Small Family, Smart Family?
Family Size and the IQ Scores of Young Men*

by

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Abstract

How do families influence the ability of children? Cognitive skills have been shown to be a strong predictor of educational attainment and future labor market success; as a result, understanding the determinants of cognitive skills can lead to a better understanding of children’s long run outcomes. This paper uses a large dataset on the male population of Norway and focuses on one family characteristic: the effect of family size on IQ. Because of the endogeneity of family size, we instrument for family size using twin births and sex composition. IV estimates using sex composition as an instrument show no negative effect of family size; however, IV estimates using twins imply that family size has a negative effect on IQ. Our results suggest that effect of family size depends on the type of family size intervention. We conclude that there are no important negative effects of expected increases in family size on IQ but that unexpected shocks to family size resulting from twin births have negative effects on the IQ of existing children.

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Cognitive skills have been shown to be a strong predictor of educational attainment and future labor market success; as a result, understanding the determinants of cognitive skills can lead to a better understanding of children’s long run outcomes. But how do families influence the cognitive ability of their children? Researchers across a variety of fields have attempted to understand the determinants of IQ as a measure of cognitive ability with only limited success. In this paper, we focus on the role of one family characteristic: the effect of family size on IQ test scores.

Family size has long been of interest to researchers, particularly because of the strong empirical regularity that children from larger families tend to have poorer outcomes. There is an extensive theoretical literature on the tradeoff between child quantity and quality within a family that dates back to Becker (1960) and Becker and Lewis (1973). The theory is often cited and is used as the basis for many macro growth models (see Becker and Barro (1988) and Doepke (2003)). A key element of the quantity-quality model is an interaction between quantity and quality in the budget constraint that leads to rising marginal costs of quality with respect to family size; this generates a tradeoff between quality and quantity. But is this tradeoff real? In particular, is it true that having a larger family has a causal effect on the “quality” (in our case IQ) of the children? Or is it the case that families who choose to have more children are (inherently) different, and the children would have lower IQs regardless of family size?

This paper uses a dataset on the male population of Norway to examine the effect of family size on children’s IQ, an outcome not previously available in datasets of this size. Importantly, we also address the issue of the endogeneity of family size. Until recent years, the empirical literature on the effects of family size on child outcomes generally relied on OLS

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1 Rosenzweig and Wolpin (1980) explicitly derive the assumptions under which an exogenous increase in family size should have a negative effect on child quality.
estimation and found a negative relationship between family size and child “quality” (usually education), even after controlling for socio-economic factors.\(^2\) However, few of these findings can be interpreted as causal; family size is endogenously chosen by parents and hence may be related to other, unobservable parental characteristics that affect child outcomes.\(^3\) Also, the absence of information on birth order often means that birth order effects are confounded with family size effects. We are unaware of any studies of IQ that have attempted to deal with both of these issues.

There is, however, a literature on the causal effects of family size on child educational attainment, starting with Rosenzweig and Wolpin (1980). Recent papers in this literature use arguably exogenous variation in family size induced by the sex composition of children (parents have a preference for variety and so are more likely to have an additional child if the first two are the same sex) and/or the birth of twins (resulting in a family size increase of two when one was expected) and have generated very little evidence for any quantity-quality tradeoff (Conley and Glauber 2006; Caceres-Delpiano 2006; Angrist et al. 2006). In recent work (Black, Devereux and Salvanes 2005), we use Norwegian data on cohorts born between 1912 and 1975 and variation in family size induced by the birth of twins to examine the effects of family size on education and earnings. We find little evidence for any family size effect either in the OLS or IV estimates. In this paper, we can study a more recent set of cohorts -- all individuals in our sample were born between 1967 and 1987 -- and are able to introduce previously unavailable information on IQ and birth endowments including birth weight.

Given the two types of interventions the instruments represent—one (sex composition) is a planned increase in family size based on parental preferences for variety in the sex composition

\(^2\) See Blake (1989) and the numerous studies cited therein.
of their children and the other (twin births) is an unplanned shock to family size resulting in two generally lower-birth weight children with zero spacing--it is somewhat surprising that researchers so far have found similar effects of family size on child outcomes using the twins and sex composition instruments (for example, Angrist et al. 2006). In contrast, our estimates differ depending on estimation method. OLS estimates using a rich set of controls suggest that there is no strong relationship between family size and IQ. Likewise, IV estimates using sex composition as an instrument show no negative effect of family size. However, our IV estimates using twins imply that family size has a negative effect on IQ; unexpected shocks to family size resulting from twin births have negative effects on the IQ of existing children.

The question then becomes why we observe a negative effect of increases in family size generated by the birth of twins on IQ when we found little or no effect on education and earnings in our earlier paper. When we examine this issue, we find that this difference is largely due to the different cohorts studied; when education is studied on a more recent cohort, we find family size effects similar to those of IQ.

Unlike the papers discussed above, we also have information on birth weight for all individuals in our sample. We use this information to address the recent critique of twin studies by Rosenzweig and Zhang (2006), who suggest that much of this literature is biased away from finding negative effects of larger families on child outcomes due to endowment-reinforcing parental investments in children. Interestingly, we find that taking account of endowments actually reduces the negative effects of family size that we find using our basic 2SLS specification, a result that is consistent with compensating rather than reinforcing parental investments.

3 There is also a literature examining the effect of family size on parental outcomes. See, for example, Bronars and Grogger (1994) and Angrist and Evans (1998).
The paper unfolds as follows. Section 2 describes the data we use and Section 3 discusses our empirical strategy. Section 4 presents our results and Section 5 discusses the internal and external validity of our estimates as well as potential mechanisms. Section 6 compares the estimates with previous estimates from Norway and Section 7 concludes.

### 2. Data

Our primary data source is the birth records for all Norwegians born in the period 1967 to 1998 obtained from the Medical Birth Registry of Norway. All births, including those born outside of a hospital, are included as long as the gestation period was at least 16 weeks. The birth records contain information on year and month of birth, birth weight, gestational length, age of mother, and a range of variables describing infant health at birth. In these data, we are also able to identify twin births. We can measure family size by counting the number of births to each woman. Our sample is composed of families in which the first birth took place during or after 1967. Given that we observe year of birth, we are able to construct family size and indicators for the birth order of each child directly from the births register.

Using unique personal identifiers, we match these files to the Norwegian Registry Data, a linked administrative dataset that covers the population of Norwegians aged 16-74 in the 1986-2002 period and is a collection of different administrative registers such as the education register, family register, and the tax and earnings register. These data are maintained by Statistics Norway and provide information about educational attainment, labor market status, earnings, and a set of demographic variables (age, gender) as well as information on families. 

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4 The data also include stillbirths, which constitute approximately 15 per 1000 births. We exclude these from the sample.

5 See Møen, Salvanes and Sørensen [2004] for a description of these data.
The IQ data are taken from the Norwegian military records from 1984 to 2005. In Norway, military service is compulsory for every able young man. Before entering the service, their medical and psychological suitability is assessed; this occurs for the great majority between their eighteenth and twentieth birthday. IQ at these ages is particularly interesting as it is about the time of entry to the labor market or to higher education.

The IQ measure is a composite score from three speeded IQ tests -- arithmetic, word similarities, and figures (see Sundet et al. [2004, 2005] and Thrane [1977] for details). The arithmetic test is quite similar to the arithmetic test in the Wechsler Adult Intelligence Scale (WAIS) [Sundet et al. 2005; Cronbach 1964], the word test is similar to the vocabulary test in WAIS, and the figures test is similar to the Raven Progressive Matrix test [Cronbach 1964]. The composite IQ test score is an unweighted mean of the three subtests. The IQ score is reported in stanine (Standard Nine) units, a method of standardizing raw scores into a nine point standard scale with a normal distribution, a mean of 5, and a standard deviation of 2.6

We have IQ scores on about 84% of the relevant population of men in Norway.7 One nice feature of our data is that we can construct family size and both instrumental variables for all persons, including those with missing IQ scores. To examine whether our missing data appears to be systematically related to our variables of interest, we have regressed a dummy variable for whether IQ is missing on family size using both OLS and 2SLS. Our finding is that individuals from larger families are more likely to have missing scores (for example, first-borns from 2-child families are 4% less likely to have missing scores than first-borns from 5-child families), but that

6 The correlation between this IQ measure and the WAIS IQ has been found to be .73 (Sundet et al., 1988).
7 Eide et al (2005) examine patterns of missing IQ data for the men in the 1967-1987 cohorts. Of those, 1.2 percent died before 1 year and 0.9 percent died between 1 year of age and registering with the military at about age 18. About 1 percent of the sample of eligible men had emigrated before age 18, and 1.4 percent of the men were exempted because they were permanently disabled. An additional 6.2 percent are missing for a variety of reasons including foreign citizenship and missing observations. There are also some missing IQ scores for individuals who showed up to the military.
this relationship disappears when family size is instrumented with the twin or sex composition instruments. Thus it is unlikely that missing IQ data is causing serious bias to our later estimates.

We exclude families in which no child is born before 1987 as no data on military outcomes of children is available for such families. We use twins to construct our instrumental variables but, as is standard in the literature, we drop twins from our estimating samples. Table 1 presents summary statistics and Appendix Table 1 shows the distribution of family sizes in our sample.

3. Empirical Strategy

Twins as an Instrument for Family Size

Rosenzweig and Wolpin (1980) first discuss the idea of using twin births as unplanned and therefore exogenous variation in family size. In their model, parents have an optimal number of children N*. The birth of twins can vary the actual N from the desired N* and it is this arguably exogenous variation that is used to estimate the effects of family size on child outcomes. Our general estimation strategy is as follows:

\[ IQ = \beta_0 + \beta_1 \text{FAMSIZE} + X\beta_2 + \varepsilon \] (1)

\[ \text{FAMSIZE} = \alpha_0 + \alpha_1 \text{TWIN} + X\alpha_2 + \nu \] (2)

In this case, IQ is the IQ of the child and FAMSIZE is the total number of children in the family. Equation (2) represents the first stage of the two stage least squares estimation, where equation (1) is the second stage. The controls in \( X \) include indicator variables for cohort, age at test, age of mother at birth, age of mother at first birth, years of education of mother, years of education of

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8 The 2SLS estimates are all small and statistically insignificant and vary in sign.
father, and log birth weight of the child. We also include controls for the IQ of the father for cases for which this is available.\(^9\)

The TWIN indicator is equal to 1 if the \(n^{th}\) birth is a multiple birth, and equal to 0 if the \(n^{th}\) birth is a singleton. We restrict the sample to families with at least \(n\) births and study the outcomes of children born before the \(n^{th}\) birth. In practice, we estimate the specification for values of \(n\) between 2 and 4. By restricting the sample to families with at least \(n\) births, we make sure that, on average, preferences over family size are the same in the families with twins at the \(n^{th}\) birth and those with singleton births. Also, by restricting the sample to children born before birth \(n\), we avoid selection problems that arise because families who choose to have another child after a twin birth may differ from families who choose to have another child after a singleton birth.

Our methodology avoids two possible pitfalls that can arise when twins are used as instruments. First, by focusing on whether a twin birth occurs for the \(n^{th}\) birth, we avoid the problem that families that have more births are more likely to have at least one twin birth. Second, by only using children born before a possible twin birth, we are able to avoid the problem that a twin birth both increases family size and shifts downwards the birth order of children born after the twins. For example, if a twin birth occurs at the second birth, the next child born is now the fourth born-child rather than the third. Thus, any estimates using children born subsequent to a twin birth will confound family size effects with birth order effects.

**Sex Composition as an Instrument for Family Size**

\(^9\) We do this by setting father’s IQ to zero for cases in which it is missing and then interacting father’s IQ with a dummy variable which is one if we have information on father’s IQ and zero otherwise. Both these variables are then included in the regression. Father’s IQ is typically unavailable for men whose father was born before 1950.
It has been well documented that parents have preferences for variety and so are more likely to have a child at parity n if the previous n-1 children are all of the same sex (Angrist and Evans, 1998). As a result, if the sex composition is random, families that randomly had two same-sex children are more likely to have a third, all other things equal, making sex composition a potential instrument. We utilize this instrument for n=3 and n=4, and restrict the sample to families with at least n-1 children.

Sample of families with at least 2 children (n=3)

When n=3, we are studying families with at least two children and the instrument is whether or not the first two children are the same sex. Given that we only have IQ data on boys, the relevant possibilities for the first two children are BB, BG, and GB, where G is for girl and B is for boy. (Note that we are looking at the outcomes of the first and second children; if the first two children are GG, we will not have IQ information on any of the existing children.). One obvious comparison is to compare the IQ of the first born children in a sample composed of BB and BG families – here the instrument can be interpreted simply as whether the second child is a boy. A second comparison is to compare the IQ of second born children in a sample composed of BB and GB families – here the instrument can be interpreted as whether the first child is a boy.

We improve the precision of our estimates by using both comparisons in estimation. To do this, we estimate the following equation using the first two children:

\[ FAMSIZE = \alpha_0 + \alpha_1 ALLBOYS + \alpha_2 2nd + X \alpha_3 + \nu \]  

(3)

Here \( ALLBOYS \) is an indicator for the first two children being the same-sex, and \( 2nd \) is an indicator for whether the child is the 2\textsuperscript{nd} born. The controls included in \( X \) are the same ones included when the twins instruments are used.
It is important to note that this approach is invalid if there are preferences for boys or girls in addition to preferences for variety, as then the probability of a second child (and hence selection into the at least 2 children sample) will depend on the sex of the first child. The evidence suggests no such sex preferences in Norway.\textsuperscript{10}

**Sample of families with at least 3 children (n=4)**

In the case of families with at least three children, our instrument equals one if all of the first three children were boys. We compare the boys in these families to the boys in families where the first two births were either both boys or both girls. In this situation, parents who have a preference for variety either got variety in the third birth (having same sex births prior), at which point they stopped having children, or had another child of the same sex, thereby increasing the probability that they would have another child in order to achieve diversity.

Thus, the only relevant sequences are BBB, BBG, and GGB; no other comparisons are useful because sequences starting with BG or GB are not comparable to those starting with GG or BB. This is because the probability of having a third child differs systematically depending on whether or not the first two children are the same sex (as is exploited in the n=3 analysis).

Analogous to the n=3 case, a simple approach is to compare the outcomes of the first two children in BBB families to those in BBG families. Also, given there are no preferences for boys (or girls), an additional comparison is between the outcomes of third-born children in BBB and GGB families. In practice we implement both using families with at least 3 children whose first two children are the same-sex. The following regression is estimated for the first three children:

\[
FAMS\text{SIZE} = \alpha_0 + \alpha_1ALLBOYS + \alpha_22nd + \alpha_33rd + X\alpha_4 + \nu
\]  \hfill (4)

\textsuperscript{10}There is no statistically significant relationship between the probability of having a second child and the sex of the first born child despite the enormous sample size of 481,731 families.
Here *ALLBOYS* is an indicator for the first three children being the same-sex, *2nd* is an indicator for whether the child is the 2nd born, and *3rd* is an indicator for whether the child is third born.

### 4. Family Size Results for IQ

In Table 2, we show the mean IQ and the distribution of IQ scores by family size. There are two clear patterns. First, only children have lower scores than the average child in 2 child families. Second, from family sizes of 2 to 5, we see a monotonic relationship that greater family size accompanies lower average scores.

The unconditional relationship between family size and IQ is only suggestive; for example, it could simply represent cohort effects or birth order effects. To better understand the relationship, we run regressions by birth order and regress IQ of children on family size dummies, presenting estimates from regressions both with and without additional controls to test sensitivity. As described in Section 3, these controls include indicator variables for cohort, age at test, age of mother at birth, age of mother at first birth, years of education of mother, years of education of father, father’s IQ, and log birth weight of the child.

The results are presented in Table 3. The omitted category in each case is Family with 5 or more children. The estimates suggest that the highest achieving children come from 2 and 3 child families, with lower achievement for only children and children from families with four or more.

However, the results are not robust to introducing additional control variables. With the control variables, almost nothing is statistically significant and the estimates are much smaller.

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11 Birth order is highly correlated with family size and has been shown to be strongly related to child outcomes in Norway (Black, Devereux, Salvanes 2005, 2007b; Bjerkedal et al. 2007).
than before. There still appears to be a penalty to being an only child, but there is little indication
that family size matters otherwise. The fact that adding family background variables and log(birth
weight) as controls largely eliminates the family size effect suggests that the family size results
may be driven by unobserved family background characteristics; if unobserved factors impacting
IQ have similar correlations with family size as the observed factors, the true causal effect of
family size is not very negative.

The IV estimates below take a more direct approach to the issue of causality.

Results Using Twins as an Instrument for Family Size

The 2SLS estimates are presented in Table 4. We also include, for comparison, two sets
of OLS coefficients estimated on exactly the same samples as the 2SLS estimates; the first set
includes controls only for birth order, the second set includes all control variables. As was the
case in Table 3, the OLS estimates without controls suggest negative effects of increased family
size, but those with controls never provide any evidence for negative family size effects. Instead
they provide very precisely estimated family size coefficients that are very close to zero.

The first stage coefficients using the twin instruments are very strong and suggest that a
twin birth increases completed family size by about 0.7 to 0.9. These first stage effects are a little
larger than we found in our 2005 paper, presumably because family sizes are smaller for these
more recent cohorts. As expected, twins at higher parity have a larger effect on family size,
presumably because they are more likely to push families above N* (the optimal number of
children). The t statistics from the first stage are typically around 40, indicating that there are no
concerns about weak instruments in this application.
The 2SLS estimate of the effect on the first child of changes in family size induced by the second birth being a twin birth is -0.15 (0.05). This implies an adverse effect of increased family size on IQ – each additional child reduces IQ of the first child by about 15% of a stanine or 8% of a standard deviation. The equivalent estimate for families that have at least 3 births is -0.17 (0.05), and for families that have at least 4 births it is -0.12 (0.08). Taken together, these three estimates are all more negative than the OLS estimates (the first two are statistically different from OLS) and suggest fairly sizeable negative effects of family size on IQ of about 15 percent of a stanine for each additional child (this is about 1.2 IQ points on a standard IQ scale). This translates into approximately a 1% difference in annual earnings as an adult. Unexpected random shocks to family size, induced by the birth of twins, lead to poorer IQ outcomes for existing children.

Results Using Same-Sex as an Instrument for Family Size

For the case where the instrument is an indicator of whether the first two children are the same sex, the first stage estimate on \texttt{ALLBOYS}, also reported in Table 4, is 0.081 with a standard error of 0.003. Similarly, Angrist et al. (2006) find a coefficient of .073 (.010) using Israeli data and Conley and Glauber (2006) find that having two same-sex children increases the probability of having a third child by about .07 in the U.S.

For the case where the instrument is whether the first three children are the same sex, the first stage estimate on \texttt{ALLBOYS} is 0.070 with a standard error of 0.005. While not directly

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12 The first stage coefficients are much bigger than those of Angrist et al. (2006) using Israeli data. For twins at second birth, they found a first stage coefficient of only 0.44. This presumably reflects the much larger proportion of large families in the Israeli data.

13 We arrive at this number by regressing log earnings in 2002 on IQ scores for individuals aged at least 30 in 2002. For both cross-sectional and family fixed effects specifications we obtain precisely estimated coefficients on IQ of about .06, suggesting an extra stanine in IQ (about half a standard deviation) leads to about 6% higher earnings.
comparable, this is smaller than the Angrist et al. (2006) finding of .120 (.015) using the sample of families with 3 or more children in Israeli data.

In sharp contrast to the estimates using twins as an instrument for family size, the 2SLS estimates using the sex composition as instruments are both positive. For the sample of families with 2 or more births the estimate is .07 (.07), and for the sample of families with 3 or more births the estimate is .08 (.17). While these family size coefficients are not very precisely estimated, the first one does rule out large negative effects of family size on IQ and is statistically significantly different from the family size estimates using twins. The contrast with the twins estimates may result from the fact that, in this case, induced increases in family size are expected and result in normally spaced singletons.

5. Internal and External Validity of the Instruments

Internal Validity of Instruments

Randomness

In order for our IV estimates to be consistent, it must be that the instrument is uncorrelated with the error term in equation (1). One concern is that the occurrence of a twin birth may not be random and may be related to unobservable family background characteristics. By definition, this is untestable, but we do examine whether the probability of twins is related to observed characteristics such as mother's and father's education and father's IQ. To do so, we estimate three linear probability models of the probability of a twin birth at the Nth birth (for N=2, 3, and 4), conditional on having at least N births, using the full set of control variables. Over these three regressions, of the nine coefficients on mother’s education, father’s education,
and father’s IQ, none is statistically significant.\textsuperscript{14} Given the enormous sample sizes, these results strongly suggest that twinning probabilities are not related to parental characteristics.

A concern might be that, in recent years, fertility treatments have led to an increase in the incidence of multiple births, and these multiple births could then be related to characteristics of the parents. However, this phenomenon did not take off in Norway until the late 1980s so, as can be seen in Figure 1, the probability of a multiple birth is fairly constant over time for the cohorts we are studying (persons born 1967 – 1987).

We use a similar approach to examine the randomness of the sex composition instruments. Once again, we find no statistically significant relationship between any of the family background characteristics and the instrument for either the sample of families with at least 2 children, or the sample of families with at least three children where the first two children are the same-sex.

In addition to examining the relationship between the instrument and the observable characteristics of the family, there are other tests we can conduct. For example, if the birth of twins is truly random, whether the second birth is a twin birth should have no impact on outcomes of the first child that occur before the second birth occurs. We do not have information on early childhood outcomes but we do know birth weight, so we use 2SLS to estimate the “effect” of family size on the log birth weight of existing children, using the full set of control variables used in the later regressions (obviously birth weight is excluded). If the instrument is random, we should see no effect of family size on the birth weight of existing children. For the twins instruments, we get an effect of family size on the birth weight of existing children of 0.013 (0.005) for twins at second birth, 0.007 (0.005) for twins at 3\textsuperscript{rd} birth, and -0.008 (0.008) for twins

\textsuperscript{14} Twin probabilities increase with parity and maternal age at birth (Jacobsen, Pearce, and Rosenbloom, 1999; Bronars and Grogger, 1994) but this is not a problem as we condition on these variables in the 2SLS estimation.
at fourth birth. Only the first coefficient is statistically significant and it is not very large (a multiple second birth is associated with the first child being about 1% heavier at birth). This probably reflects the fact that larger, healthier women are more likely to have successful twin births and suggests that, in the absence of a control for birth weight, we would have obtained a family size effect on IQ that was biased towards zero. A value added of our study is that we are able to control for birth weight in the analysis to deal with this possibility.

When we study the effect of the second child being a boy on the log birth weight of the first child, we get an estimate of 0.002 (0.009); the equivalent estimate for the second of our sex composition instruments is 0.036 (0.021). Neither of these is statistically significant. The overall conclusion from this exercise with birth weight is that, with the exception of second-birth twins, there is very little evidence that the instruments are not randomly assigned.

Other Threats to Internal Validity

The large first stage coefficients for the twins instrument (.7 - .9) are reassuring, as they imply that this can be thought of as an experiment with almost perfect compliance -- the birth of twins leads to family size increasing by one in almost all cases. This is important, as the major possible biases using twins are that we may be capturing the effects of having closely-spaced children, or of having low birth weight children, rather than the effects of having more children. Given that twin births add extra children for almost all families, these possibilities are best seen as threats to the external validity of the twins approach rather than as sources of bias. Later in this section, we consider the external validity issues in more depth.

Because the first stage coefficients are much lower for the sex composition instruments, the possibility of bias due to direct effects of sex composition on child IQ is greater. Rosenzweig
and Wolpin (2000) conjecture that there may be beneficial effects of having a sibling of the same sex and this would tend to bias away from finding family size effects using the instrument. The literature is mixed on whether sex composition has direct effects on child outcomes (Dahl and Moretti, 2004; Butcher and Case, 1994; Conley, 2000; Deschenes, 2002 all find some evidence of sex-composition effects. Kaestner (1997) and Hauser and Kuo (1998) find no evidence of such effects). In our data, we find no evidence for sex composition effects on IQ when we study situations in which the sex composition has no correlation with family size (for example whether the 3rd child is a boy in families in which the first child is a boy and the second is a girl).

**External Validity**

While it may be reasonable to view both type of instruments as being valid, they constitute very different types of interventions to increase family size. Angrist et al. (2006) discuss the many differences between the twins and sex composition instruments, in particular focusing on how the two instruments affect different parts of the family size distribution and how the ethnic composition of individuals whose fertility responds to the twins instrument differs from that of individuals who respond to the same-sex instrument. In our case these are not issues, as there is no significant ethnic variation in our sample, and the two instruments affect the family size distribution in similar ways. Therefore, we concentrate on three different aspects of the twins instrument that do differ from the same-sex one – in the case of twins, the additional family member is smaller (the difference in birth weight between twins and singletons averages about 1000 grams), there is no space between the births, and the average age of existing children at arrival of the marginal child is younger with twin births.

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15 However, the Butcher and Case study found no evidence that sex composition impacted the outcomes of boys.
Lower Birth Weight of Twins

Recent twins studies (including this one) have estimated the effect of an extra birth at parity N (due to the birth of twins) on the outcomes of children at parities 1 to N-1, for all women who have at least N births. Rosenzweig and Zhang (2006, henceforth RZ) criticize this approach by arguing that the endowments of twins are lower than those of singletons, and parents may allocate resources differentially across their children on the basis of their endowments. Thus, the effect of adding two twins (with lower birth weights on average) to a family may differ from the effect of adding two singletons. 16

As suggested by RZ, we have tried adding controls for the birth weight of the potential twin birth (using the actual birth weight of a singleton, and the mean birth weight of twins). Interestingly, adding the birth weight controls causes the negative reduced form effect of twin at second birth on first child IQ to disappear. This is counter to what RZ predict if there were reinforcing investments by parents. According to RZ, parents with low birth weight twins are more likely to invest in the heavier first born child, thereby offsetting the negative effect of increased family size on first children. Thus, once we control for birth weight, RZ would predict that the negative effect of increased family size (through twin births) should increase in absolute value. Our finding is instead consistent with compensatory investment behavior by parents.

The RZ approach can be thought of as an attempt to create a new experiment in which twins and singletons have the same average birth weight. Unfortunately, in the real world no such experiment is available and their method of conditioning on birth weight may invalidate the

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16 RZ argue that parental investment is likely to reinforce endowment differences and thus bias the literature away from finding negative effects of larger families. In addition, they provide some evidence of reinforcing parental investments using twins data from China. Datar et al. [2005] find evidence that is generally suggestive of reinforcing parental investments among singletons using U.S. data. However, much of the available evidence for developed countries suggests compensatory investment behavior (Behrman, Pollak, and Taubman 1989; Ashenfelter and Rouse, 1998; Ermish and Francesconi, 2000).
exogeneity of twin births. While twinning may be a random event that is orthogonal to family characteristics, birth weight is not. Therefore, conditioning on birth weight creates a correlation between having twins and unobserved family characteristics. For example, families that have twins may be very similar to families that have singletons, but a family that has a 2000 gram singleton is likely to be different in some ways than a family that has two 2000 gram twins (for example, the mother may be more likely to have smoked during pregnancy.) Thus, it can be seen as an attempt to increase external validity that comes at the cost of reduced internal validity i.e. potential bias. 17 Thus, while our finding here is suggestive that family size increases through twinning have particularly negative effects on existing children because of compensatory behavior by parents favoring twins, it cannot be conclusive.

**Birth Spacing**

In the case in which we use twins at second birth to study the outcomes of the first child, it may be that the first child is worse off if the second birth is a twin birth because he gets less attention if his siblings are more closely spaced. To examine this, we have looked at non-twin families and estimate the effect of the spacing between later children on earlier children’s outcomes. In particular, to simulate the timing of twins, we create a variable for whether the gap between the later children is less than or equal to one year. Also included are the standard set of control variables augmented by a control for family size and a quadratic function of the time gap between children born before the two children whose spacing is being investigated. It is

17 Recent papers have found that the impacts of birth weight on adult outcomes are generally similar in magnitude in the cross-section to when twin or sibling differences are used in estimation [Black et al. 2007, Oreopoulos et al. 2006]. To the extent that twin and sibling differences difference out confounding effects of other factors, this is consistent with the notion that birth weight is not highly correlated with these unobserved factors and hence the bias from the RZ approach may not be large.
The estimates are reported in Appendix Table 3. The reported coefficients are those relating to the distance in years between the two children immediately after the children listed in Column 1. For example, in the first box, the coefficient of -.26 suggests that the first child’s IQ is lower by .26 (about ¼ of a stanine) if the gap between the second and third child is less than or equal to one year (in families with at least 3 children).

We find that earlier children have lower IQs when later children are more closely spaced together -- children do better if the two children immediately after them are spaced further apart. If this result can be extrapolated to the case of twins, in which the space is zero, it implies that the effect of a subsequent case of twins is more adverse than the effects of having two subsequent singletons. The reduced form effect of twins on IQ of earlier born children in Table 4 is about -.1 for both twins at second and at third birth. The analogous magnitudes in Appendix Table 3 are about -.2, suggesting that the 2SLS estimates using twins could easily arise due to spacing effects and may not be generalizable to a situation in which one adds two singletons who are spaced apart.

**Age of Existing Children at Arrival of Marginal Birth**

A closely related issue is that, in the sample of families with 2 or more children, generally first-borns are older when a singleton 3rd child is born (7.9 years old on average) than when the 3rd-born child is part of a twin second birth (4.4 years old on average). It is plausible that increased family size has a more negative impact if it occurs earlier in the child’s life and this

---

18 One caveat here is that, unlike with twins, the spacing of singletons can be chosen by parents and families that closely space births are different in terms of observables than families that space out births.
could cause the twins 2SLS estimates to be more negative than OLS or the sex composition 2SLS estimates. There is some evidence that this factor can account for differences in the impacts of the family size interventions. When one uses the twin at second birth instrument using two different samples (1) where the first child is aged at least 4 when the potential twin birth occurs and (2) when the first child is aged less than 4 when the potential twin birth occurs, the 2SLS coefficients on family size are .02 (.07) and -.26 (.07) respectively. This is consistent with the idea that the exact timing of the addition of a marginal child may influence the effect on existing children and suggests that some of the differences between the twin and sex composition estimates could be because the extra child arrives "earlier" with twins.19

Mechanisms through Which Family Size Increases Affect Children

Our results point to an effect of family size that depends on exactly how family size is increased, with evidence of negative effects only from unexpected increases due to twin births. There are many possible reasons why expected increases in family size may not have deleterious consequences on the IQs of existing children. Younger siblings may actually play a beneficial role for children. For example, Zajonc (1976) suggests that older children develop intellectually through having younger children to teach. Also, parents may respond to the expense of an extra child by reducing expenditures on items that are not beneficial to child development such as alcohol or nights out. Additionally, Rosenzweig and Zhang (2006) suggest that there may be an increase in the quantity of household public goods such as internet connectivity when family size increases.

19 It seems likely that IQ scores are also impacted by investments after the first few years of life as it has been clearly demonstrated that scores on the AFQT test in the U.S. depend to a large extent on years of schooling of the test-taker (for example, by Neal and Johnson, 1996).
While we cannot test these hypotheses, there are two additional possibilities that we can shed some light on using our data. First, families with more children may be less likely to suffer from marital breakdown. We have examined this by estimating the effect of family size on the probability that both parents are living together 20 years after the first child is born (69% are). The only statistically significant IV estimate is -.032 (.011) using twins at 3rd birth. This provides some evidence that twins may destabilize families and another reason for the large negative effects of twins on child IQ.

Finally, mothers may be more likely to leave the labor market and stay home when family sizes are larger and this may be beneficial to the children. Consistent with earlier empirical work for the U.S. (Angrist and Evans, 1998), we find negative effects of family size on labor supply of mothers using all methodologies. When we study the effect of family size on the probability the mother works full time 15 years after the first birth, the twin at second birth estimate is -.044 (.012) and the sex composition estimate is -.103 (.025). Interestingly, as with Angrist and Evans, the twins estimate is smaller than the analogous sex composition one. To the extent that children benefit intellectually from having their mother at home, this suggests another reason why increases in family size due to twinning may have more negative effects than those due to sex composition.

6. Comparison to Previous Norwegian Estimates of Family Size Effects

Our estimates using the same-sex instrument are quite consistent with those we found when we looked at education as an outcome in our 2005 paper. However, the estimates using the
twin instruments are quite different from the tiny statistically insignificant family size effects we found for education using the twins instrument in that paper. The difference was quite perplexing, as we would expect an effect on IQ to lead to differences in education and earnings. However, upon further investigation, the differences appear largely due to our use of different cohorts in the two papers. When we do the education analysis on a sample as close as possible to our current sample (individuals aged at least 21 in 2002) and use twin instruments, we find sizeable negative IV effects of family size on the probability of high school graduation (see Appendix Table 2).

This discovery led us to consider differences in the effect of family size on IQ over time. In order to increase the number of cohorts in our IQ analysis, we can no longer use the birth records, as that starts in 1967. In our 2005 study, we did not have access to the birth records and so calculated family size primarily by counting individuals aged 16+ who are present in the population register in 2002. We also obtained information on twin births from the population registry. We now use this methodology to study the effect of family size on IQ for the 1950 to 1984 cohorts. Because the birth records include only Norwegian-borns, we exclude individuals not born in Norway from this sample to be consistent.

Using this longer time series data, we have constructed a 9-year moving average (by year of the potential twin birth) of the reduced form effect of twins at third birth on the IQ scores of 1st and 2nd born children. As can be seen in Figure 2, the effect of 3rd-birth twins changed around 1970 from being approximately zero and became quite negative during the 1970s. Clearly, there has been a true change in the effect of twinning as we move from older to more recent cohorts.

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22 The twin indicators are unavailable in the population register for all cohorts born after 1984. Given that IQ data are generally available for cohorts starting with those born in 1950, we restrict our data to the 1950-1984 cohorts.
23 We have found our estimates using this methodology are similar but less precisely estimated than those using the information from the birth records. The decline in precision presumably arises because of the greater accuracy of the birth records methodology.
24 We chose twins at 3rd birth because it gives the largest family size effect in Table 4.
We are not sure why this change occurred. One possibility is that falling twin mortality rates over this period may have meant that surviving twin pairs are weaker and hence have a more adverse effect on existing children. However, this does not seem to be an issue as, for the multiple births in our sample, birth weights generally increased during the decade of the 1970s. The differences between cohorts may also relate to the very different environment concerning contraception and childcare for later cohorts relative to earlier ones. Another plausible explanation is that increasing labor force participation by mothers over this period has increased the cost of unexpected multiple births, and leads to adverse child outcomes. Finally, given our finding that twin births destabilize parental relationships, the increased incidence of family breakdown over time may have led to a greater negative impact of twin births on child outcomes.

7. Conclusions

Understanding the determinants of a child’s cognitive ability is the first step to understanding their longer run outcomes. In this paper, we examine the role of family size on an individual’s IQ, attempting to isolate the causal relationship between the two. Our results point to an impact of family size that depends on exactly how family size is increased. OLS estimates with controls for family background suggest that there is no strong relationship between family size and IQ. Likewise, IV estimates using the same-sex instrument also suggest no negative effect of larger families on IQ. In contrast, our IV estimates using the birth of twins as exogenous variation in family size imply that family size has a negative effect on IQ. However, our investigation of the effects of the zero spacing of twins and their low birth weights suggests that these estimates should not be generalized to other types of variation in family size. Therefore, our
conclusion is that there are likely no important negative effect on IQ of expected increases in family size that come through normally spaced singleton births.

An intriguing result is that existing children suffer intellectually as a result of the birth of twins and that this effect is only present in recent cohorts. This may reflect that fact that unexpected shocks to family size are becoming increasingly costly as the attachment of both parents to the labor force increases. More research will be needed to examine whether this result generalizes to other countries and to children born after our sample period.
References


Bjerkedal, Tor, Petter Kristensen, Geir A. Skjeret, and John I. Brevik, “Intelligence test scores and birth order among young Norwegian men (conscripts) analyzed within and between families”, Intelligence (2007).


Eide, Martha G., Nina Øyen, Rolv Skjærven, Stein Tore Nilsen, Tor Bjerkedal and Grethe S. Tell, “Size at Birth and Gestational Age as Predictors of Adult Height and Weight,” Epidemiology, XVI (2005), 175-181.


Figure 1
Fraction of Twin Births Out of All Births
Figure 2
The Reduced Form Effect of 3rd-Born Twin Births on IQ of First Two Children
9-Year Moving Average

The year on the x-axis refers to the year of the potential twin birth.
Table 1
Summary Statistics

<table>
<thead>
<tr>
<th>Variable</th>
<th>Mean (Standard Deviation)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age in 2002</td>
<td>25 (6)</td>
</tr>
<tr>
<td>Education</td>
<td>12.3 (1.9)</td>
</tr>
<tr>
<td>IQ (stanines)</td>
<td>5.2 (1.8)</td>
</tr>
<tr>
<td>Mother’s Education</td>
<td>11.3 (2.5)</td>
</tr>
<tr>
<td>Mother’s Age in 2002</td>
<td>50.5 (6.2)</td>
</tr>
<tr>
<td>Father’s Education</td>
<td>11.8 (2.9)</td>
</tr>
<tr>
<td>Father’s Age in 2002</td>
<td>53.6 (6.7)</td>
</tr>
</tbody>
</table>

Sample Size: N=388,405 male singletons.
Education is calculated for a subsample of 282,855 men aged at least 21 in 2002.
<table>
<thead>
<tr>
<th>Family Size</th>
<th>Average IQ</th>
<th>Mother’s Education</th>
<th>Father’s Education</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>5.2</td>
<td>11.1</td>
<td>11.4</td>
<td>28,083</td>
</tr>
<tr>
<td></td>
<td>(1.8)</td>
<td>(2.5)</td>
<td>(2.8)</td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>5.3</td>
<td>11.3</td>
<td>11.8</td>
<td>170,628</td>
</tr>
<tr>
<td></td>
<td>(1.8)</td>
<td>(2.5)</td>
<td>(2.8)</td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>5.2</td>
<td>11.4</td>
<td>11.9</td>
<td>132,929</td>
</tr>
<tr>
<td></td>
<td>(1.8)</td>
<td>(2.5)</td>
<td>(2.9)</td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>5.1</td>
<td>11.2</td>
<td>11.7</td>
<td>41,805</td>
</tr>
<tr>
<td></td>
<td>(1.8)</td>
<td>(2.6)</td>
<td>(2.9)</td>
<td></td>
</tr>
<tr>
<td>5</td>
<td>4.9</td>
<td>11.0</td>
<td>11.5</td>
<td>10,214</td>
</tr>
<tr>
<td></td>
<td>(1.8)</td>
<td>(2.5)</td>
<td>(3.0)</td>
<td></td>
</tr>
<tr>
<td>6 or more</td>
<td>4.9</td>
<td>10.6</td>
<td>11.2</td>
<td>4,746</td>
</tr>
<tr>
<td></td>
<td>(1.8)</td>
<td>(2.4)</td>
<td>(2.8)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>No Controls</td>
<td>Controls</td>
<td></td>
<td></td>
</tr>
<tr>
<td>------------------</td>
<td>----------------------</td>
<td>---------------------</td>
<td>------------------</td>
<td>------------------</td>
</tr>
<tr>
<td>Family</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>One Child Family</td>
<td>.099* (.031)</td>
<td>-.037 (.028)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Two Child Family</td>
<td>.274* (.029)</td>
<td>.222* (.031)</td>
<td>.034 (.026)</td>
<td>-.015 (.028)</td>
</tr>
<tr>
<td>Three Child</td>
<td>.256* (.030)</td>
<td>.236* (.031)</td>
<td>.184* (.035)</td>
<td>.052* (.026)</td>
</tr>
<tr>
<td>Four Child Family</td>
<td>.163* (.032)</td>
<td>.144* (.034)</td>
<td>.093* (.039)</td>
<td>.063 (.046)</td>
</tr>
<tr>
<td>N</td>
<td>201789 134692 42042 7879</td>
<td>201789 134692 42042 7879</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

* indicates statistical significance at the 5% level. Standard errors (in parentheses) allow for correlation of errors within family. Each column represents a separate regression. The regressions with controls include indicators for age, test year, mother’s age, mother’s age at first birth, mother’s education, father’s education, father’s IQ, and the log of birth weight of the children studied. Omitted Category is 5 or more children.
<table>
<thead>
<tr>
<th>Instrument: Twin at Second Birth</th>
<th>OLS (no controls)</th>
<th>OLS (controls)</th>
<th>First Stage</th>
<th>Second Stage</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td>(Sample: First child in families with 2 or more births)</td>
<td>.717* (.019)</td>
<td></td>
<td></td>
<td></td>
<td>173706</td>
</tr>
<tr>
<td>Number of Children in Family</td>
<td>-.052* (.005)</td>
<td>.006 (.005)</td>
<td>-.149* (.052)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

| Instrument: Twin at Third Birth      |                   |                |             |              |     |
| (Sample: First and second children in families with 3 or more births) | .839* (.018)       |                |             |              | 135970 |
| Number of Children in Family         | -.082* (.008)     | -.002 (.007)   | -.170* (.052) |             |     |

| Instrument: Twin at Fourth Birth     |                   |                |             |              |     |
| (Sample: First, second and third children in families with 4 or more births) | .903* (.036)       |                |             |              | 44695  |
| Number of Children in Family         | -.046* (.012)     | .009 (.010)    | -.115 (.080) |             |     |

| Instrument: First Two Children Same Sex |                   |                |             |              |     |
| (Sample: First and second children in families with 2 or more births) | .082* (.003)       |                |             |              | 304755 |
| Number of Children in Family         | -.045* (.004)     | .014* (.004)   | .065 (.074)  |             |     |

| Instrument: First Three Children Same Sex |                   |                |             |              |     |
| (Sample: First, second, and third children in families with 3 or more births and first two births same-sex) | .070* (.005)       |                |             |              | 94578  |
| Number of Children in Family         | -.077* (.010)     | .008* (.009)   | .076 (.171)  |             |     |

* indicates statistical significance at the 5% level. Standard errors (in parentheses) allow for correlation of errors within family. N represents number of individuals. The regressions also include indicators for age, test year, mother’s age, mother’s age at first birth, mother’s education, father’s education, father’s IQ, and log birth weight of the children studied.
### Appendix Table 1

**Distribution of Family Size**

<table>
<thead>
<tr>
<th>Family Size</th>
<th>Percentage</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>10</td>
</tr>
<tr>
<td>2</td>
<td>47</td>
</tr>
<tr>
<td>3</td>
<td>31</td>
</tr>
<tr>
<td>4</td>
<td>9</td>
</tr>
<tr>
<td>5</td>
<td>2</td>
</tr>
<tr>
<td>6 or more</td>
<td>1</td>
</tr>
</tbody>
</table>
## Appendix Table 2
**Effect of Family Size on Whether Children Finished 12 Years of Schooling**
**Instrumental Variable Estimation**

<table>
<thead>
<tr>
<th>Instrument: Twin at Second Birth</th>
<th>OLS (No Controls)</th>
<th>OLS (Controls)</th>
<th>First Stage</th>
<th>Second Stage</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>(.021)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of Children in Family</td>
<td>-.030*</td>
<td>-.013*</td>
<td>-.029</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(.001)</td>
<td>(.001)</td>
<td>(.015)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Instrument: Twin at Third Birth</th>
<th>OLS (No Controls)</th>
<th>OLS (Controls)</th>
<th>First Stage</th>
<th>Second Stage</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>(.020)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of Children in Family</td>
<td>-.034*</td>
<td>-.016*</td>
<td>-.039*</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(.002)</td>
<td>(.002)</td>
<td>(.015)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Instrument: Twin at Fourth Birth</th>
<th>OLS (No Controls)</th>
<th>OLS (Controls)</th>
<th>First Stage</th>
<th>Second Stage</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>(.040)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of Children in Family</td>
<td>-.016*</td>
<td>-.007</td>
<td>-.040</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(.003)</td>
<td>(.003)</td>
<td>(.025)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Instrument: First Two Children Same Sex</th>
<th>OLS (No Controls)</th>
<th>OLS (Controls)</th>
<th>First Stage</th>
<th>Second Stage</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>(.003)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of Children in Family</td>
<td>-.028*</td>
<td>-.012*</td>
<td>-.019</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(.001)</td>
<td>(.001)</td>
<td>(.021)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Instrument: First Three Children Same Sex</th>
<th>OLS (No Controls)</th>
<th>OLS (Controls)</th>
<th>First Stage</th>
<th>Second Stage</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>(.005)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of Children in Family</td>
<td>-.028*</td>
<td>-.014*</td>
<td>.023</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(.001)</td>
<td>(.002)</td>
<td>(.049)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

* indicates statistical significance at the 5% level. Standard errors (in parentheses) allow for correlation of errors within family. N represents number of individuals. The regressions also include indicators for age, test year, mother’s age, mother’s age at first birth, mother’s education, father’s education, father’s IQ, and the log birth weight of the children studied. Children are aged at least 21 in 2002.
**Appendix Table 3: The Effect of Child Spacing on Previous Children’s IQ**

<table>
<thead>
<tr>
<th></th>
<th>Effect of Gap less than or equal to 1 Year</th>
<th>Effect of Gap less than or equal to 18 months</th>
</tr>
</thead>
<tbody>
<tr>
<td>First Child in Families of three or more</td>
<td>-.26*</td>
<td>-.12*</td>
</tr>
<tr>
<td>N = 74996</td>
<td>(.06)</td>
<td>(.02)</td>
</tr>
<tr>
<td>First and Second Child in Families of four or more</td>
<td>-.18</td>
<td>-.08*</td>
</tr>
<tr>
<td>N = 32772</td>
<td>(.09)</td>
<td>(.03)</td>
</tr>
<tr>
<td>First Three Children in Families of five or more</td>
<td>-.55*</td>
<td>-.19*</td>
</tr>
<tr>
<td>N = 9651</td>
<td>(.17)</td>
<td>(.06)</td>
</tr>
</tbody>
</table>

* indicates statistical significance at the 5% level. Standard errors (in parentheses) allow for correlation of errors within family. Each box represents a separate regression. The regressions include indicators for age, test year, mother’s age, mother’s age at first birth, mother’s education, father’s education, father’s IQ, birth order, and the log birth weight of the child studied. Also included is a control for family size and a quadratic function of the time gap between children born before the two children whose spacing is being investigated.

The reported coefficients are those relating to the distance in years between the two children immediately after the children listed in column 1. For example, in the first box, the coefficient of -.26 suggests that first child IQ is lower by .26 if the gap between the second and third child is less than or equal to one year (in families with at least 3 children).