributional Tests for Blood Pressure

| 3 Years | 5 Years |  | $\mathbf{1 0}$ Years |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Linear | Quad. | Linear | Quad. | Linear | Quad. |

Full Distribution

| No Controls | 0.0208 | 0.1480 | 0.0414 | 0.0262 | 0.5432 | 0.0362 |
| ---: | :--- | :--- | :--- | :--- | :--- | :--- |
| With Controls | 0.0306 | 0.2006 | 0.0552 | 0.0358 | 0.5856 | 0.0532 |


| Bottom Half |  |  |  |  |  |  |
| ---: | :--- | :--- | :--- | :--- | :--- | :--- |
| No Controls | 0.0100 | 0.0836 | 0.0112 | 0.0126 | 0.4152 | 0.0102 |
| With Controls | 0.0226 | 0.1806 | 0.0196 | 0.0266 | 0.4658 | 0.0172 |


| Top Half |  |  |  |  |  |  |
| ---: | :--- | :--- | :--- | :--- | :--- | :--- |
| No Controls | 0.0420 | 0.2772 | 0.1984 | 0.0500 | 0.6924 | 0.1502 |
| With Controls | 0.0380 | 0.2264 | 0.2008 | 0.0514 | 0.7240 | 0.1684 |

Notes: The table shows the p-values of tests of the equality of the full distribution, the bottom and top halves of the pre- and post-reform CDFs of the blood pressure index.

## Appendix Figure B11: Effect on Percentiles of Distribution of Body Size Index



Notes: The figure shed lights on the results shown in Figure 4 of the paper. It shows the fraction of study participants with a body size index below the $10^{\text {th }}, 25^{\text {th }}, 50^{\text {th }}, 75^{\text {th }}, 90^{\text {th }}$, and the $95^{\text {th }}$ percentile (of the distribution of those born between September 1,1956 and August 31 , 1957) by quarter of birth. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=266,525$.

Appendix Figure B12: Effect on Percentiles of Distribution of Lung Function Index


Notes: The figure shed lights on the results shown in Figure 5 of the paper. It shows the fraction of study participants with a body size index below the $10^{\text {th }}, 25^{\text {th }}, 50^{\text {th }}, 75^{\text {th }}, 90^{\text {th }}$, and the $95^{\text {th }}$ percentile (of the distribution of those born between September 1,1956 and August 31 , 1957) by quarter of birth. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=215,366$.

Appendix Figure B13: Effect on Percentiles of Distribution of Blood Pressure Index


Notes: The figure shed lights on the results shown in Figure 6 of the paper. It shows the fraction of study participants with a body size index below the $10^{\text {th }}, 25^{\text {th }}, 50^{\text {th }}, 75^{\text {th }}, 90^{\text {th }}$, and the $95^{\text {th }}$ percentile (of the distribution of those born between September 1,1956 and August 31 , 1957) by quarter of birth. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,647$.

## Appendix Figure B14: Distributional Effects on Body Size Index



Notes: The figure shows the difference between the pre- and post-reform CDFs for compliers and $95 \%$ confidence bands. The top figure reproduces Figure 4 in the paper, showing the pre- and post-reform CDF of body size index for compliers. The black solid line in the bottom figure shows the difference between the post- and pre-reform CDFs shown in the top figure. The blue areas show $95 \%$ confidence intervals. Inference based on these confidence intervals is problematic because it leads to a large number of highly correlated statistical tests, raising concerns about multiple hypothesis testing.

## Appendix Figure B15: Distributional Effects on Lung Function Index



Notes: The figure shows the difference between the pre- and post-reform CDFs for compliers and $95 \%$ confidence bands. The top figure reproduces Figure 5 in the paper, showing the pre- and post-reform CDF of lung function index for compliers. The black solid line in the bottom figure shows the difference between the post- and pre-reform CDFs shown in the top figure. The blue areas show $95 \%$ confidence intervals. Inference based on these confidence intervals is problematic because it leads to a large number of highly correlated statistical tests, raising concerns about multiple hypothesis testing.

## Appendix Figure B16: Distributional Effects on Blood Pressure Index



Notes: The figure shows the difference between the pre- and post-reform CDFs for compliers and $95 \%$ confidence bands. The top figure reproduces Figure 6 in the paper, showing the pre- and post-reform CDF of blood pressure index for compliers. The black solid line in the bottom figure shows the difference between the post- and pre-reform CDFs shown in the top figure. The blue areas show $95 \%$ confidence intervals. Inference based on these confidence intervals is problematic because it leads to a large number of highly correlated statistical tests, raising concerns about multiple hypothesis testing.

Appendix Figure B16: Body Mass Index


Notes: The figure shows average BMI by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 schoolleaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,019$.

## Appendix Figure B17: Overweight



[^1]
## Appendix Figure B18: Obese



Notes: The figure shows the fraction of study participants who were obese by quarter of birth. Obesity is defined as having a BMI greater or equal to 30 . The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,019$.

## Appendix Figure B19: Blood Pressure Systolic



Notes: The figure shows average systolic blood pressure by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,647$.

## Appendix Figure B20: Diastolic Blood Pressure



Notes: The figure shows average diastolic blood pressure by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,647$.

## Appendix Figure B21: Stage 1 Hypertension



Notes: The figure shows the fraction of participants with stage 1 hypertension by quarter of birth. Stage 1 hypertension is defined as having a diastolic blood pressure greater or equal to 80 or having a systolic blood pressure greater or equal to 130 . The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15. The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,647$.


Notes: The figure shows the fraction of participants with stage 2 hypertension by quarter of birth. Stage 2 hypertension is defined as having a diastolic blood pressure greater or equal to 90 or having a systolic blood pressure greater or equal to 140 . The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=270,647$.

## Appendix Table B4: BMI, Overweight, and Obesity

|  | BMI |  | Overweight |  | Obesity |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Reduced-form |  |  |  |  |  |  |
| Post | -0.061 | -0.070 | 0.003 | 0.001 | -0.011 | -0.012 |
|  | [0.063] | [0.063] | [0.006] | [0.006] | [0.005]** | [0.005]** |
| Two stages least squares |  |  |  |  |  |  |
| Stayed in school until 16 | -0.407 | -0.504 | 0.020 | 0.009 | -0.075 | -0.085 |
|  | [0.420] | [0.453] | [0.040] | [0.043] | [0.036]** | [0.039]** |
| Controls? | No | Yes | No | Yes | No | Yes |
| Mean of Y | 27.41 | 27.41 | 0.65 | 0.65 | 0.25 | 0.25 |
| $N$ Observations | 270,019 | 270,019 | 270,019 | 270,019 | 270,019 | 270,019 |

Notes: The table shows the effects on average BMI, the fraction overweight, and the fraction obese. The first two rows show reduced-form effects of the 1972 Raising of the School Leaving Age. The last two rows show two stages least squares estimates of the effect of staying in school until age 16 obtained by using an indicator for being born on or after September 1, 1957 to instrument for staying in school until age 16. Robust standard errors. Controls include male, age in days and age squared, dummies for calendar month of birth, dummies for ethnicity, and dummies for country of birth.

## Appendix Table B5: Blood Pressure

|  | Systolic Blood Pressure |  | Diastolic Blood Pressure |  | Stage 1 <br> Hypertension |  | Stage 2 <br> Hypertension |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Reduced-form |  |  |  |  |  |  |  |  |
| Post | 0.426 | 0.311 | 0.243 | 0.208 | 0.012 | 0.010 | 0.006 | 0.005 |
|  | [0.213]** | [0.209] | [0.130]* | [0.128] | [0.006]** | [0.006]* | [0.006] | [0.006] |
| Two stages least squares |  |  |  |  |  |  |  |  |
| Stayed in school until 16 | 2.836 | 2.234 | 1.619 | 1.492 | 0.082 | 0.071 | 0.041 | 0.037 |
|  | [1.420]** | [1.508] | [0.866]* | [0.921] | [0.039]** | [0.042]* | [0.041] | [0.043] |
| Controls? | No | Yes | No | Yes | No | Yes | No | Yes |
| Mean of Y | 133.80 | 133.80 | 82.66 | 82.66 | 0.68 | 0.68 | 0.38 | 0.38 |
| $N$ Observations | 270,647 | 270,647 | 270,647 | 270,647 | 270,647 | 270,647 | 270,647 | 270,647 |

Notes: The table shows the effects on average systolic blood pressure, diastolic blood pressure, stage 1 hypertension, and stage 2 hypertension. Stage 1 hypertension is defined as having a systolic blood pressure greater or equal to 130 or a diastolic blood pressure greater or equal to 80 . Stage 2 hypertension is defined as having a systolic blood pressure greater or equal to 140 or a diastolic blood pressure greater or equal to 90 . The first two rows show reduced-form effects of the 1972 Raising of the School Leaving Age. The last two rows show two stages least squares estimates of the effect of staying in school until age 16 obtained by using an indicator for being born on or after September 1, 1957 to instrument for staying in school until age 16. Robust standard errors. Controls include male, age in days and age squared, dummies for calendar month of birth, dummies for ethnicity, and dummies for country of birth.

Appendix C

Appendix Figure C1: Missing Body Mass Index


Notes: The figure shows the fraction of study participants for whom data on BMI was missing by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure C2: Missing Body Fat Percentage



[^2]
## Appendix Figure C3: Missing Waist-Hip Ratio



Notes: The figure shows the fraction of study participants for whom data on waist-hip ratio was missing by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

Appendix Figure C4: Missing Body Size Index


Notes: The figure shows the fraction of study participants for whom data on body size index was missing by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

Appendix Figure C5: Missing Lung Function Index


Notes: The figure shows the fraction of study participants for whom spirometry data was missing by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

Appendix Figure C6: Missing Blood Pressure Index


Notes: The figure shows the fraction of study participants for whom data on blood pressure was missing by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

## Appendix Figure C7: Missing Summary Index



Notes: The figure shows the fraction of study participants for whom data on the summary index was missing by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

Appendix Table C1: Missing Outcomes

| Post | BMI |  | Body Fat Percentage |  | Waist-hip Ratio |  | Body Size Index |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\begin{gathered} -0.002 \\ {[0.001]^{* * *}} \end{gathered}$ | $\begin{gathered} -0.002 \\ {[0.001]^{* * *}} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} -0.000 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.001]} \end{gathered}$ | $\begin{gathered} -0.001 \\ {[0.002]} \end{gathered}$ |
| Controls? <br> Mean of Y | $\begin{gathered} \mathrm{N} \\ 0.00467 \end{gathered}$ | $\begin{gathered} \mathrm{Y} \\ 0.00467 \end{gathered}$ | $\begin{gathered} \mathrm{N} \\ 0.0144 \end{gathered}$ | $\begin{gathered} \mathrm{Y} \\ 0.0144 \end{gathered}$ | $\begin{gathered} \mathrm{N} \\ 0.00320 \end{gathered}$ | $\begin{gathered} \mathrm{Y} \\ 0.00320 \end{gathered}$ | $\begin{gathered} \mathrm{N} \\ 0.0147 \end{gathered}$ | $\begin{gathered} \mathrm{Y} \\ 0.0147 \end{gathered}$ |
|  | Lung Function Index |  | Blood Pressure Index |  | Summary Index |  |  |  |
| Post | $\begin{gathered} -0.006 \\ {[0.005]} \end{gathered}$ | $\begin{gathered} -0.005 \\ {[0.005]} \end{gathered}$ | $\begin{gathered} 0.000 \\ {[0.000]} \end{gathered}$ | $\begin{gathered} -0.000 \\ {[0.000]} \end{gathered}$ | $\begin{gathered} -0.005 \\ {[0.005]} \end{gathered}$ | $\begin{gathered} -0.005 \\ {[0.005]} \end{gathered}$ |  |  |
| Controls? |  | Y |  | Y |  | Y |  |  |
| Mean of Y | 0.200 | $0.200$ | $0.00130$ | $0.00130$ | $0.210$ | $0.210$ |  |  |

Notes: The table investigates whether there are discontinuities in missing outcomes at the September 1, 1957 cutoff. It reports the coefficient on an indicator for being born on or after September 1, 1957 (i.e., "Post") from regressions where the dependent variables is listed in the column. The regressions also included quadratic polynomials in date of birth, which were allowed to differ on either side of the cutoff. The mean of Y corresponds to the fraction of study participants born in the 12 months before September 1, 1957 for whom the outcome of interest was missing.

## Appendix D

## Appendix Figure D1: Map with Locations of 22 Assessment Centers



Notes: The figure shows the location of the 22 assessment centers (as well as the location of the pilot study).

## Appendix Figure D2: Fraction Staying in School until Age 17 by Quarter of Birth



Notes: The figure shows the fraction of study participants who stayed in school until age 17 by quarter of birth for different specifications. The left-hand side column uses quadratic trends in quarter of birth. The right-hand side column uses linear trends in quarter of birth. The top row uses a 10-year bandwidth. The middle row uses a 5 -year bandwidth. The bottom row uses a 3 -year bandwidth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The circumference of each circle reflects the number of participants born in that quarter.

Appendix Table D1: Effect of 1972 ROLSA on Fraction Staying in School until Age 17

10 Years
Quadratic
Linear

|  | Quadratic |  | Linear |  |
| :---: | :---: | :---: | :---: | :---: |
| Post | 0.028 | 0.018 | -0.011 | -0.017 |
|  | [0.006]*** | [0.006]*** | [0.004]*** | [0.004]*** |


| Controls? <br> $N$ | N | Y N |  | Y |
| :---: | :---: | :---: | :---: | :---: |
|  |  | 271,082 |  |  |
|  | 5 Years |  |  |  |
|  | Quadratic |  | Linear |  |
| Post | $\begin{gathered} 0.027 \\ {[0.009]^{* * *}} \end{gathered}$ | $\begin{gathered} 0.008 \\ {[0.009]} \end{gathered}$ | $\begin{gathered} 0.015 \\ {[0.006]^{* *}} \end{gathered}$ | $\begin{gathered} 0.005 \\ {[0.006]} \end{gathered}$ |
| Controls? | N | Y | N | Y |
| $N$ | 129,222 |  |  |  |
|  | 3 Years |  |  |  |
|  | Quadratic |  | Linear |  |
| Post | 0.038 | 0.009 | 0.024 | 0.009 |
|  | [0.011]*** | [0.012] | [0.008]*** | [0.008] |


| Controls? | N | Y | N | Y |
| ---: | :--- | :--- | :--- | :--- |
| $N$ |  |  | 76,901 |  |

[^3]

Notes: The figure shows the joint distribution of body size and blood pressure indices among compliers born in the 12 months before September 1, 1957. The circumference of each circle reflects the mass in that interval. $N=2,210$.

Appendix Table D2: The categories of the 2000 National Statistics Socio-economic Classification (NS-SEC)

1 Higher managerial and professional occupations
2 Lower managerial and professional occupations
3 Intermediate occupations
4 Small employers and own account workers
5 Lower supervisory and technical occupations
6 Semi-routine occupations
7 Routine occupations
Notes: The table shows the cateogires of the 2000 National Statistics Socio-economic
Classification (NS-SEC) of occupations.

Appendix Figure D4: Pre-Reform Cumulative Distribution of Body Size Index for Compliers and for Entire Population


Notes: The figure shows the pre-reform CDFs of body size index for compliers (black dashed) and for the entire population (red solid). The pre-reform CDF is the CDF in the limit when date of birth is converging to September 1, 1957 from the left. $N=33,228$ (compliers) and 158,707 (all).

## Appendix Figure D5: Pre-Reform Cumulative Distribution of Lung Function Index for Compliers and for Entire Population



## Appendix Figure D6: Pre-Reform Cumulative Distribution of Blood Pressure Index for Compliers and for Entire Population



Notes: The figure shows the pre-reform CDFs of blood pressure index for compliers (black dashed) and for the entire population (red solid) The pre-reform $C D F$ is the CDF in the limit when date of birth is converging to September 1, 1957 from the left. $N=33,882$ (compliers) and 161,264 (all).

Appendix Figure D7: Comparison with Clark and Royer (2010) and Davies et al. (2018)






Notes: The figures compare the point estimates and the $95 \%$ confidence intervals for 2 SLS estimates of the effect of staying in school until age 16. See respective papers for details about bandwidth, controls, and trends.

Appendix Table D3: Comparison with Clark and Royer (2010) and Davies et al. (2018)

|  | Blood Pressure |  |  | Body Mass Index |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Diastolic | Diastolic > 90 | Systolic | BMI | $B M I>25$ | $B M I>30$ |
| Clark \& Royer |  |  |  |  |  |  |
| Point estimate | 0.99 | 0.01 |  | 0.16 | 0.03 | -0.02 |
| Lower Bound 95\% CI | -1.44 | -0.09 |  | -0.71 | -0.07 | -0.10 |
| Upper Bound 95\% CI | 3.41 | 0.11 |  | 1.02 | 0.13 | 0.05 |
| $N$ | 15,097 | 15,097 |  | 18,473 | 18,473 | 18,473 |
| Barcellos et al. |  |  |  |  |  |  |
| Point estimate | 1.49 | 0.01 | 2.23 | -0.50 | 0.01 | -0.08 |
| Lower Bound 95\% CI | -0.31 | -0.06 | -0.72 | -1.39 | -0.07 | -0.16 |
| Upper Bound 95\% CI | 3.30 | 0.08 | 5.19 | 0.38 | 0.10 | -0.01 |
| $N$ | 270,647 | 270,647 | 270,647 | 269,970 | 269,970 | 269,970 |
| Davies et al. |  |  |  |  |  |  |
| Point estimate | -0.30 |  | -2.67 | -1.09 |  |  |
| Lower Bound 95\% CI | -1.21 |  | -3.93 | -1.40 |  |  |
| Upper Bound 95\% CI | 0.61 |  | -1.42 | -0.78 |  |  |
| $N$ | 21,494 |  | 21,492 | 22,055 |  |  |

Notes: The table compares the point estimates and the $95 \%$ confidence intervals for 2SLS estimates of the effect of staying in school until age 16 on the health outcomes shown in the columns. See respective papers for details about bandwidth, controls, and trends.

Appendix Table D4: Clustering Standard Errors by Day-Month-Year of Birth


Notes: The table shows how the standard error estimates change when we cluster the standard errors by day-month-year of birth.

Appendix E

Appendix Table E1: Effect on Distribution of Annual Household Income
Annual household income below

Stayed in school until 16

| $£ 18,000$ | $£ 31,000$ | $£ 52,000$ | $£ 100,000$ |
| :---: | :---: | :---: | :---: |
| -0.070 | -0.192 | -0.061 | 0.016 |
| $[0.031]^{* *}$ | $[0.043]^{* * *}$ | $[0.045]$ | $[0.026]$ |
| 0.129 | 0.328 | 0.622 | 0.915 |

Notes: The figure shows the effect of staying in school until age 16 on the distribution of annual household income. $N=240,880$.

Appendix Table E2: Effect on Occupation SES

| Stayed in school until 16 | Socioeconomic Class |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $=7$ | $\geq 6$ | $\geq 5$ | $\geq 4$ | $\geq 3$ | $\geq 2$ |
|  | -0.032 | -0.073 | -0.084 | -0.117 | -0.070 | -0.041 |
|  | [0.023] | [0.037]** | [0.040]** | [0.043]*** | [0.050] | [0.044] |
| Mean of $Y$ | 0.0545 | 0.163 | 0.203 | 0.251 | 0.437 | 0.750 |

Notes: The figure shows estimates of the staying in school until age 16 on the socioeconomic class of the participants' occupations. Lower values correspond to higher SES. $N=207,533$.

## Appendix Table E3:

Effect on Car and Home Ownership, Neighborhood SES and Pollution

|  |  | Number | of Cars |  | Home | Townsend | Pollution |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $=0$ | $\leq 1$ | $\leq 2$ | $\leq 3$ | Ownership | Townsend | Index |
| Stayed in school until 16 | -0.030 | -0.091 | -0.017 | -0.017 | 0.005 | -0.497 | -0.056 |
|  | [0.022] | [0.041]** | [0.033] | [0.019] | [0.026] | [0.254]* | [0.088] |
| Mean of Y | 0.0790 | 0.412 | 0.801 | 0.947 | 0.899 | -1.331 | 5.48e-11 |
| $N$ Observations |  | 270, | , 055 |  | 269,363 | 270,705 | 248,333 |

Notes: The figure shows estimates of the staying in school until age 16 on car and home ownership and neighborhood SES and pollution.

Appendix Table E4: Effect on Diet

|  | Calories | \% Sugars | \% Fat | \% Saturated <br> Fat | \% Carbo- <br> hydrates |
| ---: | :---: | :---: | :---: | :---: | :---: |
| Stayed in school until 16 | -86.868 <br> $[153.979]$ | 0.020 <br> $[0.017]$ | -0.030 <br> $[0.017]^{*}$ | -0.019 <br> $[0.008]^{* *}$ | 0.011 <br> $[0.021]$ |
| Mean of Y |  | 2108 | 0.221 | 0.329 | 0.126 |

Notes: The figure shows estimates of the effects of staying in school until age 16 on diet. Study participants were asked about their diet in five different waves (at baseline and four online surveys), such that there are sometimes multiple observations by participant. For this reason, standard errors are clustered at the individual level. $N=268,957$ observations, corresponding to 122,665 study participants.

## Appendix Table E5:

Effect on Smoking, Physical Activity, and Hypertension Diagnosis and Medication

|  | Currently <br> Smoke | Ever <br> Smoke | Hypertension <br> Diagnosis | Hypertension <br> Medication | Physical <br> Activity |
| ---: | :---: | :---: | :---: | :---: | :---: |
| Stayed in school until 16 | -0.002 | 0.043 | -0.035 | 0.002 | 0.456 |
|  | $[0.027]$ | $[0.042]$ | $[0.033]$ | $[0.022]$ | $[2.704]$ |
| Mean of Y | 0.118 | 0.396 | 0.205 | 0.0724 | 28.87 |
| N Observations | 270,937 | 267,384 | 270,700 | 268,315 | 61,701 |

Notes: The figure shows estimates of the effects of staying in school until age 16 on smoking, physical activity, and hypertension diagnosis and medication.

Appendix F

## Power to Detect Mean Effects Versus Distributional Effects

## Introduction

In this section, we illustrate a case where an estimate of average treatment effects will be lower powered than distributional treatment effects. More precisely, we examine a setting where only individuals in the upper portion of the outcome distribution are responsive to treatment. We calculate analytically the power to estimate the treatment effect on the average value of the outcome in the population.

Due to the analytic and computation complexity of the Anderson-Darling-based test used in this paper, we are unable to calculate the power of that test analytically nor by simulation. Instead, we evaluate the power to measure the treatment effect on an indicator of whether the outcome variable is above or below certain values at different parts of the outcome distribution. (See details below.) While this is not a perfect comparison, it is meant to give intuition for when distributional tests may perform better than tests of the average.

We find in this simplified setting that when most individuals are affected by some treatment, estimates of average effects are better powered than estimates of distributional effects. However, when fewer individuals are affected, distributional tests may be better powered.

## Data Generating Process

Let $Y_{i}$ denote some outcome of interest for individual $i$. We are interested in the effect of some treatment $X_{i}$ on $Y_{i}$, where $X_{i}$ is an indicator variable of whether individual $i$ was treated. Let $Y_{0, i}$ denote the potential outcome of individual $i$ in the case that they were not treated and $Y_{1, i}$ denote the potential outcome of individual $i$ in the case that they were.

To simulate distributional effects, we assume that only the top $p_{\tau}$ fraction of the potential outcome distribution is affected by treatment. More precisely, we assume that

$$
Y_{1, i}=\left\{\begin{array}{cc}
Y_{0, i} & \text { if } Y_{0, i} \leq \tau \\
Y_{0, i}+\delta & \text { otherwise }
\end{array}\right.
$$

where $\tau$ is the $\left(1-p_{\tau}\right)$-th percentile of the potential outcome distribution, $\tau \equiv \Phi^{-1}\left(1-p_{\tau}\right)$, and $\delta$ is the effect of treatment on those affected. To maintain monotonicity and simplify this derivation, we assume that $\delta>0$.

In this exercise, we assume that we draw a sample of $N$ individuals from the population, treat a fraction $p_{x}$ of them, and measure their realized outcome $Y_{i}$. In the sections below, we calculate the power to find a statistically significant effect of treatment by looking at the effect of treatment on the average of the outcome and by looking at the effect of treatment on specific parts of the outcome distribution.

## The Distributional Test

Because the distributional test used in this paper is very complicated, both analytically and computationally, it will not be possible to calculate the power of that test. We instead calculate the power of an alternative but related test that is meant to provide intuition for why and under which circumstances a distributional test may be better powered than a test of an average treatment effect.

Specifically, we will consider an indicator variable, $T_{i}$, for whether $Y_{i}$ is greater than some value $t$, and calculate the power to detect an effect of the treatment on $T_{i}$. Recall that the distributional test in this paper is a weighted integral of these treatments effects across a range of values of $t$. Thus, if for the values of $t$ in that range, the power of each corresponding test is greater than the power to detect an effect on the average value of $Y_{i}$, it is likely that the power of the distributional test will similarly be greater.

## Some Intermediate Calculations

In order to perform the power calculations below, we will need to know the values of $\mathrm{E}\left(Y_{i}\right)$, $\operatorname{Var}\left(Y_{i}\right), \mathrm{E}\left(T_{i}\right)$, and $\operatorname{Var}\left(T_{i}\right)$. We first calculate

$$
\begin{gathered}
\mathrm{E}\left(Y_{i}\right)=\mathrm{P}\left(X_{i}=0\right) \mathrm{E}\left(Y_{i} \mid X_{i}=0\right)+\mathrm{P}\left(X_{i}=1\right) \mathrm{E}\left(Y_{i} \mid X_{i}=1\right) \\
=\mathrm{P}\left(X_{i}=1\right) \mathrm{E}\left(Y_{i} \mid X_{i}=1\right) \\
=p_{x} p_{\tau} \delta
\end{gathered}
$$

Next, we calculate

$$
\begin{aligned}
\mathrm{E}\left(Y_{i}^{2}\right)=\mathrm{P}\left(X_{i}\right. & =0) \mathrm{E}\left(Y_{i}^{2} \mid X_{i}=0\right)+\mathrm{P}\left(X_{i}=1\right) \mathrm{E}\left(Y_{i}^{2} \mid X_{i}=1\right) \\
& =\left(1-p_{x}\right)+p_{x} \mathrm{E}\left(Y_{i}^{2} \mid X_{i}=1\right)
\end{aligned}
$$

Note that the variable $\left(Y_{i} \mid X_{i}=1\right)$ is the same as $(Z+\delta I)$ where $Z$ is a standard normal random variable and $I$ is an indicator variable for whether $Z>\tau$. We therefore continue

$$
\begin{gathered}
\mathrm{E}\left(Y_{i}^{2}\right)=\left(1-p_{x}\right)+p_{x} \mathrm{E}\left[(Z+\delta I)^{2}\right] \\
=\left(1-p_{x}\right)+p_{x}\left[\mathrm{E}\left(Z^{2}\right)+2 \mathrm{E}(Z I) \delta+\mathrm{E}\left(I^{2}\right) \delta^{2}\right] \\
=\left(1-p_{x}\right)+p_{x}\left[1+2 \phi(\tau) \delta+p_{\tau} \delta^{2}\right] \\
=1+2 p_{x} \phi(\tau) \delta+p_{x} p_{\tau} \delta^{2},
\end{gathered}
$$

where $\phi($.$) is the standard normal pdf. Finally, this implies that$

$$
\begin{gathered}
\operatorname{Var}\left(Y_{i}\right)=\mathrm{E}\left(Y_{i}^{2}\right)-\mathrm{E}\left(Y_{i}\right)^{2} \\
=1+2 p_{x} \phi(\tau) \delta+p_{x} p_{\tau} \delta^{2}-\left(p_{x} p_{\tau} \delta\right)^{2} \\
=1+2 p_{x} \phi(\tau) \delta+p_{x} p_{\tau}\left(1-p_{x} p_{\tau}\right) \delta^{2} .
\end{gathered}
$$

For the binary variable, we first calculate

$$
\begin{gathered}
\mathrm{E}\left(T_{i}\right)=\mathrm{P}\left(X_{i}=0\right) \mathrm{E}\left(T_{i} \mid X_{i}=0\right)+\mathrm{P}\left(X_{i}=1\right) \mathrm{E}\left(T_{i} \mid X_{i}=1\right) \\
=\left(1-p_{x}\right)[1-\Phi(t)]+p_{x}[1-\Phi(t-\delta)] \\
=1-\left(1-p_{x}\right) \Phi(t)-p_{x} \Phi(t-\delta)
\end{gathered}
$$

This implies that

$$
\operatorname{Var}\left(T_{i}\right)=\left[\left(1-p_{x}\right) \Phi(t)+p_{x} \Phi(t-\delta)\right]\left[1-\left(1-p_{x}\right) \Phi(t)-p_{x} \Phi(t-\delta)\right]
$$

## Power to Detect Changes in the Average Outcome

To estimate the effect of treatment on average health, $\beta_{\text {avg }}$, we regress $Y_{i}$ on $X_{i}$. In that case, we see that

$$
\begin{gathered}
\beta_{\text {avg }}=\frac{\operatorname{Cov}\left(X_{i}, Y_{i}\right)}{\operatorname{Var}\left(X_{i}\right)} \\
=\frac{\mathrm{E}\left(X_{i} Y_{i}\right)-\mathrm{E}\left(X_{i}\right) \mathrm{E}\left(Y_{i}\right)}{\operatorname{Var}\left(X_{i}\right)} \\
=\frac{p_{\tau} \delta p_{x}-p_{\tau} \delta p_{x}^{2}}{p_{x}\left(1-p_{x}\right)} \\
=p_{\tau} \delta .
\end{gathered}
$$

Using the derivations from the previous section, the standard error an estimator of $\beta_{\text {avg }}$ is

$$
\begin{aligned}
& \operatorname{SE}\left(\beta_{\mathrm{avg}}\right)=\sqrt{\frac{\operatorname{Var}\left(Y_{i}\right)-\operatorname{Var}\left(X_{i} \beta_{\mathrm{avg}}\right)}{N \operatorname{Var}\left(X_{i}\right)}} \\
& =\sqrt{\frac{1+2 p_{x} \phi(\tau) \delta+p_{x} p_{\tau}\left(1-p_{\tau}\right) \delta^{2}}{N p_{x}\left(1-p_{x}\right)}}
\end{aligned}
$$

From these expressions, we see that the z -statistic for the average effect will be distributed as

$$
N\left(p_{\tau} \delta \sqrt{\frac{N p_{x}\left(1-p_{x}\right)}{1+2 p_{x} \phi(\tau) \delta+p_{x} p_{\tau}\left(1-p_{\tau}\right) \delta^{2}}}, 1\right)
$$

So the power of a test of whether there is a non-zero mean effect of the policy will be equal to the fraction of the time that this normally distributed random variable achieves a value greater than 1.96 in magnitude.

## Power to Detect Changes in the Distribution of the Outcome

Let $T_{i}$ be defined for some threshold $t$, as described above. We first note that if $t<\tau$, then the treatment effect, $\beta_{t}$, will be equal to zero since the treatment will not induce any individuals below the threshold to cross the threshold. This means that power to estimate an effect of treatment with a $p$-value of less than 0.05 is $5 \%$.

We next consider that case that $t>\tau+\delta$. In this setting, every individual less than $\delta$ units below the threshold $t$ will be above the threshold after treatment. Therefore

$$
\beta_{t}=E\left(T_{i} \mid X_{i}=1\right)-E\left(T_{i} \mid X_{i}=0\right)
$$

$$
\begin{gathered}
=E\left(Y_{i}>t \mid X_{i}=1\right)-E\left(Y_{i}>t \mid X_{i}=0\right) \\
=1-\Phi(t-\delta)-1+\Phi(t) \\
=\Phi(t)-\Phi(t-\delta) \\
=p_{t}-p_{t-\delta}
\end{gathered}
$$

where $p_{t} \equiv \Phi(t)$ is the fraction of individuals where $Y_{0, i} \leq t$.
The standard error is therefore

$$
\begin{gathered}
\operatorname{SE}\left(\beta_{t}\right)=\sqrt{\frac{\operatorname{Var}\left(T_{i}\right)-\operatorname{Var}\left(X_{i} \beta_{\mathrm{t}}\right)}{N \operatorname{Var}\left(X_{i}\right)}} \\
=\sqrt{\frac{\left(1-p_{x}\right) p_{t}\left(1-p_{t}\right)+p_{x} p_{t-\delta}\left(1-p_{t-\delta}\right)}{N p_{x}\left(1-p_{x}\right)}}
\end{gathered}
$$

This means the z-statistics for a test of a treatment effect on $T_{i}$ is distributed

$$
N\left(\left(p_{t}-p_{t-\delta}\right) \sqrt{\frac{N p_{x}\left(1-p_{x}\right)}{\left(1-p_{x}\right) p_{t}\left(1-p_{t}\right)+p_{x} p_{t-\delta}\left(1-p_{t-\delta}\right)}}, 1\right)
$$

As with the calculation of power in the average treatment effect-case, the power in this setting is the fraction of time that this normally distributed random variable exceeds one in absolute value.

We do not calculate the power of a test for values of $t$ in $[\tau, \tau+\delta]$ but rather note that the power will be somewhere between the cases when $t$ is above or below this interval.

## Illustration of Power Calculations in Various Settings

In this illustration, we compare the power of the estimates of the treatment on the average outcome and the effect of the treatment on the binary outcomes. In these calculations, we set $\delta$ $=0.025$, which is approximately the same magnitude as the estimated average effect for all three health indices in this paper. Qualitatively, the results of this illustration are the same at any sample size, but we set $N=100,000$ here because it makes the results easier to display. We set $p_{x}=0.5$.

In order to investigate our claim that distributional effects may be better powered when not everyone is affected homogeneously by the treatment, we consider the cases $p_{\tau} \in\{0.25,0.5$, $0.75,1\}$. We evaluate the power for the continuous and binary case for a dense range of values of $p_{t} \in(0,1)$. The results of these calculations are found in Appendix Figure 1 below.


## Discussion

Panel (a) corresponds to a setting when every individual is affected by the treatment homogeneously. Unsurprisingly, the power of the average effect estimate is greater than the power of the distributional effect estimate for all values of $p_{t}$.

Panel (b) corresponds to a setting where only $25 \%$ of individuals are unaffected by the policy. Note that for values of $p_{t}$ less than 0.25 , the distributional effect has very low power, a result of individuals in that part of the distribution being unaffected by the treatment. Nevertheless, over a large interval of values for $p_{t}$, the distributional test is slightly better powered than the test on the average. The power of the distributional test quickly decays outside of this range however, suggesting that a test that considers all values of $p_{t}$ in the range 0.5 to 1 (e.g., the AndersonDarling test used in this paper) may not be better powered than a test on the average.

In panel (c), however, where $50 \%$ of individuals are unaffected, the difference in power becomes substantial over nearly the whole range of values in the upper half of the distribution. This is particularly relevant to our setting because, observing the pre- and post-reform CDF of body size index, it appears that the ROSLA affected only for those in the upper half of the distribution.

The patterns in paned (d), with $75 \%$ of the population unaffected, are similar to those of the first three panels. For the values of $p_{t}$ corresponding to affected individuals, the difference in power
between the distributional and average effect is even larger, though the range of individuals who are affected is much narrower than in the other panels.

This discussion shows how tests of distributional effects may be better powered than tests for average effects in certain cases. Obviously, this framework is simplified in order to make the math tractable, but the general principle will hold that when only a portion of the distribution is affected by some treatment, methods that focus on those segments of the distribution may be better powered than those that consider the whole distribution. This appears to be increasingly true as the fraction affected becomes smaller.

Appendix G

We conducted a back-of-envelope calculation to estimate the mortality consequences of the estimated reduction in BMI caused by staying in school until age 16. To map BMI into mortality, we used Aune et al. (2016)'s estimates of the association between BMI and all cause mortality. In particular, we used the estimates from the column "All participants" of Table 2.

In the first step, we fitted a fractional polynomial of second order through the points in Table 2 in order to obtain a continuous mapping of relative mortality as a function of BMI. A fractional polynomial is a polynomial that may include logarithms, noninteger powers, and repeated powers. Here is the Stata output from this estimation:

```
. fp <BMI>: regress RR <BMI>
(fitting 44 models)
(... 10%.... 20%. . . 30%. . . 40% . . 50%. . . 60%. . . . 70%. . . 80%. . . 90% . . . 100%)
Fractional polynomial comparisons:
```

| BMI | df | Deviance | Res. s.d. | Dev. dif. | $P(*)$ | Powers |
| ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| omitted | 0 | 22.177 | 0.500 | 76.310 | 0.000 |  |
| linear | 1 | 20.454 | 0.490 | 74.588 | 0.000 | 1 |
| m = 1 | 2 | 16.741 | 0.436 | 70.875 | 0.000 | 3 |
| m $=2$ | 4 | -54.133 | 0.049 | 0.000 | -- | -23 |

(*) $\mathrm{P}=$ sig. level of model with $\mathrm{m}=2$ based on F with 11 denominator dof.

| Source | ss | df | MS | Number of obs$F(2,13)$ |  | 16 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Model | 3.71445463 | 2 | 1.85722731 | Prob > F |  | 0.0000 |
| Residual | . 031789143 | 13 | . 002445319 | R-squared |  | 0.9915 |
|  |  |  |  | Adj R-squared |  | 0.9902 |
| Total | 3.74624377 | 15 | . 249749585 | Root MSE |  | . 04945 |
| RR | Coef. | Std. Err. | t | $P>\|t\|$ | [95\% Con | Interval] |
| BMI_1 | 540.4458 | 16.46243 | 32.83 | 0.000 | 504.8809 | 576.0107 |
| BMI_2 | . 0000281 | $7.26 \mathrm{e}-07$ | 38.70 | 0.000 | . 0000265 | . 0000297 |
| _cons | -. 3610887 | . 0493578 | -7.32 | 0.000 | -. 4677197 | -. 2544577 |

where $R R$ is the relative mortality and $B M I_{-} 1=B M I^{-2}$ and $B M I_{-} 2=B M I^{3}$.
Let $\widehat{R R}(x)$ be the predicted relative risk of death for an individual with a BMI of $x$.
In the second step, we estimated the vintiles of the distribution of BMI among the compliers born between September 1, 1956 and August 31, 1957, where compliers are defined as those born before the reform who dropped out of school at age 15 and younger. These vintiles define the grid used to estimate the pre- and the post-reform cumulative distribution functions of BMI.

In the third step, we estimated $F_{\text {pre }}(\tau)$ and $F_{\text {post }}(\tau)$ at each vintile $\tau$ (following the procedure explained in Section 3 of the paper).

In the fourth step estimated, we estimated the pre- and post-reform relative risk of death:

$$
\begin{aligned}
& \widehat{R l s k}_{\text {pre }}=\widehat{R R}\left(\tau_{5}\right) * \widehat{F}_{\text {pre }}\left(\tau_{5}\right)+\sum_{k=10,15, \ldots, 90} \widehat{R R}\left(\tau_{k}\right) *\left[\widehat{F}_{\text {pre }}\left(\tau_{k}\right)-\widehat{F}_{\text {pre }}\left(\tau_{k-5}\right)\right]+ \\
&\left.+\widehat{R R}\left(\tau_{95}\right) *\left(1-\tau_{90}\right)\right) \\
& \widehat{R l s k}_{\text {post }}=\widehat{R R}\left(\tau_{5}\right) * \widehat{F}_{\text {post }}\left(\tau_{5}\right)+\sum_{k=10,15, \ldots, 90} \widehat{R R}\left(\tau_{k}\right) *\left[\widehat{F}_{\text {post }}\left(\tau_{k}\right)-\widehat{F}_{\text {post }}\left(\tau_{k-5}\right)\right]+ \\
&\left.+\widehat{R R}\left(\tau_{95}\right) *\left(1-\tau_{90}\right)\right)
\end{aligned}
$$

$\left[\left(\widetilde{R L S k}_{\text {post }}\right)-1\right] * 100$ is the estimated reduction in mortality associated with our distributional treatment effects.

To calculate the reduction in mortality implied by the average treatment effect, we first estimated the average BMI for compliers born at September 1, 1957. To do that, we restricted the sample to participants born before September 1, 1957 who dropped out of school at age 15 or younger and run a regression of BMI on quadratic trends for date of birth. The coefficient on the constant, 28.51707, is our estimate of the pre-reform average BMI among compliers. Next, we estimated the 2SLS effect of staying in school until age on BMI, -.4066152. The post-reform average BMI among compliers is equal to $28.1104548=28.51707-.4066152$. Finally, the estimated reduction in mortality is $\left[\left(\frac{\overparen{R R}(28.1104548)}{\overparen{R R}(28.51707)}\right)-1\right] * 100$

## REFERENCE

Aune, D., Sen, A., Prasad, M., Norat, T., Janszky, I., Tonstad, S., Romundstad, P. and Vatten, L.J., 2016. BMI and all cause mortality: systematic review and non-linear dose-response meta-analysis of 230 cohort studies with 3.74 million deaths among 30.3 million participants. Bmj, 353, p.i2156.

Appendix H

## Appendix Figure H1: Comparison of Distribution of BMI of Compliers

 in the UK Biobank and in the Health Survey for England

[^4]Appendix Figure H2: Comparison of Distribution of Waist-Hip Ratio of Compliers in the UK Biobank and in the Health Survey for England


Notes: The figure compares the distribution of waist-hip ratio of compliers in the UK Biobank and in the Health Survey for England (years 2006, 2007, 2008, 2009, and 2010). The UK Biobank sample is restricted to respondents living in England. We do not have data on the date of birth of HSE respondents so we had to rely on age and month of interview to identify respondents who were born unambiguously before September 1957. We applied the same sample restrictions in terms of age (and month of interview) to both samples. We approximate the population of compliers as those born before September 1957 who dropped out of school at age 15 or younger. The HSE estimates include sample weights. The distributions are adjusted for differences in gender and age (using the HSE as reference). $N=39,249$ (UKB), 2,259 (HSE).

## Appendix Figure H3: Comparison of Distribution of Diastolic Blood Pressure of Compliers in the UK Biobank and in the Health Survey for England



[^5]Appendix Figure H4: Comparison of Distribution of Systolic Blood Pressure of Compliers in the UK Biobank and in the Health Survey for England


[^6]We used genetic data to re-weight the UK Biobank (UKB) sample in an attempt to make it nationally representative. Genetic data may be useful in this regard because it is fixed at conception. We used the English Longitudinal Survey of Ageing (ELSA) as a nationally representative benchmark. We restricted both samples to individuals of European ancestry in each data set who were born in the years 1954-1959.

We use a polygenic score (PGS) for educational attainment (EA) to compare the two samples. A PGS is a weighted sum of molecular genetic markers called single-nucleotide polymorphisms (SNPs). The weights for this PGS are based on a genome-wide association study (GWAS) of EA (Lee et al. 2018) and are derived using a standard procedure, LDpred (Vilhjálmsson et al. 2015). A PGS for EA constructed in this way reduces dimensionality of the genetic data while still maximizing its predictive power. This PGS has been shown to explain $11-13 \%$ of the variation in EA, slightly less than the variation explained by a parent's EA (Lee et al. 2018).

To compare the distribution of the EA PGS between the ELSA and UKB samples, it is important to construct the polygenic score from the same set of SNPs. The ELSA sample contains imputed genotypes for $39,131,557$ SNPs. The UKB sample contains $16,642,384$ imputed SNPs. In order to include only those SNPs that are well-imputed, we restrict SNPs in the ELSA sample with Rsq values of exactly 1, where Rsp is a measure of imputation quality between 0 and 1. Due to computational constraints, we were unable to include an imputation quality criterion on the UKB sample. Keeping only SNPs that are in the intersection of the UKB sample, ELSA sample, and the GWAS summary statistics, we are left with 281,700 SNPs. The PGS is constructed based on this set of SNPs.

Appendix Figure H5 shows the distribution of the polygenic score in the two samples.

## Appendix Figure H5: Distribution of Polygenic Scores for Educational Achievement in the UK Biobank (UKB) and in the English Longitudinal Survey of Ageing (ELSA)



Notes: PGSs are standardized using the mean and standard deviation of PGSs in the ELSA sample.

To create weights, the distribution of scores is divided into 512 evenly-spaced points. At each point, the weight equals the ratio of the density of UKB observations to the density of ELSA observations. Then, for each individual in the UKB, we find the score distribution point nearest to the individual's score, and assign the weight associated with this point as the individual's weight.

Appendix Table H1 shows the results with and without weighting.


Notes: The table shows the effects of the school reform on education. Each cell corresponds to a separate regression. We report the coefficient on the indicator variable for being born on or after September 1, 1957 (i.e., "Post"). The dependent variable mean in the bottom row is the weighted mean among those born in the 12 months before September 1, 1957. Controls include male, age in days and age squared, dummies for calendar month of birth, dummies for ethnicity, and dummies for country of birth. Robust standard errors. $N=264,066$. The estimates in the first and third columns are slightly different from the estimates in Table 1 in the paper because the sample here is restricted to those of European ancestry in the UK Biobank who were genotyped.

Vilhjálmsson, B.J., J. Yang, H.K. Finucane, A. Gusev, S. Lindström, S. Ripke, ... \& T. Hayeck. (2015). "Modeling linkage disequilibrium increases accuracy of polygenic risk scores." The American Journal of Human Genetics, 97(4), 576-592.

Lee, James J., Robbee Wedow, Aysu Okbay, Edward Kong, Omeed Maghzian, Meghan Zacher, Tuan Anh Nguyen-Viet, Peter Bowers, ..., David I. Laibson, Jian Yang, Magnus Johannesson, Philipp D. Koellinger, Patrick Turley, Peter M. Visscher, Daniel J. Benjamin, and David Cesarini (2018). "Gene discovery and polygenic prediction from a genome-wide association study of educational attainment in 1.1 million individuals." Nature Genetics, 50, 1112-1121.


[^0]:    Notes: The figure shows the pre- and post-reform CDFs of north coordinate of place of birth. The pre-reform CDF is the CDF in the limit when date of birth is converging to September 1, 1957 from the left. The post-reform CDF is the CDF in the limit when date of birth is converging to September 1, 1957 from the right. $N=266,883$.

[^1]:    Notes: The figure shows the fraction of study participants who were overweight by quarter of birth. Overweight is defined as having a BMI greater or equal to 25 . The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=$ 270,019.

[^2]:    Notes: The figure shows the fraction of study participants for whom data on body fat percentage was missing by quarter of birth. The dashed vertical line marks the first birth cohort affected by the 1972 school-leaving age reform. Cohorts born to the right of the line had to stay in school until age 16 while cohorts born before could leave at age 15 . The curves show quadratic polynomials in quarter of birth that capture birth cohort trends. The circumference of each circle reflects the number of participants born in that quarter. $N=271,082$.

[^3]:    Notes: The table investigates whether the 1972 school-leaving age reform affected the fraction of study participants who stayed in school until age 17. Each cell corresponds to a separate regression of an indicator variable for whether the study participant stayed in school until (at least) age 17 on an indicator variable for whether the study participant was born on or after September 1, 1957 (i.e., "Post"), and quadratic or linear trends in date of birth. The set of controls include gender, age in days (at the time of the baseline assessment) and age squared, dummies for ethnicity, dummies for country of birth, and dummies for calendar month of birth.

[^4]:    Notes: The figure compares the distribution of body mass index (BMI) of compliers in the UK Biobank and in the Health Survey for England (years 2006, 2007, 2008, 2009, and 2010). The UK Biobank sample is restricted to respondents living in England. We do not have data on the date of birth of HSE respondents so we had to rely on age and month of interview to identify respondents who were born unambiguously before September 1957. We applied the same sample restrictions in terms of age (and month of interview) to both samples. We approximate the population of compliers as those born before September 1957 who dropped out of school at age 15 or younger. The HSE estimates include sample weights. The distributions are adjusted for differences in gender and age (using the HSE as reference). $N=39,186$ (UKB), 2,700 (HSE)

[^5]:    Notes: The figure compares the distribution of diastolic blood pressure of compliers in the UK Biobank and in the Health Survey for England (years 2006, 2007, 2008, 2009, and 2010). The UK Biobank sample is restricted to respondents living in England. We do not have data on the date of birth of HSE respondents so we had to rely on age and month of interview to identify respondents who were born unambiguously before September 1957. We applied the same sample restrictions in terms of age (and month of interview) to both samples. We approximate the population of compliers as those born before September 1957 who dropped out of school at age 15 or younger. The HSE estimates include sample weights. The distributions are adjusted for differences in gender and age (using the HSE as reference). $N=39,316$ (UKB), 1,899 (HSE).

[^6]:    Notes: The figure compares the distribution of systolic blood pressure of compliers in the UK Biobank and in the Health Survey for England (years 2006, 2007, 2008, 2009, and 2010). The UK Biobank sample is restricted to respondents living in England. We do not have data on the date of birth of HSE respondents so we had to rely on age and month of interview to identify respondents who were born unambiguously before September 1957. We applied the same sample restrictions in terms of age (and month of interview) to both samples. We approximate the population of compliers as those born before September 1957 who dropped out of school at age 15 or younger. The HSE estimates include sample weights. The distributions are adjusted for differences in gender and age (using the HSE as reference). $N=39,316$ (UKB), 1,899 (HSE).

