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Multiple Experiments for the Causal Link between the Quantity and Quality of Children

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This article presents evidence on the child-quantity/child-quality trade-off using quasi-experimental variation due to twin births and preferences for a mixed sibling sex composition, as well as ethnic differences in the effects of these variables. Our sample includes groups with very high fertility. An innovation in our econometric approach is the juxtaposition of results from multiple instrumental variables strategies, capturing the effects of fertility over different ranges for different sorts of people. To increase precision, we develop an estimator that combines different instrument sets across partially overlapping parity-specific subsamples. Our results are remarkably consistent in showing no evidence of a quantity-quality trade-off.

Family Planning: The Way to Prosperity. (Slogan on the back of Indonesia's Rp 5 coin)

I. Introduction

The question of how family size affects economic circumstances is one of the most enduring in social science. Beginning with Becker and Lewis

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(1973) and Becker and Tomes (1976), economists have developed a rich theoretical framework that sees both the number of children and parental investment per child as household choice variables that respond to economic forces. An important implication of this framework is that exogenous reductions in family size should increase parental investment in children, thereby improving human capital and welfare. By the same token, events that lead to otherwise unplanned increases in family size should reduce parental investment and therefore reduce inframarginal “child quality.”

On the policy side, the view that smaller families and slower population growth are essential for economic development motivates many international agencies and some governments to promote, or even to require, smaller families. In addition to China’s One Child Policy, examples of government-sponsored family planning efforts include a forced-sterilization program in India and the aggressive public promotion of family planning in Mexico and Indonesia.¹ Bongaarts (1994) notes that by 1990, 85% of people in the developing world lived in countries in which the government considers fertility to be too high. The Becker and Lewis (1973) model, as well as recent economic analyses of the role of the demographic transition, provides additional theoretical support for the view that large families keep living standards low (e.g., Galor and Weil 2000; Hazan and Berdugo 2002; Moav 2005).

Most of the scholarly evidence pointing to an empirical quantity-quality trade-off comes from the widely observed negative association between family size, on the one hand, and schooling or academic achievement, on the other. For example, Leibowitz (1974) and Hanushek (1992) find that children’s educational attainment and achievement growth are negatively correlated with family size. Many other microeconomic and demographic studies show similar relations.² The principal problem with re-

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¹ These episodes are recounted by Weil (2005, chap. 4), who also mentions the antinatalist slogan on the Indonesian rupiah.

² See, e.g., the review by Schultz (2005). Johnson (1999) notes that the relation between family size and economic well-being or growth is less clear-cut at the time series or cross-country level.

search of this type is the likelihood of omitted variables bias in estimates of the effects of childbearing. This is highlighted by Angrist and Evans (1998), who used instrumental variables (IVs) derived from multiple births and same-sex sibling pairs to estimate the causal effect of family size on mothers' labor supply. IV estimates, while still negative, are considerably smaller than the corresponding ordinary least squares (OLS) estimates.

This article provides new evidence on the quantity-quality trade-off using exogenous variation in family size in low- and high-fertility subsamples. We begin by looking at the effect of third and higher-parity births on first- and second-born children's completed schooling, labor market status, adult earnings, and marital status and fertility. These are important long-run "quality" indicators that are likely to be affected by the home environment. Effects on marriage and fertility also play a role in some theories of the demographic transition (Lutz and Skirbekk 2005).

Two of the instruments used here are dummies for multiple second births and for same-sex sibling pairs in families with two or more children. We also extend the sex-composition and twins identification strategies in a number of ways. First, we introduce a new source of exogenous variation in family size based on sharp differences in the effects of multiple births and sex composition across ethnic groups in the Israeli population. Second, as an alternative to instruments based on sex mix, we exploit preferences for boys at higher-order births in some ethnic groups.³ Third, we combine twins and sex-composition instruments at different parities to produce more precise IV estimates and increase the range of variation covered by our experiments. This parity-pooled analysis includes third- and fourth-born children. Finally, we present evidence for the exclusion restrictions that justify IV by estimating reduced form effects in samples with no first stage.

The fact that our analysis combines evidence from multiple sources of variation is important for a number of reasons. First, both twins and sex-composition instruments are potentially subject to omitted variables biases. For example, twin rates vary with maternal characteristics like age at birth and race, and twin births affect child spacing and child health in a manner that seems likely to accentuate any negative effects of childbearing. IVs derived from sibling sex composition are not subject to these considerations, although sex composition may affect outcomes due, say, to economies of scale through room or clothes sharing (as suggested by Rosenzweig and Wolpin 2000). A comparison of twins and sex-compo-

³ Traditional Jewish preferences over sibling sex composition can be traced to the Mishnah (oral law): a man shall not stop having children until he has two. Beit Shammai (a relatively strict rabbinic tradition) says two sons, while Beit Hillel (a more forgiving rabbinic tradition) says a boy and a girl. As it is written in Genesis, "male and female he created them" (Mishnah Nashim—Yebhamoth 6: 7).

sition estimates therefore provides a specification check since the omitted variables biases associated with each type of instrument should differ. The use of instruments based on preferences for male children per se also provides a simple check on IV estimates derived from sex mix.

A related consideration arises from the fact that the estimates generated by any particular IV strategy capture effects on individuals affected by that instrument (Imbens and Angrist 1994). Moreover, in models with variable treatment intensity, IV results are specific to the range of variation induced by the instrument (Angrist and Imbens 1995). As noted by Moffit (2005), these limitations lead to concerns about the external validity of IV estimates. Our analysis addresses these concerns by juxtaposing results from different quasi-experimental research designs. On the one hand, as we show below, twins instruments identify the effect of treatment on the nontreated since compliance is perfect when a multiple birth occurs. On the other hand, the average causal response (ACR) due to a twin birth reflects the impact of increasing family size only (or mostly) at the parity of occurrence. Sex-composition instruments, in contrast, shift the fertility distribution at parities as high as nine. Moreover, the ethnic composition of same-sex compliers tends to vary in a manner opposite to that for twins. We therefore argue that the fact that instruments affecting different people and inducing differing ranges of variation generate similar results provides considerable evidence for the external validity of our estimates.

A limitation of our quasi-experimental identification strategies is that they fail to capture effects on the marginal child. Specifically, we can study the effect on an older child of having an extra younger sibling but not the effect on a younger child of being born into a larger family (whether the family is large because of twinning, sex preferences, or any other reason). At the same time, the similarity of our results across alternative identification strategies, subpopulations, and fertility increments offers no evidence of substantially heterogeneous effects by parity or birth order.

Our article is related to a burgeoning empirical literature that uses multiple births to estimate the causal effects of family size. Rosenzweig and Wolpin (1980) appear to have been the first to use twins to estimate a child-quantity/child-quality trade-off. Other estimates using multiple births include those by Duflo (1998), who looks at effects on child mortality in Indonesia; Caceres (2006), who looks at effects on private schooling and grade retention in U.S. census data; and Black, Devereux, and Salvanes (2005), who use twins to estimate family-size effects on education and earnings in Norway. As in our article, Black et al. (2005) look at human capital variables with a large administrative sample. In contrast to the original Rosenzweig and Wolpin study, this literature has uncovered surprisingly little evidence for an adverse effect of family size on human capital. However, a recent paper by Rosenzweig and Zhang (2009), using the effect of twins instruments on twins themselves to bound quantity-

quality effects, suggests there is a trade-off. Because twins probably differ from nontwins for reasons both observed and unobserved, we prefer empirical strategies that look at the effects of twins on older siblings.⁴ Nevertheless, we also briefly discuss results from the Rosenzweig and Zhang (2009) approach.

To the best of our knowledge, none of this previous work has attempted to combine or reconcile evidence from multiple natural experiments. Our article also differs from Black et al. (2005) and Caceres (2006) in that we study a higher-fertility population with demographic and social characteristics much closer to developing country populations. Of particular interest is the Asia-Africa (AA) subsample, that is, Sephardic Jews of North African and Middle Eastern origin. Sephardic Jews are poor, relative to the Israeli average, and typically have larger families.⁵

On the methodological side, our article has features in common with Oreopoulos (2006), which compares IV estimates of the returns to schooling using changes in compulsory schooling laws in different countries. Oreopoulos argues that this comparison can be used to gauge the importance of treatment-effect heterogeneity when the size of the compulsory-schooling first stage varies. A final contribution stems from the relative precision of our estimates. Having established that different instruments and samples generate broadly similar effects, we develop a simple two-stage least squares (2SLS) procedure that combines parity-specific IV estimates into a single estimate that is more precise than the twins estimates reported by Black et al. (2005).

The next section describes the data and the construction of the analysis samples and discusses data quality issues such as the relationship between instruments and match rates. Section III discusses the first-stage estimates and their implications for treatment effect heterogeneity and nonlinearity, while Section IV presents the main OLS and 2SLS results. On balance, the results reported here offer little evidence for an effect of family size on schooling, work, or earnings, although we do find some effects on girls' marital status, age at marriage, and fertility. Section V discusses possible explanations for these findings, and Section VI concludes and suggests directions for further work.

II. Data and Samples

The main sources of data used here are the 20% public-use microdata samples from the 1995 and 1983 Israeli censuses, linked with information

⁴ Similarly, when using sex mix as an instrument, we look only at older siblings because the outcomes of the last child born come from an endogenously selected sample if fertility is endogenous.

⁵ In 1975, when the subjects we study were growing up, Israel was an upper-middle-income country, with gross domestic product per capita similar to that of Greece and Argentina; see Heston, Summers, and Aten (2002).

on parents and siblings from the population registry. The Israeli census microfiles are one-in-five random samples that include information collected on a fairly detailed long-form questionnaire similar to the one used to create the public use microdata sample files for U.S. censuses.⁶ The set of Jewish long-form respondents age 18–60 provides our initial study sample. In the discussion that follows, we refer to these individuals as “subjects,” to distinguish them from their parents and siblings, for whom we also collected data. The link from census to registry is necessary for our purposes because in a sample of adult respondents, most of whom no longer live with their parents and siblings, the census provides no information about sibship size, multiple births, or sibling sex composition.⁷

A. Match Rates and Sample Selection

The vast majority of our census subjects appear in the population registry. This can be seen in table 1, which reports sample sizes and subject-to-registry match rates, grouped according to whether subjects’ parents were Israeli born, birth cohort, and whether subjects were Israeli born (results are shown for each census). Subject-to-registry match rates range from 95% to 97%, regardless of cohort and nativity. The first coverage shortfall from our point of view is the failure to obtain an administrative record for subjects’ mothers. This failure arises for a number of reasons. First, subject’s mothers may have been alive but not at home in 1948 when the registry was created, or a mother may have been deceased. Second, children are more likely to be linked to parents and siblings when a subject’s mother gave birth to all of her children in Israel.

Table 1 also describes the impact of these record-keeping constraints on our census-to-registry match rates. The mothers of 1995 census subjects with Israeli-born fathers were found 90% of the time for cohorts born during or after 1955. However, for those born before 1955, only 17% of mothers were found. Likewise, for those with foreign-born fathers, there is a similar age gradient in mothers’ match rates. Even in this group, however, 87% of mothers were found for younger Israeli-born subjects in the 1995 census. The 1955 birth cohort marks a useful division

⁶ Documentation can be found at the Israel Social Sciences Data Center Web site: http://isdc.huji.ac.il/mainpage_e.html (data sets 115 [1995 demographic file] and 301 [1983 file]). The census includes residents of dwellings inside the state of Israel and Jewish settlements in the occupied territories. This includes residents abroad for less than 1 year, new immigrants, and noncitizen tourists and temporary residents living at the indicated address for more than a year.

⁷ About 80% of the Israeli population is Jewish. The study sample is limited to Jews because census-to-population-registry match rates are considerably lower for other groups. Additional information related to data set construction appears in the appendix.

Table 1
Match Rates and Sample Selection

	Foreign-Born Father											
	Israeli-Born Father			Foreign-Born Subject								
	Subject Born < 1955 (1)	Subject Born ≥ 1955 (2)		Subject Born < 1955 (3)	Subject Born ≥ 1955 (4)	Subject Born < 1955 (5)	Subject Born ≥ 1955 (6)					
1995 census:												
All subjects:	9,453	56,534		118,633	72,340	58,767	161,331					
Matched to registry (N, %)	9,057	95.8	54,073	95.6	115,123	97.0	68,788	95.0	57,098	97.2	156,096	96.8
Matched mother + siblings (N, %)	1,573	16.6	50,597	89.5	7,600	6.4	32,472	44.9	11,351	19.3	139,783	86.6
Selected sample:												
Mothers born ≥ 1930, age 15–45 at first birth:	494	48,683		1,166	26,217	2,556					119,928	
Israeli-born mothers or immigrants who arrived since 1948 and before age 45	419	47,022		1,127	22,704	2,211					115,783	
First and second borns in families with two or more births	349	34,778		1,008	15,443	1,937					67,952	
1983 census:												
All subjects:	11,049	12,665		160,459	25,025	66,761					70,662	
Matched to registry (N, %)	9,704	87.8	10,867	85.8	140,932	87.8	20,691	82.7	60,105	90.0	62,141	87.9
Matched mother + siblings (N, %)	1,289	11.7	9,258	73.1	7,380	4.6	14,557	58.2	10,767	16.1	50,785	71.9
Selected sample:												
Mothers born ≥ 1930, age 15–45 at first birth:	421	7,854		1,065	9,197	2,438					34,560	
Israeli-born mothers or immigrants who arrived since 1948 and before age 45:	318	6,952		1,045	8,913	2,138					32,368	
First and second borns in families with two or more births	232	3,657		730	3,425	1,519					12,095	

NOTE.—This table shows sample sizes and match rates at each step of the link from census data to the population registry. The target population consists of Jewish census respondents age 18–40 in 1995 and 1983. The impact of sample selection criteria on sample sizes is also shown.

for our purposes because mothers of subjects born after 1954 gave birth to most of their children in post-1948 Israel (the mothers in this group were mostly born after 1930, and, assuming childbirth starts at 18, this dates their first births at 1948 or later).

Given the match rates in table 1, our analysis sample is weighted toward post-1955 cohorts (i.e., 40 or younger in 1995). This accounts for about two-thirds of the 1995 population age 18–60. Among the children of immigrant fathers, we are also much more likely to find mothers of those who are Israeli born. The coverage rates for post-1955 Israeli-born cohorts seem high enough that we are likely to have information on mothers for a representative sample of younger cohorts, regardless of fathers' nativity. We also used information on mothers in the matched sample to discard mothers who were born before 1930 (as the match rates for this group appeared to be very low anyway). Subjects with mothers whose first birth was before age 15 or after age 45 were also dropped. These restrictions eliminate almost all subjects born before 1955, primarily because most of those born earlier have mothers born before the 1930 maternal age cutoff. We also restricted the sample of subjects with foreign-born mothers to those whose mothers arrived in 1948 or later and before age 45 (in this case so that an immigrant mother with children is likely to have come with all her children, who would then have been included in the registry, either in the first census or at the time identification numbers, IDs, were issued to the family).

The final sample restriction retains only first- and second-born subjects since these are the people exposed to the natural experiments exploited by the twins and sex-composition research designs. Note that the restriction to first- and second-born subjects naturally eliminates a higher percentage of younger rather than older cohorts. This restriction also has a bigger effect on the Israeli-born children of foreign-born fathers than on other nativity groups, probably because these children were disproportionately likely to have been born to immigrant fathers who arrived with a large wave of immigrants from Asia and Africa in the 1950s. Immigrants from this group typically formed large families after arrival and will therefore have contributed more higher-parity births to the sample.

A key issue for the internal validity of our study is whether the instruments affect the likelihood of a successful match between census and registry files. Specifically, the matched censuses might constitute a non-random sample that is selected in some manner related to the key variables in our study. We explored this by looking at match rates from registry to census as a function of family size and our instruments. The maximum theoretical match rate here is 20%. This "reverse match" finds about 15% of registrants who were alive in 1983 and about 17% of registrants who were alive in 1995. Discrepancies between theoretical and actual match rates can arise because of differences in the definition of the base popu-

lation in the registry and the census, missing or invalid IDs on census records, and census nonresponse rates.⁸ The means and first-stage regressions coming out of the reverse match are almost identical to those generated by our working extract.⁹ Most importantly, there is no significant relation between the twins and same-sex instruments and the probability of appearing in either the 1983 or the 1995 census data. This is documented in table A1 and discussed in the appendix. There is a small statistically significant decrease in the probability that someone from a larger family appears in the 1983 census, which can be explained by the higher census nonresponse rates of ultraorthodox households (Central Bureau of Statistics 1985). Nevertheless, since the bulk of our sample is from 1995, the results are similar in the two data sets, and there is no differential selection as a function of the instruments, it seems unlikely that differential match rates affect the results.

B. Description of Analysis Samples

We work with two main analysis samples, both described in table 2. One consists of firstborn subjects in families with two or more births (the 2+ sample; $N = 89,445$). The second sample consists of first- and second-born subjects in families with three or more births (the 3+ sample; $N = 65,673$ firstborns and $52,964$ second borns). These samples are defined conditional on the number of births instead of the number of children so that multiple-birth families can be included in the analysis samples without affecting the sample selection criteria.¹⁰ Twin subjects were dropped from both samples, however.

Roughly three-quarters of the observations in each sample were drawn from the 1995 census. On average, subjects were born in the mid-sixties, and their mothers were in their early twenties at first birth. Because out-of-wedlock childbearing is rare in Israel, especially among the cohorts studied here, virtually all subjects in both samples were born to married mothers. Naturally, however, some marriages have since broken up, and

⁸ The 1983 and 1995 censuses are documented in Central Bureau of Statistics (1985, 2001).

⁹ For technical reasons related to the Central Bureau of Statistics' data-handling protocols, the population of registrants used for the reverse match differs slightly from the registry population used to construct our main extract. In practice, these differences have no bearing on our analysis.

¹⁰ A 3+ sample defined as including firstborn children from families with three or more children instead of three or more births would include all families with multiple second births, even though these families differ in that they choose to love only two children. An additional advantage of birth-based sample definitions is that sibling sex composition can be defined without the need to determine which, say, of two twins, constitutes the second child.

Table 2
Analysis Samples

	Full Sample			Asia-Africa Sample		
	2+	3+		2+	3+	
	Firstborns (1)	Firstborns (2)	Second Borns (3)	Firstborns (4)	Firstborns (5)	Second Borns (6)
1995 census	.758	.753	.775	.706	.705	.732
Mother married or widowed in 2003	.910	.926	.932	.921	.932	.937
Endogenous variables:						
No. of children	3.63	4.22	4.32	4.31	4.67	4.76
More than two children	.74	1.00	1.00	.87	1.00	1.00
More than three children	.400	.545	.573	.593	.686	.704
Family composition:						
Twins at second birth	.009	.006008	.006	...
Twins at third birth010	.010	.008	.009	.009
Boy at first birth	.517	.518	.527518	.528
Boy at second birth	.514	.515	.507	.516	.514	.504
Boy at third birth515	.517509	.516
Girl12 = 1	.233	.239	.237	.232	.236	.234
Boy12 = 1	.265	.272	.272	.265	.267	.267
Girl123 = 1115	.114117	.113
Boy123 = 1140	.140138	.138
Control variables:						
Age on census day	26.2	26.4	25.5	27.4	27.5	26.4
Year of birth	1966	1965	1967	1964	1964	1965
Mother's age on census day	49.1	48.8	50.4	49.7	49.5	50.7
Mother's year of birth	1943	1943	1942	1942	1942	1941
Mother's age at first birth	22.7	22.2	22.1	22.0	21.7	21.7
Mother's age at immigration (for non-Israeli mothers)	17.4	15.7	15.9	15.6	15.4	15.7
Mother's ethnicity:						
Israel	.344	.354	.315	.167	.161	.138
Asia-Africa	.397	.468	.507	.792	.805	.830

Former USSR	.115	.068	.064	.011	.009	.007
Europe-America	.144	.111	.113	.030	.025	.025
Father's ethnicity:						
Israel	.274	.282	.248
Asia-Africa	.426	.501	.535	1	1	1
Former USSR	.114	.068	.068
Europe-America	.186	.149	.148
Subject ethnicity:						
Israel	.836	.869	.887	.856	.852	.878
Asia-Africa	.061	.074	.065	.144	.148	.122
Former USSR	.066	.029	.024	.000	.000	.000
Europe-America	.037	.027	.025	.000	.000	.000
Education outcomes:						
Highest grade completed	12.6	12.5	12.3	12.2	12.1	12.0
Schooling ≥ 12	.824	.813	.802	.759	.754	.752
Some college (age ≥ 24)	.291	.262	.224	.177	.169	.143
College graduate (age ≥ 24)	.202	.180	.153	.117	.111	.093
Labor market outcomes (age ≥ 22):						
Worked during the year	.827	.820	.809	.812	.809	.798
Hours worked last week	32.6	32.4	31.7	32.5	32.4	31.7
Monthly earnings (1995 shekels)	2,997	2,920	2,721	2,847	2,820	2,621
ln(earnings) for full-time workers	8.24	8.23	8.18	8.20	8.19	8.15
Marriage and fertility:						
Married on census day	.446	.465	.418	.519	.530	.479
Married by age 21 (age ≥ 21)	.172	.183	.171	.198	.205	.194
No. of own children (women only)	1.00	1.08	.98	1.28	1.32	1.20
N	89,445	65,673	52,964	38,063	32,875	28,357

NOTE.—This table shows descriptive statistics for the main analysis samples. The 2+ sample consists of firstborn census subjects from families with two or more births including the subject. The 3+ sample consists of first- and second-born census subjects from families with three or more births including the subject. The Asia-Africa subsample consists of census subjects whose fathers' ethnicity is identified as Asia-Africa in the census.

some wives have been widowed. This is reflected in the 2003 marital status variables available in the registry.¹¹

The Jewish Israeli population is often grouped by ethnicity, with Jews of African and Asian origin (AA; e.g., Moroccans) distinguished from Jews of European and North American (EA) origin. The 2+ sample is about 40% AA (defined using father's place of birth), while the 3+ sample is over half AA. A preference for larger families in the AA population is also reflected in the statistics on number of children. Average family size ranges from 3.6 in the 2+ sample to 4.2 in the 3+ sample (4.3 for second borns). In the AA subsample, however, the corresponding family sizes are about 4.3 and 4.7.

Table 2 also reports statistics for the variables used to construct IVs. The twin rate was 9/10 of 1% at second birth in the 2+ sample and 1% at third birth in the 3+ sample, with similar rates in the AA and full samples.¹² As expected, about 51% of births are male, regardless of birth order. Consequently, about half of the 2+ sample was born into a same-sex sibling pair and about one-quarter of the 3+ sample was part of a same-sex threesome.

The outcome variables described in table 2 measure subjects' educational attainment, labor market status and earnings, and marital status and fertility. Most Israelis are high school graduates, while 20% are college graduates. In the AA subsample, however, the proportion of college graduates is much lower. Most of our subjects were working at the time they were interviewed and earned about 3,000 shekels (about \$1,000) per month on average (including zeros). About 45% of subjects were married, although marriage rates are higher in the AA subsample.

III. First-Stage Estimates, Interpretation of IV Estimates, and Instrument Validity

Different instruments generate different average causal effects. Of particular importance in this context are the links between first-stage effects and the subpopulations affected by each underlying natural experiment and the relation between first-stage effects and the range of variation induced by each instrument. These points are detailed below.

¹¹ The 2+ sample of firstborns includes the 3+ sample of firstborns. In the 3+ sample, about 10% of first and second borns have the same mother (both must appear in the 20% census sample and be in the relevant age range). We therefore cluster analyses that pool parities by mothers' IDs.

¹² The second-birth twin rate in the 3+ sample is not comparable to the second-birth twin rate in the 2+ sample because the 3+ sample consists of those who had three or more *births*. Families with a second-born twin in the 3+ sample chose to have a fourth child and are therefore unusual.

Table 3
Twins First Stage

	Two or More Births		Three or More Births			
	Firstborns		Firstborns		First and Second Borns	
	(1)	(2)	(3)	(4)	(5)	(6)
Twins2	.437 (.050)	.625 (.057)
Twins2 × Asia-Africa	...	-.484 (.105)
Twins3522 (.045)	.583 (.045)	.585 (.043)	.692 (.049)
Twins3 × Asia-Africa	-.132 (.094)	...	-.226 (.086)
Male	-.018 (.010)	.000 (.012)	.016 (.018)	.018 (.023)	.014 (.011)	.006 (.015)
Male × Asia-Africa	...	-.041 (.022)	...	-.005 (.035)015 (.022)
Asia-Africa	.242 (.015)	.267 (.019)	.166 (.016)	.161 (.027)	.083 (.014)	.069 (.021)

NOTE.—This table shows first-stage effects of twins2 and twins3 on number of children. The sample includes nontwins age 18–60 in the 1983 and 1995 censuses as described in table 1. In addition to the effects reported here, the regressions include indicators for age, missing month of birth, mother's age, mother's age at first birth, mother's age at immigration (when relevant), father's and mother's place of birth, and census year. Regressions for cols. 3–6 also include controls for girl12, boy12, and twins at second birth. Regressions for cols. 5–6 also include indicators for second born and birth spacing between first and second birth. Robust standard errors are reported in parentheses. Standard errors in cols. 5–6 are clustered by mother's ID.

A. Twins First Stages

A multiple second birth increases the average number of siblings in the 2+ sample by about half a child, a finding reported in column 1 of table 3, which gives first-stage estimates for the twins experiment. In particular, column 1 reports estimates of the coefficient α in the equation

$$c_i = X_i'\beta + \alpha t_{2i} + \eta_i, \quad (1a)$$

where c_i is subject i 's sibship size (including the subject), X_i is a vector of controls that includes a full set of dummies for subjects' and subjects' mother's ages, mothers' age at first birth, mothers' age at immigration (when relevant), fathers' and mothers' place of birth, census year and a dummy for missing month of birth. The variable t_{2i} (which we call twins2) indicates multiple second births in the 2+ sample.

The Israeli twins2 first stage is smaller than the twins2 first stage of about .6 in the Angrist and Evans (1998) sample, reflecting the fact that Israelis typically have larger families than Americans. Multiple births result in a smaller increase in family size when families would have been large even in the absence of a multiple birth. Within Israel, however, there are marked differences in the twins first stage by ethnicity. This can be seen in column 2 of table 3, which reports the twins2 main effect and an

interaction term between *twins2* and a dummy for AA ethnicity (a_i) in the equation

$$c_i = X_i'\beta + \alpha_0 + \alpha_1 a_i t_{2i} + \eta_i. \quad (1b)$$

The *twins2* main effect, α_0 , captures the effect of a multiple birth in the non-AA population, while the interaction term, α_1 , measures the AA/non-AA difference.¹³ The estimates in column 2 show that non-AA family size goes up by about .63 in response to a multiple birth (similar to the Angrist and Evans [1998] first stage), while AA family size increases by only .63 – .48 = .15. Both α_0 and α_1 are very precisely estimated.

The remaining columns of table 3 report the first-stage effect of a multiple third birth in the 3+ sample. The *twins3* effects were estimated in the 3+ sample by replacing t_{2i} with t_{3i} , a dummy for multiple third births, in equations (1a) and (1b). These results are reported in columns 3–4 for firstborns and columns 5–6 for the pooled sample of first and second borns. The first-stage effect of a multiple birth is bigger in the 3+ sample than in the 2+ sample because the desire to have additional children diminishes as family size increases. For the same reason, the effect of t_{3i} differs less by ethnicity in the 3+ sample than in the 2+ sample, although, as the estimates in column 6 show, there is still a significant difference by ethnicity when first- and second-born subjects are pooled.

B. Heterogeneity and Nonlinearity in Response to a Multiple Birth

The difference in first-stage effects across ethnic groups has a useful interpretation in the average causal response (ACR) framework laid out by Angrist and Imbens (1995). To see this, define potential endogenous variables C_{0i} and C_{1i} to be the number of children a woman would have if a generic binary instrument, Z_i , is equal to zero or one. Because we observe C_{0i} for those with Z_i equal to zero and C_{1i} for those with Z_i equal to one, the realized number of children is

$$c_i = C_{0i} + (C_{1i} - C_{0i})Z_i.$$

For a model without covariates, the IV estimand using this instrument is the Wald estimator (see, e.g., Angrist 1991):

$$\beta_w = \frac{E[y_i | Z_i = 1] - E[y_i | Z_i = 0]}{E[c_i | Z_i = 1] - E[c_i | Z_i = 0]},$$

where y_i is the outcome variable. The observed y_i is related to potential outcomes, $Y_i(j)$, where j indexes possible values of c_i , as follows:

$$y_i = Y_i(0) + \sum_j [Y_i(j) - Y_i(j-1)]1[c_i \geq j], \quad (2)$$

¹³ The a_i main effect is included in the vector of covariates, X_i . Note that the covariate effects, labeled β , differ as the first-stage specification and sample change.

and the summation is from $j = 1, \dots, J$. Potential outcome $Y_i(j)$ tells us what would happen to woman i if her $c_i = j$.

A linear constant-effects model imposes the restriction $Y_i(j) - Y_i(j-1) = \rho$, for all i and j , in which case the Wald estimator equals this parameter. More generally, Angrist and Imbens (1995) show that

$$\beta_w = \sum_j E[Y_i(j) - Y_i(j-1) | C_{1i} \geq j > C_{0i}] \omega(j), \quad (3)$$

where the weighting function, $\omega(j)$, is

$$\omega(j) = \frac{P(C_{1i} \geq j > C_{0i})}{\sum_j P(C_{1i} \geq j > C_{0i})}.$$

Thus, the Wald estimator is a weighted ACR for people from families induced by an instrument to go from having fewer than j to at least j children, weighted over j by the probability of crossing this threshold.¹⁴

It is straightforward to show that the denominator normalizing the weights, $\omega(j)$, is the Wald first stage. In other words,

$$E[c_i | Z_i = 1] - E[c_i | Z_i = 0] = E[C_{1i} - C_{0i}] = \sum_j P(C_{1i} \geq j > C_{0i}).$$

This relation is important because we can think of individuals with $C_{1i} \geq j > C_{0i}$ for any j in the support of c_i as compliers in the sense of Angrist, Imbens, and Rubin (1996). In this context, the subpopulation of compliers consists of individuals who switch from having fewer than j to at least j children because of the instrument. Differences in the size of the first stage across demographic or ethnic groups measure differences in the probability of compliance between these groups.

As a practical matter, we can use the ratio of first stages for the AA and overall sample to measure the likelihood that twins2 compliers are of AA ethnicity. To see this, note that

$$\frac{E[C_{1i} - C_{0i} | a_i = 1]}{E[C_{1i} - C_{0i}]} = \sum_j \frac{P(a_i = 1 | C_{1i} \geq j > C_{0i})}{P(a_i = 1)} \omega(j).$$

Thus, the ratio of the first stage for the AA subsample to the overall first stage summarizes the extent to which compliers are AA, relative to the population proportion AA. The fact that AA family size increases by

¹⁴ The assumptions that lay behind the ACR theorem are that (a) potential outcomes and treatment assignments are independent of the instrument and that (b) the instrument moves fertility in one direction only (monotonicity); i.e., $C_{1i} \geq C_{0i}$. With covariates, the interpretation of the ACR is more elaborate, but the basic idea is preserved. Because some parents may prefer a mixed sibship while others may prefer same-sex sibships, monotonicity need not hold for sex-composition instruments. As a partial check on monotonicity, we estimated the same-sex first stage separately by intervals of individual year of birth, maternal age at first birth, and ethnicity. Only three out of 36 cells generated negative estimates, and all 16 significant estimates were positive.

only .15 in response to a second twin birth while the overall first stage is .44 means that the population of twins compliers is less than half as likely as the overall population to be AA. In contrast, sex-composition compliers are disproportionately likely to be AA, as we show below.

A second important feature of the twins identification strategy is the fact that twins instruments capture the causal effect of childbearing in a narrow range. Figure 1, which plots first-stage estimates of the effect of twins2 and twins3 on $d_{ji} \equiv 1(c_i \geq j)$ ($j = 1, \dots, 11$), along with the associated confidence bands, documents this. The normalized cumulative distribution function (CDF) differences plotted in figure 1 are the weights in the ACR decomposition of β_w in equation (3). The figure therefore implies that twins instruments capture an average causal effect over a range of fertility variation that is close to the parity of the multiple birth. For example, a multiple third birth increases the likelihood of having a fourth child by about .35 in the AA 3+ subsample, with a much smaller effect on the likelihood of having a fifth child and no significant effect at higher parities (see the lower-left panel of fig. 1).¹⁵

The last distinctive econometric feature of the twins estimates is that they generate the average causal effect of treatment on the nontreated, where treatment is defined as a dummy for having another child. Specifically, the subpopulation of compliers affected by the twins2 instrument is the entire population with two children. This is a consequence of the causal treatment-effects framework outlined in Angrist et al. (1996), which divides the population into three types of responders affected by a Bernoulli instrument: always-takers who always get treated, never-takers who never get treated, and compliers who get treated when the instrument is switched on but not otherwise. The treated consist of always-takers plus compliers with the instrument switched on, while the nontreated consist of never-takers plus compliers with the instrument switched off. But with twins there are no never-takers, so the nontreated consist only of compliers with the twins instrument switched off. Because twinning is as good as randomly assigned, causal effects for the latter population are the same as causal effects on all compliers. From this we conclude that the parameter identified by twins instruments is the average effect on the nontreated.¹⁶

¹⁵ The twins instrument induces small shifts in fertility at parities beyond the twinning parity because a multiple birth leads to tighter spacing, thereby lengthening the biological window for continued childbirth. This is most likely to be relevant for the members of the ultraorthodox minority who have very high fertility.

¹⁶ Here is a more formal argument: note first that, for the twins instrument, $P(C_{1i} \geq 3 > C_{0i}) = P(C_{0i} = 2)$ since $C_{1i} \geq 3$ and $C_{0i} \geq 2$ for everybody in the 2+ sample. Moreover, $P(C_{1i} \geq j > C_{0i})$ is close to zero for $j > 3$ since a multiple second birth has little effect on childbearing at higher parities. Therefore, $\beta_w = E[Y_i(3) - Y_i(2) | C_{0i} = 2]$. Finally, because Z_i is independent of potential outcomes

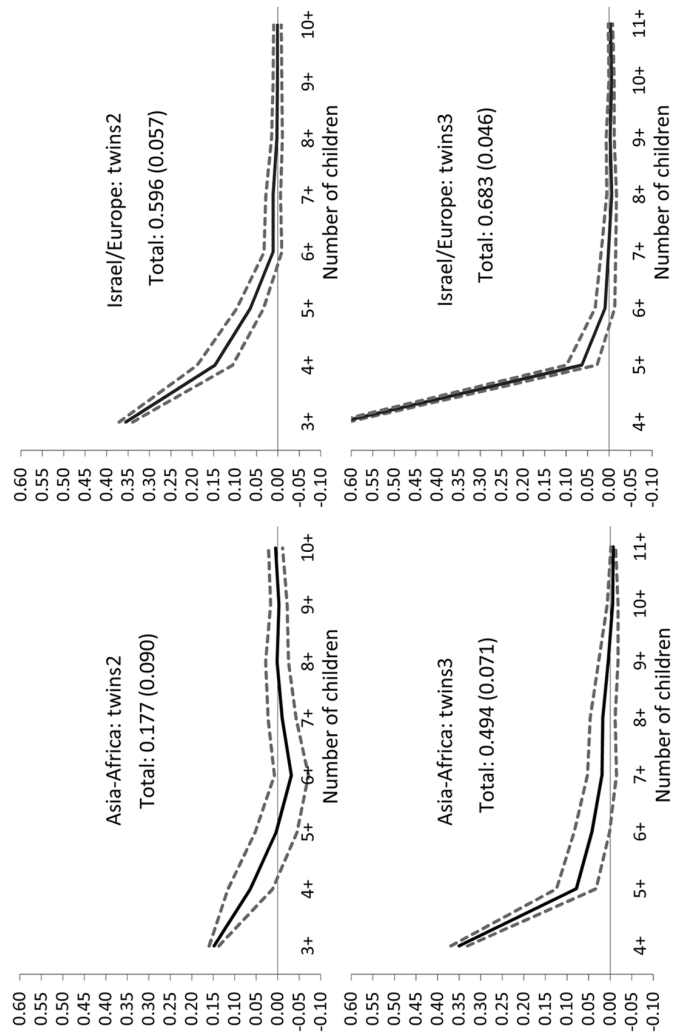


FIG. 1.—Firstborns in the 2+ sample: first-stage effects of twins2 (*top*). First and second borns in the 3+ sample: first-stage effects of twins3 (*bottom*). Dashed lines are confidence bands. Overall first stages and standard errors are shown in the plot area.

Sex-composition instruments identify average causal effects that differ in two ways from the effects captured by twins. On one hand, the compliers population is less complete; not all the nontreated are affected by sex composition. On the other hand, as we show below, sex-composition instruments (including a dummy for third-born male children) shift the fertility distribution over a wider range than does a multiple birth, especially in the event of an all-female sibship.

C. Sibling Sex-Composition First Stages

Sex-composition first stages in the 2+ sample were estimated using the following two models:

$$c_i = X_i'\beta + \gamma_1 b_{1i} + \gamma_2 b_{2i} + \pi_s s_{12i} + \eta_i, \quad (4a)$$

$$c_i = X_i'\beta + \gamma_1 b_{1i} + \pi_b b_{12i} + \pi_g g_{12i} + \eta_i, \quad (4b)$$

where b_{1i} (boy-first) and b_{2i} (boy-second) are dummies for boys born at first and second birth, the variable $s_{12i} = b_{1i}b_{2i} + (1 - b_{1i})(1 - b_{2i})$ is a dummy for same-sex sibling pairs, and $b_{12i} = b_{1i}b_{2i}$ and $g_{12i} = (1 - b_{1i})(1 - b_{2i})$ indicate two boys and two girls. Note also that b_{1i} indicates the subject's sex in the 2+ sample and that $s_{12i} = b_{12i} + g_{12i}$. The first model controls for boy-first and boy-second main effects, while the excluded instrument is a same-sex effect common to boy and girl pairs. The second model allows the effect of two boys and two girls to differ, although one of the boy main effects must be dropped since $\{b_{1i}, b_{2i}, b_{12i}, g_{12i}\}$ are linearly dependent.¹⁷ We also report results from models with AA interaction terms, as in table 3.

The first-stage effect of s_{12i} in the 2+ sample, reported in column 1 of table 4, is .073 children. The AA interaction term in this case is essentially zero, so that in contrast with the twins first stage, the overall sex-composition effect in the 2+ sample is the same for the AA and non-AA populations.

In models with common effects across ethnic groups, two girls increases family size by .11, while the effect of two boys is .039. This can be seen in columns 3 and 4 of table 4, which report estimates of π_b and π_g in equation (4b). Models allowing different coefficients by ethnicity generate

and potential treatment assignments, $\beta_w = E[Y_i(3) - Y_i(2) | C_{0i} = 2, Z_i = 0]$. But this is the same as $E[Y_i(3) - Y_i(2) | c_i = 2]$ because all those with two children have singleton births and $C_{0i} = 2$ and vice versa. A similar argument leads to the conclusion that the twins3 estimator in the 3+ sample identifies $E[Y_i(4) - Y_i(3) | c_i = 3]$.

¹⁷ For example, $g_{12i} = 1 - b_{1i} - b_{2i} + b_{12i}$. Control for boy-first and boy-second main effects is motivated by the fact that the same-sex interaction term is, in principle, correlated with the main effects when the probability of male birth exceeds .5. In practice, however, this matters little both because the correlation is small and because the main effects are small.

Table 4
Sex-Composition First Stages

	2+				3+							
	Firstborns				Firstborns							
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
Same sex	.073 (.010)	.071 (.012)133 (.017)	.095 (.022)120 (.015)	.070 (.019)
Same sex × Asia-Africa005 (.022)	-.076 (.034)	-.097 (.029)
All girls110 (.015)	.088 (.017)180 (.025)	.095 (.032)183 (.022)	.072 (.027)
All girls × Asia-Africa051 (.028)169 (.051)217 (.043)
All boys039 (.015)	.055 (.016)092 (.023)	.095 (.029)065 (.021)	.070 (.025)
All boys × Asia-Africa	-.038 (.026)	-.005 (.047)	-.009 (.041)
Boy3 × (1 - sameex12)	-.080 (.018)	-.053 (.023)	-.077 (.015)	-.044 (.019)
Boy3 × (1 - sameex12) × Asia-Africa	-.054 (.035)	-.064 (.030)

NOTE.— This table shows first-stage effects of sex composition and boy3 on number of children. The sample for cols. 1–4 includes firstborn nontwins from families with two or more births (2+). The sample for cols. 5–8 includes firstborn nontwins from families with three or more births (3+). The sample for cols. 9–12 includes first- and second-born nontwins from families with three or more births. Regression estimates are from models that include the control variables specified in the note to table 3. Robust standard errors are reported in parentheses. Standard errors in cols. 9–12 are clustered by mother's ID.

a two-girls effect equal to .088 in the non-AA population, while the effect of two girls in the AA sample is larger by .051. In contrast, the two-boys effect is only .055 in the non-AA population, and the AA two-boys effect is smaller by .038. As a result, the AA population appears to increase childbearing in response to the birth of two girls but not in response to the birth of two boys.

The sex-composition first stage in the 3+ sample captures the effect of an all-boy or an all-girl triple on first- and second-born subjects, controlling for the sex composition of earlier births. The first stage therefore conditions on b_{12i} and g_{12i} as well as a subject-sex main effect and a birth-order dummy. Additional variables included in these models are dummies for the sex of the third child, an effect that is defined conditional on a mixed-sex sibling pair at first and second birth (because for families with $b_{12i} = 1$, the boy-third effect is the same as having an all-male triple, while for families with $g_{12i} = 1$, the boy-third effect is the same as having an all-female triple). The resulting model can be written as follows (we spell out notation only for the model that allows for separate all-male and all-female effects):

$$c_i = X_i'\beta + \gamma_1 b_i + \delta_b b_{12i} + \delta_g g_{12i} + \gamma_3(1 - s_{12i})b_{3i} \\ + \lambda_b b_{123i} + \lambda_g g_{123i} + \eta_i, \quad (5)$$

where b_{123i} and g_{123i} are indicators for all-male and all-female triples and b_i is the subject's sex.¹⁸ The term b_{3i} (boy3) is also used as an instrument, although we postpone a discussion of the associated first stage for the moment. The sex-composition effects in this model are reported in columns 5–12 of table 4.

The overall same-sex effect in the 3+ sample is .12 among first and second borns. This can be seen in column 9 of table 4 (results for firstborns only, reported in cols. 5–8, are similar). The AA interaction term generates a large ethnic differential in sex-composition effects. For example, the same-sex effect among first- and second-born non-AA subjects, reported in column 10 of table 4, is .070 (SE = .019), while the AA subsample responds to a same-sex triple by more than twice as much. This again contrasts with the twins estimates, where first-stage effects are smaller in the AA subsample.

First-stage effects in the 3+ sample show large differences when stratified by both sex and ethnicity, as can be seen in columns 7–8 and 11–

¹⁸ This model is almost saturated in the sense that it controls for all lower-order interaction terms in the estimation of the effects of the two same-sex triples except for one: in the $(1 - s_{12i})b_{3i}$ term, we do not distinguish mixed-sibling pairs according to whether a boy or a girl was born first. A saturated model can be obtained by replacing the single term, $(1 - s_{12i})b_{3i}$, with two terms, $b_{1i}(1 - b_{2i})b_{3i}$ and $b_{2i}(1 - b_{1i})b_{3i}$. In practice, this substitution matters little.

12 of table 4. The overall effect of three girls on first and second borns is .183, almost triple the corresponding effect of three boys, .065. The effect of three girls is also much larger in the AA population. The estimate for non-AA in column 12 is .072, and the increment for AA is .217, so that the effect of three girls in the first- and second-born AA subsample is .29 (.26 for firstborns only). This is considerably larger than the twins effect on AA subjects in the 2+ sample.

D. Heterogeneity and Nonlinearity in Response to Sibling Sex Composition

The difference in first-stage effects by AA status documented in table 4 shows that the population of sex-composition compliers is disproportionately more likely to come from an AA background. This is especially true for the response to an all-girl sibship. For example, the two-girl effect on AA fertility in column 4 is .14, while the EA effect is about .09. The AA differential in the effects of sex composition on family size is largest for the response to same-sex triples. This pattern stands in marked contrast to the composition of twins compliers, among which the AA subsample is underrepresented. Thus, any comparison of twins and sex-composition IV estimates is implicitly a comparison for very different groups.

A second noteworthy distinction between the sex-composition and twins first stages is in the different ranges of effects traced out by the two types of instruments. As we noted above, the twins2 instrument in the 2+ sample increases family size from two to three with relatively little effect at higher parities, while twins3 in the 3+ sample primarily increases family size from three to four, with virtually no other impact on fertility. In contrast, a same-sex sibship leads some families to keep having children at higher parities in pursuit of a more balanced sex composition.

The fertility shift due to sex composition in the 2+ sample is documented in figure 2, which reports first-stage estimates of effects of b_{12i} and g_{12i} on $d_{ji} \equiv 1(c_i \geq j)$, for j up to 11, along with the associated confidence bands. In the AA population, b_{12i} increases the likelihood that families have three or more children, with no significant effects at higher-order births. In contrast, the effect of two girls on d_{ji} increases from $j = 3$ to $j = 4$ and then tails off gradually, with a marginally significant effect on the likelihood of having seven or more children. Effects in the non-AA population drop off more sharply as the number of children increases and are similar for two boys and for two girls. If anything, the non-AA population seems to increase childbearing more sharply in response to two boys than to two girls.

The CDF differences plotted in figure 2 imply that sex-composition instruments capture an average causal effect that reflects the impact of having as many as seven children in the AA population and as many as

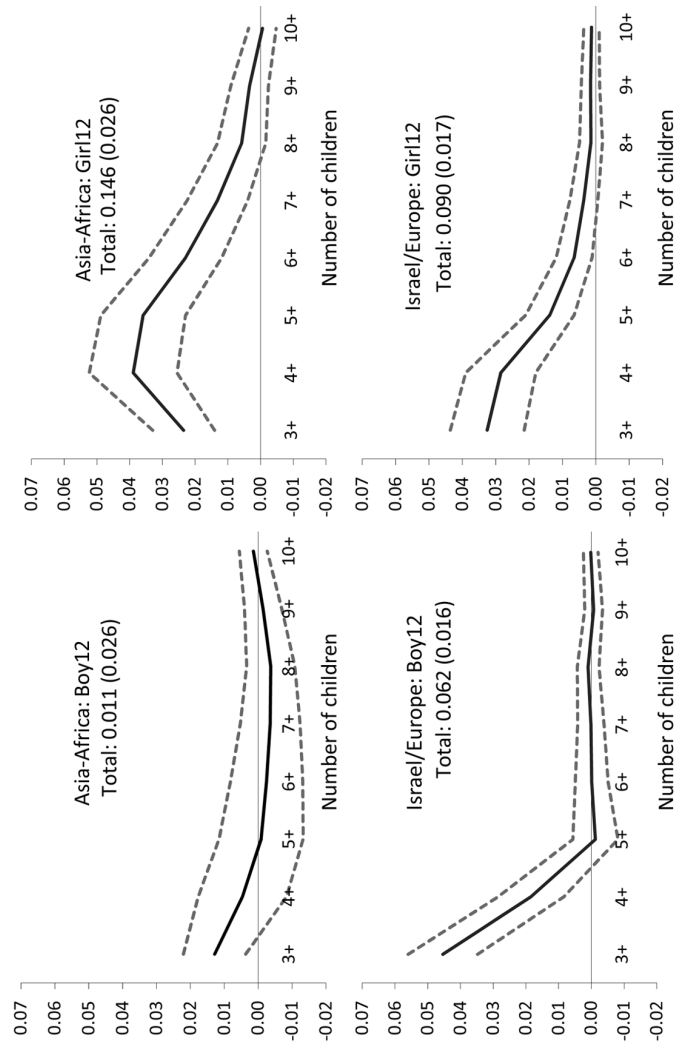


FIG. 2.—Firstborns in the 2+ sample: first-stage effects by ethnicity and type of sex mix. Dashed lines are confidence bands. Overall first stages and standard errors are shown in the plot area.

six children in the non-AA population. The range of fertility variation induced by sex composition is even wider in the 3+ sample. This can be seen in figure 3, which reports CDF differences in response to b_{123i} and g_{123i} along with the associated confidence bands. The figure shows that, in the AA population, b_{123i} increases the likelihood of having four or more children, with a small and marginally significant effect on the likelihood of having five or more children. The effect of three boys is similar in the AA and the non-AA populations. In contrast, the effect of three girls differs considerably by ethnicity, reaching .29 for three girls in the AA sample. Also in the AA population, the effect of g_{123i} increases from $j = 3$ to $j = 4$ and then diminishes gradually for higher values of j , remaining marginally significant even at $j = 10$. In the non-AA population, in contrast, the effect of g_{123i} is considerably smaller and differs little from the effect of b_{123i} .

E. The Boy3 Instrument

The bottom rows of columns 5–12 in table 4 show the effect of having a boy at third birth in families with a mixed-sex sibship at first and second birth. We expect the boy3 instrument to operate through preferences for male children that are common in more traditional Israeli households. In addition to providing additional variation, the boy3 instrument is useful because it is implicitly used only for families with a mixed-sex sibship at parities one and two. The boy3 instrument is therefore unlikely to be subject to the same violations of the exclusion restriction as instruments derived from sex mix alone.

A boy at third birth reduces childbearing in the families of first and second borns with a mixed-sex sibship by .077. Models allowing different coefficients by ethnicity generate an effect of $-.044$ in the non-AA population, while the AA interaction term generates a further .064 reduction. Figure 4 summarizes the effects of b_{3i} on fertility increments separately by ethnicity. The sample used to construct this figure includes both first and second borns.

Figure 4 shows that, as with the sex-mix instruments, boy3 affects fertility over a wider range than do multiple births. In the AA population in particular, b_{3i} reduces the likelihood of having more than four children, as well as the likelihood of higher-order births up to seven, beyond which the effect is no longer significant. In the non-AA population, however, b_{3i} reduces the likelihood of having four or more children, with no significant effect at higher-order births.

F. Instrument Validity

A possible concern in any IV study is correlation between the instruments and potential outcomes because of either confounding or violations of

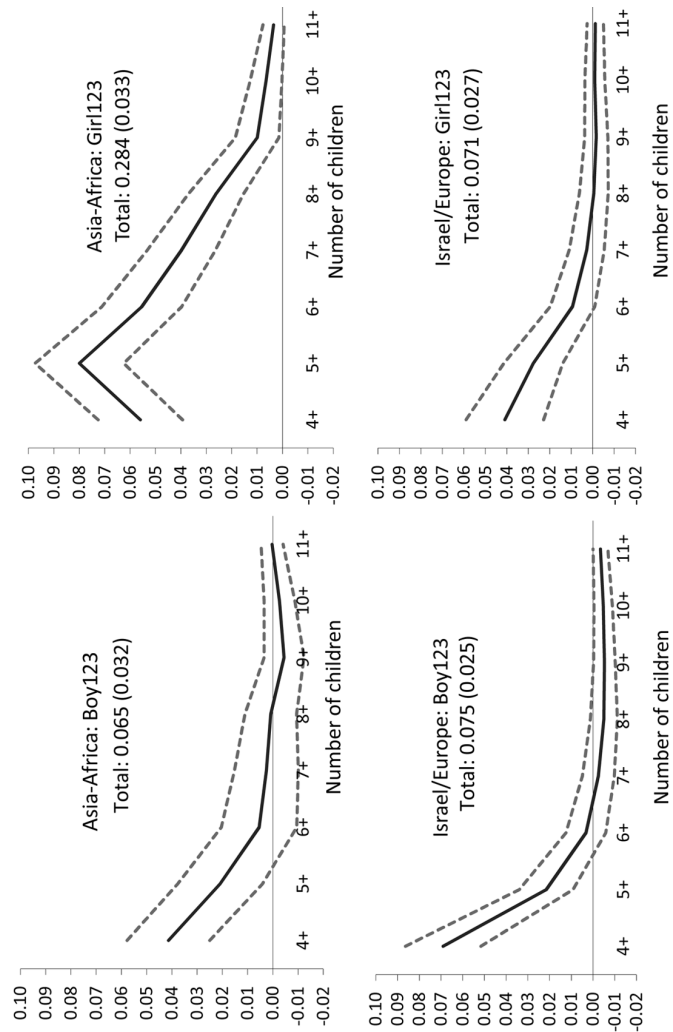


FIG. 3.—First and second borns in the 3+ sample: first-stage effects by ethnicity and type of sex mix. Dashed lines are confidence bands. Overall first stages and standard errors are shown in the plot area.

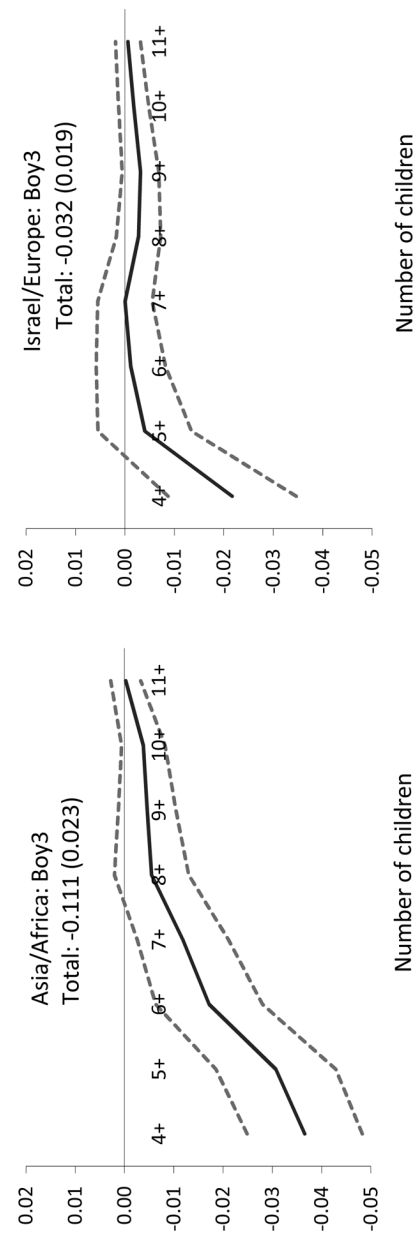


FIG. 4.—First and second borns in the 3+ sample: first-stage effects of boy3. Dashed lines are confidence bands. Overall first stages and standard errors are shown in the plot area.

the exclusion restriction. As in the Angrist and Evans (1998) study using sex-composition instruments, however, there is no relation between sex mix and any of the background variables or covariates in our matched data set (detailed results available on request). We also replicated the common finding that twin births are associated with older maternal age. For example, the mothers of first and second borns who had twins at second or third birth were .3–.5 years older at first birth than those who had singletons. Twinning is not otherwise associated with subject demographics, with one exception: in the 1995 sample of 2+ subjects, twin rates are higher for younger cohorts. Since twins can be identified only when birth records are complete, the fact that the quality of birth records improved over time seems likely to explain this finding. In any case, the 3+ sample does not exhibit this pattern. Because the results are similar in the 2+ and 3+ samples, the change in quality of birth records seems unlikely to have had a major impact on our findings.

It is also worth noting that multiple-birth-enhancing fertility treatments, a possible source of bias when using twins instruments, became available in Israel only in the mid-1970s. The effect of this on twin rates is first evident in vital statistics data starting in the mid-1980s (Blickstein and Baor 2004). Since fewer than 5% of the third-born siblings in our 3+ sample and fewer than 1% of second-born siblings in our 2+ sample were born after 1984, the spread of fertility treatments is unlikely to be a factor in our analysis.

A further concern with twins instruments, raised by Rosenzweig and Zhang (2009), is the possible violation of exclusion restrictions due to the fact that twins have lower average birth weight than singletons and, perhaps, worse health or cognitive achievement later on. Rosenzweig and Zhang argue that some parents therefore allocate resources away from twins and toward older singleton-birth children. Such parental behavior may offset any quantity-quality effects, making them harder to find using twins instruments to estimate effects on nontwins.

To see if resource reallocation is a problem, we estimated reduced-form twins effects on outcomes in samples in which twins have little effect on family size. If the Rosenzweig and Zhang household resource story is true, first- and second-born children who have younger twin siblings should come out better in these no-first-stage samples since they benefit from the resources shifted away from less promising twins, with no offsetting increase in family size. Households with little or no twins first stage include those likely to have large families anyway, such as mothers who gave birth early or spaced births closely. Columns 1–3 of table A2 therefore report estimated reduced-form twin effects in 2+ samples with young mothers (first birth before age 21), closely spaced births (less than 2 years), and AA ethnicity. The first-stage effect of twins2 in the young-mother and closely spaced samples are small and insignificant; the first-

stage effect of twins2 in the AA subsample is much smaller than in the complementary sample, although marginally significant at .177 (SE = .09).

The results presented in table A2 fail to support the claim that parents favor older children after a twin birth. There are no twins effects on the outcomes of the firstborn children of mothers who gave birth before 21, as can be seen in column 1 of the table. The reduced-form effects of twinning are also zero in samples stratified by birth spacing and ethnicity.¹⁹

The same-sex instruments might also violate the exclusion restriction, a possibility raised by Rosenzweig and Wolpin (2000). Specifically, Rosenzweig and Wolpin argue for pure sex-composition effects on family size due to household efficiencies in families with same-sex sibships. To check this, columns 4–6 of table A2 report on an investigation that parallels the no-first-stage investigation for twins instruments. The no-first-stage samples for sex composition are again defined as those with young mothers, from families with tight spacing, and with AA subjects, looking in all cases at firstborn boys in the 2+ sample. Because sex composition has no effect on family size in these subsamples, effects of confounding factors related to sex composition should therefore surface. Consistent with a causal interpretation of the sex-composition IV estimates, however, there is no reduced-form relation between the two-boy instrument and any outcome variable in any subsample.

IV. OLS and 2SLS Estimates

The causal effect of interest is the coefficient ρ in the model

$$y_i = W_i'\mu + \rho c_i + \varepsilon_i, \quad (6)$$

where y_i is an outcome variable and W_i includes the covariates X_i as well as instrument- and sample-specific controls (e.g., b_i). As discussed in the previous section, 2SLS estimates of this equation capture siblings' weighted average response to the birth of an additional child for those whose parents were induced to have an additional child by the instrument at hand. The outcome variables measure human capital, economic well-being, and social circumstances. In particular, we look at measures of subjects' educational attainment (highest grade completed and indicators of high school completion and college attendance), labor market status (indicators of work last year and hours worked last week), earnings (monthly earnings and the natural log of earnings for full-time workers), marital status (indicators of being married at census day and married by age 21), and fertility.

¹⁹ We also implemented a version of the twins/nontwins strategy discussed by Rosenzweig and Zhang (2009). This is discussed at the end of our results section.

A. The 2+ Sample

As is typical for regressions of this sort, OLS estimates of the coefficient on family size in equation (6) indicate a negative association between family size and measures of human capital and economic circumstances. Larger families are also associated with earlier marriage and increased fertility. These results can be seen in column 2 of table 5, which presents OLS estimates for firstborns in the 2+ sample (col. 1 reports the means). Not surprisingly given the sample sizes, all the OLS estimates are very precise. Control for covariates reduces but does not eliminate this negative relationship, as can be seen in column 3.

In contrast with the negative OLS estimates, 2SLS estimates point to zero or even positive effects. These results appear in columns 4–8 of table 5, which report 2SLS estimates using different sets of instruments. For example, the effect on schooling estimated using twins instruments with AA interaction terms, reported in column 5, is .105 (SE = .131). The corresponding estimate using sex-composition instruments with AA interaction terms, reported in column 7, is .222 (SE = .176).

To increase precision, we also estimated specifications that combine twins and sex-composition instruments within a given sample (in this case, 2+) to produce a single more efficient IV estimate. Although each instrument potentially generates its own local average treatment effect, the combination of instruments in this context can be justified by the desire to pin down what appears to be a common effect (of zero) as precisely as possible.

Combining twins and sex-composition instruments generates an estimate of .16 (SE = .106) for the effect on schooling, reported in table 5, column 8.²⁰ The combination of instruments generates a substantial gain in precision relative to the use of each instrument set separately: the schooling effect in the first row of column 8 is significantly different from the corresponding OLS estimate of $-.145$ reported in column 3. Likewise, the estimated effect on college attendance is small, positive, and reasonably precise.

This discussion highlights the fact that a key concern with the IV analysis is whether the estimates are precise enough to be informative. Of particular interest is the ability to distinguish IV estimates from the corresponding OLS benchmarks. As it turns out, the estimates constructed by pooling twins and sex-composition instruments with AA interaction terms, meet this standard of precision remarkably often. In particular, six out of seven estimates of effects on nonmarriage and fertility outcomes presented in this column are estimated precisely enough that the associated 95% confidence interval excludes the corresponding OLS estimates re-

²⁰ The combined first stages are reported in the appendix.

Table 5
Estimates for Firstborns in the 2+ Sample

	OLS			2SLS Instrument List				
	Mean (1)	Basic Covariates (2)	All Covariates (3)	Twins (4)	TwinsAA (5)	Girl12, Boy12AA (6)	Girl12, Boy12, Girl12AA, Boy12AA (7)	All (8)
Schooling:								
Highest grade completed	12.6	-.252 (.005)	-.145 (.005)	.174 (.166)	.105 (.131)	.294 (.184)	.222 (.176)	.160 (.106)
Years of schooling ≥ 12	.824	-.037 (.001)	-.029 (.001)	.030 (.028)	.024 (.021)	-.009 (.028)	-.015 (.028)	.007 (.017)
Some college (age ≥ 24)	.291	-.049 (.001)	-.023 (.001)	.017 (.052)	.026 (.046)	.089 (.048)	.089 (.046)	.057 (.032)
College graduate (age ≥ 24)	.202	-.036 (.001)	-.015 (.001)	-.021 (.045)	-.006 (.041)	.115 (.046)	.115 (.044)	.054 (.028)
Labor market outcomes (age ≥ 22):								
Worked during the year	.827	-.025 (.001)	-.024 (.001)	-.005 (.038)	.002 (.033)	.062 (.043)	.072 (.043)	.034 (.026)
Hours worked last week	32.6	-1.06 (.05)	-1.20 (.06)	-.97 (2.58)	.00 (2.18)	1.46 (2.06)	1.06 (1.98)	.51 (1.45)
Monthly earnings (1995 shekels)	2,997	-217.0 (7.4)	-179.1 (8.0)	-7.7 (394.1)	73.0 (324.5)	266.7 (283.6)	429.1 (292.1)	264.1 (214.2)
ln(earnings) for full-time workers	8.24	-.045 (.002)	-.028 (.002)	-.082 (.116)	.125 (.100)	.120 (.124)	.180 (.215)	.435 (3.852)
Marriage and fertility:								
Married on census day	.446	.023 (.001)	.020 (.001)	.043 (.029)	.060 (.025)	.118 (.034)	.101 (.032)	.078 (.020)
Married by age 21 (age ≥ 21)	.172	.027 (.001)	.022 (.001)	-.006 (.037)	.024 (.032)	.197 (.047)	.192 (.046)	.110 (.026)
Any children	.448	.029 (.001)	.019 (.001)	.090 (.056)	.013 (.036)	.135 (.041)	.134 (.041)	.079 (.026)

NOTE.—Means of dependent variables are in col. 1, and ordinary least squares (OLS) estimates of the coefficient on family size are in cols. 2–3. Two-stage least squares (2SLS) estimates using different sets of instruments appear in cols. 4–8. Instruments with an “AA” suffix are interaction terms with an Asia-Africa dummy. Sample includes firstborns from families with two or more births (2+) as described in table 1. OLS estimates for col. 2 include controls for age and sex. Estimates for cols. 3–8 are from models that include the control variables specified in table 3. Robust standard errors are reported in parentheses.

ported in column 3. Moreover, most of the estimates of effects on schooling variables are very close to zero. A few of the estimated effects on college attendance are significant and positive, although, given the large number of reported effects, this may be a chance finding.

A second set of noteworthy results are those for marriage and fertility. The IV estimates of effects on marital status suggest that subjects from larger families are more likely to be married and married sooner. Using both twins and sex-composition instruments, the estimated effects on marital status are significantly different from zero and substantially larger than the corresponding OLS estimates. The marriage effects generated by sex-composition instruments are larger than the twins estimates, a point we return to below.

The marriage effects are paralleled by (and are perhaps the cause of) an increase in fertility: the combination IV estimate of the effect on the probability of having any children is .079, four times larger than the corresponding OLS estimate, .019. In addition to the likelihood that increased marriage rates increase fertility, these fertility effects may reflect an intergenerational causal link in preferences over family size, a possibility suggested by Fernandez and Fogli (2005).²¹

B. The 3+ Sample

Estimates in the 3+ sample, reported in table 6, are broadly similar to those for the 2+ sample, although there are some noteworthy differences. Columns 2–6 in table 6 parallel columns 4–8 in table 5 in that they report results from a similar sequence of instrument lists, with the modification that the twins instruments were generated by the event of a multiple third birth, and the sex-composition instruments are dummies for same-sex triples. A further change in table 6 is the addition of column 7, which reports results combining all instruments (with AA interaction terms) and a dummy for boy3 (also with an AA interaction term). This addition provides a modest further gain in precision.

The OLS results in tables 5 and 6 are virtually identical. The 2SLS estimates in the 3+ sample exploit more sources of variation than were used to construct estimates in the 2+ sample, so here we might expect some differences. The key finding, however, remains: 2SLS estimates using both twins and sibling sex composition generate no evidence of an adverse effect of larger family size on human capital or labor market variables. Moreover, as in table 5, a few of the estimated effects on schooling out-

²¹ To see whether the earnings effects are driven by the fact that many subjects are in their early twenties, we reestimated the models in table 5, restricting the sample to those at least age 30. The results based on this restricted sample, which are not reported here but are available from the authors, are virtually identical to those using the full sample.

Table 6
Estimates for First and Second Borns in the 3+ Sample

	2SLS Instrument List						
	OLS, All Covariates (1)	Twins (2)	TwinsAA (3)	Girl123, Boy123 (4)	Girl123AA, Boy123AA (5)	All (6)	All, Boy3, Boy3AA, TwinsAA, Boy3, TwinsAA, Boy3, Boy3AA (8)
Schooling:							
Highest grade completed	-.143 (.005)	.167 (.117)	.187 (.110)	-.116 (.134)	-.067 (.120)	.069 (.080)	.072 (.076)
Years of schooling ≥ 12	-.031 (.001)	.024 (.019)	.025 (.018)	.000 (.023)	-.009 (.022)	.009 (.014)	.016 (.017)
Some college (age ≥ 24)	-.021 (.001)	.059 (.036)	.060 (.036)	-.051 (.031)	-.025 (.025)	.011 (.021)	.029 (.027)
College graduate (age ≥ 24)	-.014 (.001)	.052 (.032)	.055 (.032)	-.060 (.028)	-.032 (.022)	.004 (.018)	.030 (.024)
Labor market outcomes (age ≥ 22):							
Worked during the year	-.027 (.001)	.029 (.025)	.034 (.024)	.033 (.029)	.031 (.027)	.032 (.018)	.035 (.017)
Hours worked last week	-1.40 (.05)	2.35 (1.45)	2.51 (1.43)	1.31 (1.36)	1.44 (1.28)	1.94 (.94)	2.08 (1.23)
Monthly earnings (1995 shekels)	-184.5 (6.8)	47.2 (204.1)	63.9 (203.8)	118.8 (176.5)	109.2 (162.2)	90.0 (128.7)	76.3 (175.7)
ln(earnings) for full-time workers	-.030 (.002)	-.010 (.047)	-.016 (.047)	.021 (.069)	.053 (.061)	.014 (.039)	-.021 (.071)
Marriage and fertility:							
Married on census day	.020 (.001)	.021 (.018)	.015 (.018)	.024 (.024)	.045 (.022)	.029 (.014)	.022 (.016)
Married by age 21 (age ≥ 21)	.023 (.001)	.029 (.022)	.028 (.021)	.046 (.029)	.052 (.028)	.039 (.017)	.036 (.019)
Any children	.022 (.001)	.021 (.028)	.015 (.026)	-.002 (.025)	.016 (.023)	.015 (.017)	.010 (.023)

NOTE.—Ordinary least squares (OLS) estimates of the coefficient on family size are in col. 1. Two-stage least squares (2SLS) estimates using different sets of instruments appear in cols. 2–8. Instruments with an “AA” suffix are interaction terms with an Asia-Africa dummy. The sample includes first and second borns from families with three or more births (3+) as described in table 2. Regression estimates are from models that include the control variables specified in table 3. Robust standard errors are reported in parentheses and are clustered by mother’s ID.

comes are positive and (marginally) significant, although the significant estimates are fewer and smaller in this case. The marriage effects in the 3+ sample are also smaller and less consistently significant than in the 2+ sample. In particular, the twins instruments generate no significant marriage estimates when used alone, although they are still positive. Likewise, there are no longer any significant fertility effects.

As a check on the exclusion restrictions for sex-composition instruments, we also looked at estimates omitting these instruments but retaining boy3. These results, reported in table 6, column 8, again provide no evidence of any adverse effects of family size. In general, same-sex instruments appear to generate smaller 2SLS estimates (i.e., closer to zero or less likely to be positive) than do twins instruments or the combination of twins with boy3. This is inconsistent with Rosenzweig and Wolpin's (2000) conjecture regarding possible beneficial effects of having a sibling of the same sex. The boy3 instrument may also have direct effects, as suggested by Butcher and Case (1994) for girls, but others have found little evidence for this (e.g., Kaestner 1997).

Interpreting Average Causal Response

The results in tables 5 and 6 are largely consistent across instruments, samples, and subjects' birth order. This is important because, as shown in the previous section, different instruments shift the fertility distribution very differently for different ethnicities. Moreover, sex-composition instruments shift fertility over a wide range of parities, with substantial shifts in large families, especially for the AA sample. Twins instruments, by contrast, increase completed fertility close to the parity at which a multiple birth occurred. The twins and sex-composition IV estimates therefore capture the effects of different fertility increments. A related point is that the fertility shifts induced by both sets of instruments are over different ranges in the 2+ and 3+ samples. Finally, we might expect different types of instruments to have different omitted variables biases, if any. Overidentification tests generate a formal measure of the equality of a set of IV estimates in models with multiple instruments (see, e.g., Angrist 1991). Although not reported here in detail, the overidentification tests for the 2SLS estimates in tables 5 and 6 generate no evidence of significant differences across instrument sets.

Because the effects of a family size on older siblings might differ at different ages (perhaps because parental investments before a fertility shock are unaffected by the shock), it is also noteworthy that multiple birth and sex-composition experiments expose children to an increase in family size at a wide range of ages. For example, firstborn children in the 2+ sample were about 7 years old on average when a singleton third child was born but only 4 years old upon the arrival of a twin. Similarly,

firstborn children in the 3+ sample were about 9.5 years old when a singleton fourth child was born but only 7.75 years old when the fourth born was a twin. On average, firstborn children exposed to a parity-six singleton birth were about 12 years old at the time. We also observe significant ethnic variation in age of exposure due to tighter birth spacing in AA families. Of course, as noted in the introduction, we have no direct evidence on the effects of family size on the last child born. Nevertheless, the consistency of our results across widely ranging parities and ages of exposure weighs against substantially heterogeneous effects by birth order.

The Rosenzweig and Zhang Bounding Strategy

As noted in the discussion of instrument validity, Rosenzweig and Zhang (2009) argue that lower average birth weight may induce some parents to allocate household resources away from twins and toward older singleton-birth children. Such parental behavior could offset any quantity-quality effects, making them harder to find in studies using twins instruments to estimate effects on older nontwin siblings. As discussed above, our direct investigation of the exclusion restriction generated no evidence of this behavior. Nevertheless, we explore the Rosenzweig and Zhang argument further here.

Rosenzweig and Zhang (2009) specifically argue that estimates of the effect of twins2 on firstborns underestimate the quantity-quality trade-off. To avoid this bias, they suggest that comparisons of twins and nontwins at parity two be taken as an upper bound on the magnitude of the quantity-quality effect.²² Comparisons of twins and nontwins tend to overestimate any negative effects of larger family size because twins have lower average birth weight than nontwins and may differ in other ways. Rosenzweig and Zhang therefore also suggest that when looking at the impact of twins on nontwins, it is useful to control for birth weight as a measure of twin quality, although control for birth weight is problematic because birth weight is an endogenous variable that is itself affected by twinning. We do not have data on birth weight, but we can compare twins and nontwins in the spirit of Rosenzweig and Zhang's suggestion that such comparisons provide an upper bound on quantity-quality trade-offs.

We implemented the bounding approach by comparing twin and nontwin outcomes for second- and third-born individuals using regression models similar to those used to produce our OLS and 2SLS estimates. As before, these regressions control for gender, age, missing month of birth, mother's age, mother's age at first birth, mother's age at immigra-

²² "The effect of twinning at the second pregnancy on the outcomes of second- (first-) birth children provides an upper (lower) bound on the average negative effect on child outcomes of increasing family size" (Rosenzweig and Zhang 2009, 1157).

tion, father's and mother's place of birth, and census year. The regression-adjusted twin/nontwin comparisons show no significant differences between twins and singletons for outcome variables related to schooling, earnings, or labor supply.²³ Thus, the sort of contrasts seen by Rosenzweig and Zhang as bounding the size of the causal effect of interest also produce an estimate of zero in our data.²⁴

C. Combining 2+, 3+, 4+, and 5+ Samples

To further increase precision, we also pooled estimates across the 2+, 3+, and two higher-parity samples. For example, we constructed a single twins IV estimate using t_{2i} , t_{3i} , t_{4i} , and t_{5i} as instruments in a data set that stacks 2+, 3+, 4+, and 5+ samples while restricting the IV estimates from the different parity-specific subsamples to be the same. Because the instrument list and conditioning variables are different in each parity-specific subsample, this procedure requires a modification of conventional 2SLS.

The Parity-Pooled Setup

Our pooled analysis works with the union of subjects from 2+, 3+, 4+, and 5+ subsamples. The total sample therefore includes individuals who are firstborn subjects in the 2+ sample, first- and second-born subjects in the 3+ sample, first-through-third-born subjects in the 4+ sample, and first-through-fourth-born subjects in the 5+ sample. In other words, the sample includes all birth orders up to $p - 1$ from families with at least p children, for $p \leq 5$. The $p+$ subsamples are not mutually exclusive; for example, a given firstborn subject in the 5+ sample must also be a member of the 2+, 3+, and 4+ subsamples.

Pooled estimation can be motivated by assuming that the causal effect of childbearing is a constant, denoted ρ_0 (tables 5 and 6 suggest $\rho_0 = 0$). In terms of potential outcomes, we have

$$Y_i(j) = Y_i(0) + \rho_0 j. \quad (7)$$

In addition, let $Y_i(0) = X_i' \mu_0 + v_i$ denote the regression of $Y_i(0)$ on X_i in the population from which the parity-pooled sample is drawn. The residual, v_i , is orthogonal to X_i in this population by construction. The observed outcome, y_i , is linked to this causal model by

$$y_i = X_i' \mu_0 + \rho_0 c_i + v_i. \quad (8)$$

Note that the residual, v_i , may be correlated with c_i .

²³ Detailed tables showing twin/nontwin contrasts are available on request.

²⁴ Twins usually have lower birth weight (LBW) than singletons, but the question of whether this matters for adult outcomes remains controversial (see, e.g., Behrman and Rosenzweig 2004; Almond, Chay, and Lee 2005; Conley, Strully, and Bennett 2006; Black, Devereux, and Salvanes 2007; Royer 2009).

The following lemma provides the econometric justification for pooled estimation:

LEMMA. Let d_{pi} denote membership in a $p+$ sample, and let Z_{pi} denote an IV satisfying $Z_{pi} \perp Y_i(0) | W_{pi}$, $d_{pi} = 1$, where W_{pi} includes X_i plus possibly additional instrument-specific controls. Let $Z_{pi}^* = Z_{pi} - W_{pi}'\Gamma$, where Γ is the coefficient vector from a regression of Z_{pi} on W_{pi} in the $p+$ population. Assume there is a first stage for Z_{pi} , that is, $E(Z_{pi}^*c_i | d_{pi} = 1) \neq 0$. Then $E(d_{pi}Z_{pi}^*v_i) = 0$, where v_i is the error term in equation (8), and the expectation is taken in the population containing subjects of birth order up to $p - 1$ from families with at least p children.

Proof. $E(d_{pi}Z_{pi}^*v_i) = E[d_{pi}Z_{pi}^*(y_i - X_i'\mu_0 - \rho_0c_i)] = E[Z_{pi}^*(y_i - X_i'\mu_0 - \rho_0c_i) | d_{pi} = 1]P(d_{pi} = 1)$. Note that $E(Z_{pi}^*X_i | d_{pi} = 1) = 0$ by construction. Given the constant-effects causal model, equation (7), and the conditional independence assumption at the beginning of the lemma, $\rho_0 = E(Z_{pi}^*y_i | d_{pi} = 1) / E(Z_{pi}^*c_i | d_{pi} = 1)$. Therefore, $E[Z_{pi}^*(y_i - \rho_0c_i) | d_{pi} = 1] = 0$. QED

This lemma shows how a common causal parameter can be estimated in a parity-pooled sample. For example, we can combine t_{2i} in the 2+ sample and t_{3i} in the 3+ sample. The data set required for this is the union of the 2+ and 3+ samples, that is, firstborns in the 2+ sample and second borns in the 3+ sample (firstborns in the 3+ sample are included in the 2+ sample). After partialing out the relevant set of covariates as described in the lemma, $d_{2i}t_{2i}^*$ and $d_{3i}t_{3i}^*$ are valid instruments for equation (8) in the pooled sample. Similarly, we can combine $d_{2i}b_{12i}^*$, $d_{2i}g_{12i}^*$, $d_{3i}b_{123i}^*$, $d_{3i}g_{123i}^*$ and $d_{3i}[(1 - s_{12i})b_{3i}]^*$, where the first-step regression adjustment of each instrument accounts for the fact that sex-composition instruments involve different sets of controls in the 2+ and 3+ sample, in addition to the set of common covariates, X_i .

Before turning to a discussion of parity-pooled empirical results, we briefly discuss first-stage estimates in higher-parity samples, focusing on the sex-composition instruments. Figures 5 and 6 report the effects of sex composition on fertility in the 4+ sample, using a format similar to the one used in figures 3 and 4. The figures document the fact that sex mix sharply increases family size in this sample. Effects are again larger for all-female than for all-male sibships and for the AA population. In the AA samples, an all-girl sibship increases the likelihood of family sizes as large as nine. A full set of first-stage estimates is given in table A3. The largest first-stage effect for sex mix is .365 (.242 + .123) among the AA population, a result of five girls. However, an all-male sibship still increases fertility in both the 4+ and the 5+ samples. The effect of a multiple fourth birth, reported in column 5, is almost one child for non-AA Jews in the 4+ sample. For this group, the twins experiment amounts to a randomized trial with perfect compliance.

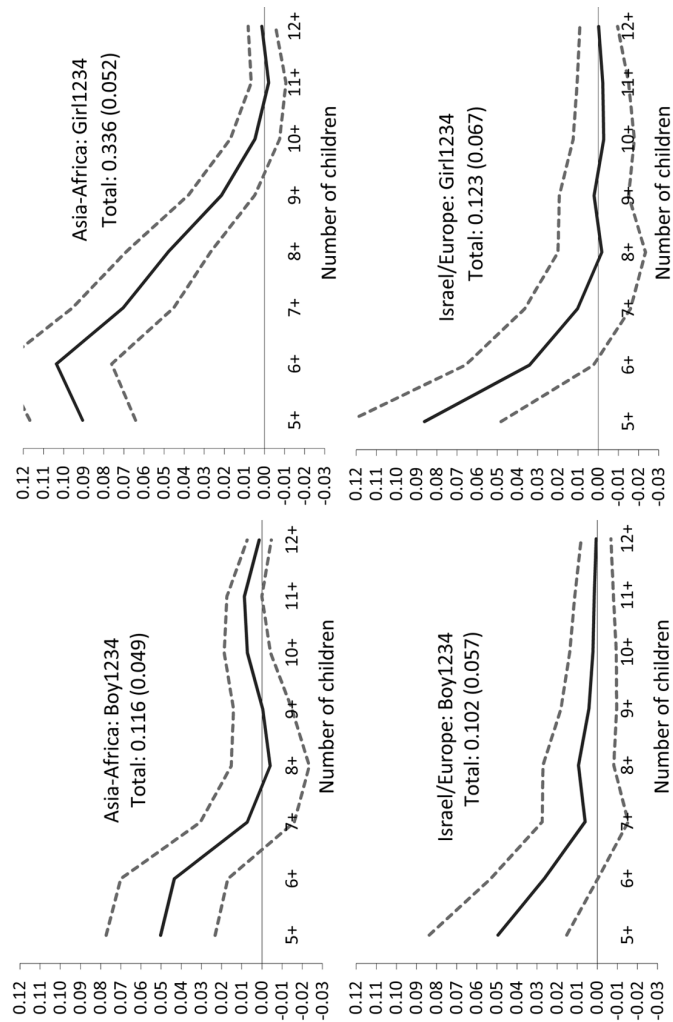


FIG. 5.—First, second, and third borns in the 4+ sample: first-stage effects by ethnicity and type of sex mix. Dashed lines are confidence bands. Overall first stages and standard errors are shown in the plot area.

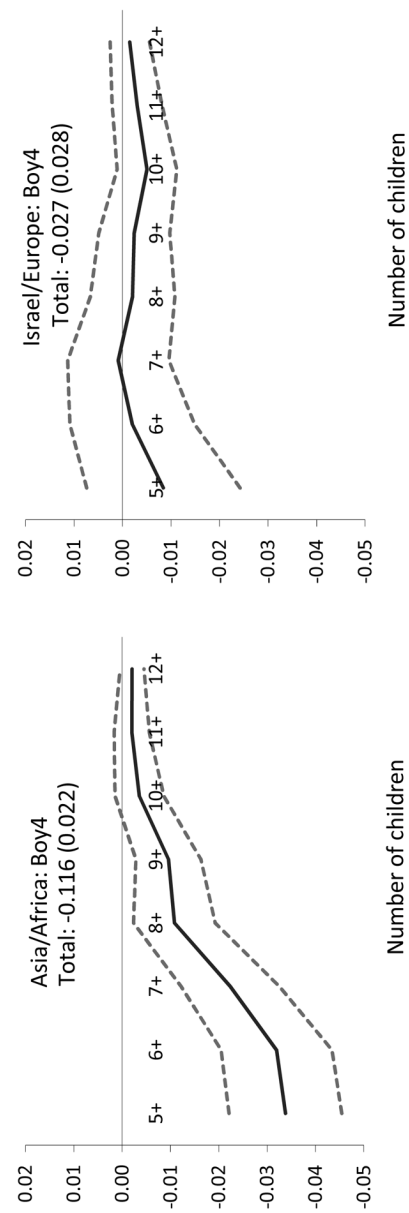


FIG. 6.—First, second, and third borns in the 4+ sample: first-stage effects of boy4. Dashed lines are confidence bands. Overall first stages and standard errors are shown in the plot area.

Parity-Pooled Results

The empirical strategy using parity-pooled samples leads to a considerable gain in precision, while most of the estimated effects on outcomes other than marriage and fertility remain small and insignificant. This can be seen in table 7, which reports pooled results using twins instruments in columns 1–3, results pooling sex-composition instruments in columns 4–6, and the results of pooling all instruments in columns 7–9. The table shows results from three samples for each instrument set: the union of subjects from the 2+ and 3+ subsamples; the union of subjects from 2+, 3+, and 4+ subsamples; and the union of subjects from 2+, 3+, 4+, and 5+ subsamples. For example, the estimated effect on highest grade completed using all available twins instruments in the union of the 2+, 3+, 4+, and 5+ samples is .031 (SE = .055), shown in the first row of table 7. The corresponding estimate using all available sex-composition instruments is .054 (SE = .068), in column 6.

The estimates combining both twins and sex-composition instruments in the union of 2+, 3+, 4+, and 5+ samples, reported in column 9 of table 7, are the most precise we have been able to construct. For example, the estimated effect on highest grade completed is .040 (SE = .043), in comparison with .072 (SE = .076) in table 6. Similarly, the estimated effect on annual employment is .005 (SE = .009), compared to .035 (SE = .017) reported in table 6. All estimates of effects on nonmarriage and fertility outcomes in column 9 of table 7 generate confidence intervals that exclude the corresponding OLS estimates with covariates.

Most of the parity-pooled estimates of effects on marriage and some of the estimated effects on fertility remain at least marginally significantly different from zero. For example, the estimated effect on marriage using the twins instrument in the pooled 2+, 3+, 4+, and 5+ sample is .023 (SE = .010), and the corresponding estimate using sex-composition instruments is .045 (SE = .012). While sex-composition instruments generate larger effects on marriage than do the twins instruments, the fact that this effect turns up in both IV strategies suggests the IV estimates reflect the causal effect of childbearing and not just a propensity for older girls to marry in response to the birth of a younger sister, a point discussed further in the next section.

D. Analyses by Ethnicity and Gender

Large numbers of Sephardic Jews came to Israel from the Arab countries of Asia and North Africa in the 1950s. Although fertility among Sephardic Jews ultimately fell to close to the Israeli average, the AA cohorts in our sample come from much larger families than other Jews. While almost 60% of AA Jews in our 2+ sample come from families with four or more

Table 7
Estimates for Parity-Pooled Samples

	Twins Instruments			Sex Composition Instruments			All Instruments		
	2+, 3+ (1)	2+, 3+, 4+ (2)	2+, 3+, 4+, 5+ (3)	2+, 3+ (4)	2+, 3+, 4+ (5)	2+, 3+, 4+, 5+ (6)	2+, 3+ (7)	2+, 3+, 4+ (8)	2+, 3+, 4+, 5+ (9)
Schooling:									
Highest grade completed	.141 (.079)	.060 (.059)	.031 (.055)	.036 (.093)	.036 (.074)	.054 (.068)	.092 (.060)	.050 (.046)	.040 (.043)
Years of schooling ≥ 12	.022 (.013)	.002 (.011)	-.006 (.010)	-.013 (.016)	-.015 (.014)	-.010 (.013)	.006 (.010)	-.005 (.008)	-.008 (.008)
Some college (age ≥ 24)	.039 (.026)	.010 (.016)	.004 (.014)	.002 (.019)	.009 (.015)	.002 (.014)	.017 (.016)	.010 (.011)	.003 (.010)
College graduate (age ≥ 24)	.024 (.023)	.012 (.014)	.011 (.012)	.007 (.017)	.011 (.013)	.007 (.012)	.014 (.014)	.012 (.010)	.010 (.009)
Labor market outcomes (age ≥ 22):									
Worked during the year	.016 (.018)	-.007 (.014)	-.001 (.013)	.044 (.020)	.004 (.016)	.011 (.014)	.031 (.013)	-.002 (.010)	.005 (.009)
Hours worked last week	1.31 (1.09)	.37 (.76)	.66 (.69)	1.25 (.98)	-.41 (.77)	.11 (.71)	1.27 (.72)	.00 (.54)	.40 (.49)
Monthly earnings (1995 shekels)	31.8 (158.2)	-37.4 (121.8)	36.6 (107.1)	187.5 (132.5)	48.8 (97.5)	57.2 (88.8)	118.7 (101.6)	5.3 (78.6)	47.0 (70.2)
ln(earnings) for full-time workers	-.022 (.230)	-.007 (.032)	-.003 (.029)	.024 (.057)	.012 (.033)	.014 (.031)	.015 (.063)	-.004 (.023)	.006 (.021)
Marriage and fertility:									
Married on census day	.030 (.014)	.029 (.010)	.023 (.010)	.046 (.017)	.053 (.013)	.045 (.012)	.037 (.010)	.039 (.008)	.033 (.007)
Married by age 21 (age ≥ 21)	.024 (.016)	.023 (.012)	.020 (.012)	.078 (.021)	.058 (.017)	.052 (.015)	.052 (.013)	.039 (.010)	.035 (.009)
Any children	.019 (.020)	.017 (.014)	.021 (.013)	.040 (.018)	.049 (.015)	.042 (.013)	.031 (.013)	.033 (.010)	.032 (.009)

NOTE.—This table shows two-stage least squares estimates of fertility effects using twins, sex composition, and all available instruments in parity-pooled samples. Robust standard errors are reported in parentheses and are clustered by mother's ID.

children, only 26% of other Jews in the sample come from families this large.

In addition to having higher fertility, the AA group is less educated and poorer than other Jewish ethnic groups. For example, only 12% of AA Jews in our 2+ sample are college graduates, while the overall college graduation rate in the 2+ sample is 20%. The gap in living standards by ethnicity is especially big in larger households. Among those born in Israel, the average 1990 income in AA households with five or more members was about 60% of the income of similarly sized EA households, only 15% larger than the income of non-Jews (Central Bureau of Statistics 1992, table 11.4). These differences suggest that estimates in the AA sub-sample may be especially relevant for poorer populations.

OLS estimates by ethnicity, reported in columns 1 and 3 of table 8, generally show somewhat larger adverse effects on schooling and labor market outcomes in the non-AA sample than in the AA sample. The estimates in table 8 are for the full parity-pooled sample including the union of subjects from 2+, 3+, 4+, and 5+ families, and the 2SLS estimates use the full set of instruments. The resulting 2SLS estimates by ethnicity, reported in columns 2 and 4, generate no evidence of an effect on human capital or labor market variables for either ethnic group. For example, the estimated effect on highest grade completed in the non-AA sample is .043 (SE = .064), while the corresponding estimate for AA is .031 (SE = .057). The estimated effect on hours worked is .30 (SE = .87) for non-AA subjects and .45 (SE = .59) for AA.

As in the sample that does not differentiate by ethnicity, there is again evidence for an effect of family size on marriage rates or timing in both ethnic groups. For example, the estimated effects on marriage are .030 (SE = .011) for non-AA subjects and .035 (SE = .010) for AA subjects. Effects on early marriage are similar in the two groups. The effects on fertility are also positive in both samples and are slightly larger for the AA population.

Also of interest are separate estimates for men and women, especially in view of the effects on marital status discussed above. We therefore estimated separate models by sex, using the full set of instruments in the largest parity-pooled sample, with results reported in columns 6 and 8 of table 8. The OLS estimates reported in columns 5 and 7 are similar for men and women. Again, however, 2SLS estimates by sex show no evidence of negative effects on schooling or labor market variables for either group. For example, the estimated effects on log earnings are .009 (SE = .031) for men and .015 (SE = .026) for women.

2SLS estimates of effects on marriage rates are more pronounced for women than for men and more precise. For example, the effect on women, reported in table 8, column 8, is .04 (SE = .009), while the corresponding effect for men, reported in column 6, is .021 (SE = .011). Moreover, the

Table 8
Estimates by Ethnicity and Sex in the Largest Parity-Pooled Sample (2+, 3+, 4+, 5+)

	By Ethnicity				By Sex			
	Israel-Europe		Asia-Africa		Males		Females	
	OLS (1)	2SLS (2)	OLS (3)	2SLS (4)	OLS (5)	2SLS (6)	OLS (7)	2SLS (8)
Schooling:								
Highest grade completed	-.117 (.006)	.043 (.064)	-.117 (.005)	.031 (.057)	-.150 (.005)	.078 (.073)	-.096 (.005)	.020 (.050)
Years of schooling ≥ 12	-.029 (.001)	.000 (.011)	-.023 (.001)	-.014 (.011)	-.033 (.001)	.002 (.014)	-.019 (.001)	-.011 (.009)
Some college (age ≥ 24)	-.033 (.001)	.010 (.019)	-.014 (.001)	-.001 (.011)	-.017 (.001)	.010 (.016)	-.019 (.001)	.000 (.012)
College graduate (age ≥ 24)	-.021 (.001)	.019 (.017)	-.010 (.001)	.003 (.009)	-.011 (.001)	.021 (.014)	-.013 (.001)	.006 (.011)
Labor market outcomes (age ≥ 22):								
Worked during the year	-.039 (.001)	-.007 (.016)	-.014 (.001)	.011 (.012)	-.024 (.001)	.013 (.014)	-.021 (.001)	.000 (.012)
Hours worked last week	-2.23 (.07)	.30 (.87)	-.67 (.05)	.45 (.59)	-1.31 (.06)	.56 (.91)	-1.01 (.05)	.23 (.55)
Monthly earnings (1995 shekels)	-239.1 (8.6)	119.6 (147.0)	-120.5 (6.5)	7.0 (74.2)	-219.6 (8.7)	65.5 (162.1)	-109.1 (5.4)	33.2 (59.6)
ln(earnings) for full-time workers	-.030 (.004)	.072 (.037)	-.025 (.002)	-.032 (.025)	-.027 (.002)	.009 (.031)	-.026 (.003)	.015 (.026)
Marriage and fertility:								
Married on census day	.034 (.001)	.030 (.011)	.010 (.001)	.035 (.010)	.016 (.001)	.021 (.011)	.022 (.001)	.040 (.009)
Married by age 21 (age ≥ 21)	.039 (.001)	.034 (.014)	.014 (.001)	.038 (.012)	.013 (.001)	-.012 (.010)	.027 (.001)	.065 (.013)
Any children	.034 (.002)	.024 (.016)	.015 (.001)	.038 (.012)020 (.001)	.032 (.009)

NOTE.—This table shows ordinary least squares (OLS) and two-stage least squares (2SLS) results from models estimated separately by ethnicity and sex. The 2SLS estimates use the full set of instruments (i.e., corresponding to col. 9 in table 7). Robust standard errors are reported in parentheses and are clustered by mother's ID.

estimated effect on early marriage for women is about 6–7 percentage points and significantly different from zero. In contrast, the corresponding estimate for men is negative and insignificant.

The consistency and relative precision of results across instrument sets suggests that early marriage may indeed be a consequence of increased family size, especially for older daughters. The marriage effects seem to generate a small effect on fertility as well (also apparent in table 7). Stronger marriage effects for women may reflect the fact that marriage is the main route to an independent household for girls in traditional Jewish families. Moreover, older daughters in Israel may be tempted to marry sooner when crowded by younger sisters. This is consistent with traditional Jewish values and can be traced back to the biblical story of Rachel and Leah's joint betrothal to Jacob. We might therefore expect marriage effects estimated using sex-composition instruments to be larger than effects estimated using twins instruments, as in table 7.

V. Possible Explanations

Exogenous increases in family size in a Becker-Lewis model (due, say, to a change in contraceptive costs) reduce child quality since an increase in quantity increases the shadow price of quality. Along these lines, Rosenzweig and Wolpin (1980) interpret twin births as a subsidy to the cost of further childbearing. They argue that this price change should reduce quality unless quantity and quality are strong complements in parental utility functions. While the quantity-quality trade-off is less clear-cut in more recent theoretical discussions, the traditional view provides an intellectual foundation for policies that attempt to reduce family size in less developed countries.

The most important question our findings raise is what might account for the absence of a causal link between sibship size and later outcomes. A definitive answer to this question must await future empirical research. Here, we briefly review a number of possible explanations. One theoretical possibility is that, as far as investment in human capital goes, parents use perfect capital markets to fund investment irrespective of resource constraints. It seems unlikely, however, that capital markets are so nearly perfect, especially in Israel during the period we are studying, when financial markets were not well developed.²⁵

A more relevant possibility is that parents adjust to exogenous increases in family size on margins other than quality inputs. For example, parents may work longer hours or take fewer or less expensive vacations (i.e.,

²⁵ Another theoretical possibility, outlined by Deaton and Paxson (1998), is that larger households are better off at the same level of per capita expenditure due to household scale economies. This seems unlikely to explain our results since we are investigating effects without holding per capita resources constant.

consume less leisure). Parents may also substitute away from personal as opposed to family consumption (e.g., by drinking less alcohol). Direct evidence on this point is difficult to obtain since consumption data rarely come in the form needed to replicate our research design.

The Angrist and Evans (1998) results for wives raise the possibility of an explanation linked to female labor supply. Clearly one effect of additional childbearing is to increase the likelihood of at-home child care for older siblings (an effect also documented by Gelbach 2002). It may be that home care is better, on average, than commercial or other out-of-home care, at least in the families affected by the fertility shocks we study. However, estimates of Angrist-and-Evans-type models for samples of Israeli mothers show only modest effects of childbearing on labor supply (Marmer 2000).

On the institutional side, the quantity-quality trade-off within households may be partially offset by welfare payments and public schooling. If so, this may limit the external validity of our findings or at least their applicability to countries with a less developed welfare state. However, the Israeli setting is especially interesting because different cohorts were exposed to different institutions. The Israeli Compulsory Schooling Law enacted in 1949 allowed 9 years of free and compulsory education starting from kindergarten until eighth grade. In 1969, the law was changed to provide 2 additional years of free and compulsory education, until tenth grade. In 1978, a further extension provided free (although not compulsory) schooling for grades 11 and 12. When our subjects were children, school enrollment of Jewish children below age 14 was about 95%, whereas school enrollment for those age 14–17 increased from 67% in 1970 to 80% in 1980 (Central Bureau of Statistics 1996). Despite this sharp increase in educational attainment, we find no significant cross-cohort differences in IV estimates of family size effects.

As in many developed and middle-income countries, Israel offers tax concessions to larger families in the form of child allowances. But these payments were low during the period when members of our samples were young (Manski and Mayshar 2003) and therefore seem unlikely to explain the absence of a quantity-quality trade-off. We confirmed this in a 2SLS analysis introducing interaction terms for changes in eligibility for child allowances and the level of child allowances by cohort.

An additional explanation for the absence of a causal link between sibship size and the outcomes studied here might be called “marginally irrelevant inputs.” Using research designs similar to ours, Caceres (2006) finds some evidence for a decreased likelihood of private school enrollment. However, private school attendance in Israel may matter little for human capital and earnings. A final explanation consistent with our findings is that the presence of siblings directly enhances child welfare, perhaps because children with siblings benefit socially or take on more respon-

sibility sooner. This conjecture is consistent with Qian's (2009) IV estimates for China, which show that the presence of a younger sibling increases older children's school enrollment.

VI. Summary and Directions for Further Work

We studied the causal link running from sibship size to human capital, economic well-being, and family structure using a unique sample combining population registry and census data. Our research design exploits variation in fertility due to multiple births and preferences for a mixed-sibling sex composition, along with ethnicity interactions and preferences for male children. The natural experiments embodied in these IV strategies capture a wide range of fertility variation.

The evidence reported here is remarkably consistent across research designs and samples: while all instruments exhibit a strong first-stage relation and OLS estimates are substantial and negative, IV estimation generates no evidence for negative consequences of increased sibship size on outcomes. The estimates do suggest, however, that girls from larger families marry sooner. This marriage effect may have a modest impact on fertility, but it does not appear to reduce schooling, employment, or earnings. In future work, we hope to shed light on possible explanations for these findings by generating new evidence on the effect of family size on resource allocation across generations.

Appendix

The Israeli population registry, our source of information on families of origin, contains updated administrative records for Israeli citizens and residents, whether currently living or dead, including most Israelis who have moved abroad. This database also includes the Israeli ID numbers held by citizens and temporary residents. ID numbers are issued at birth for the native born and on arrival for immigrants. In addition to basic demographic information on individuals (date of birth, sex, country of birth, year of immigration, marital status, religion, and nationality), the registry records parents' names and registrants' parents' ID numbers.

The construction of our analysis file proceeded by first using subjects' ID numbers to link to non-public-use versions of census long-form files that include ID numbers with registry records for as many subjects as we could find. In a second step, we used the registry to find subjects' mothers. Finally, once mothers were linked to census respondents, we then located all the mothers' children in the registry, whether or not these children appear in the census. In this manner, we were able to observe the sex and birth dates of most adult census respondents' siblings.

The likelihood of successful matches at each stage of our linkage effort is determined primarily by the inherent coverage limitations of the reg-

istry. Israel's population registry was first developed in 1948, not long after the creation of the state of Israel. Census enumerators went from house to house, simultaneously collecting information for the first census and for the administrative system that became the registry. Later, the registry was updated using vital statistics data. Thus, in principle, the sample of respondents available for a census interview in 1983 and 1995 should appear in the registry, along with their mothers' ID numbers, if they were resident in 1948, born in Israel after 1948, or immigrated to Israel after 1948.

To assess whether the limitations in the matching process outlined above introduce a bias that might affect our estimates, we constructed a reverse match starting with the registry and going forward to the censuses. The reverse match includes Jews in the registry alive on the 1983 or 1995 census dates. We applied the same sample restrictions related to mother's year of birth and year of immigration, mother's age at first birth, and subject's year of birth used to construct the main extract. Table A1 reports the effect of family size and the instruments on the probability of appearing in either the 1983 or 1995 census data using the reverse match. Each cell reports estimates from a separate regression. Estimates reported in even columns come from regressions that control for an indicator for age at census date, missing month of birth, mother's age at census date, mother's age at first birth, mother's age at immigration, and father's and mother's place of birth. Columns 1–4 report estimates for firstborns in 2+ families; columns 5–8 and 9–12 report estimates from a sample of 3+ families, firstborns and second borns, respectively. This table suggests match rates are unrelated to the instruments. As noted in the text, there is a small decrease in the 1983 match rates with family size, but this is unlikely to affect our empirical strategy. We also reproduced the first-stage regressions using the entire registry population alive on the census date and the reverse-match sample. Both samples produced first-stage estimates virtually identical to those reported in the article.²⁶

²⁶ A table with first-stage estimates based on these two samples is available on request.

Table A1
Relationship between Family Structure and Selection into Census Data

	Two or More Births				Firstborns				Three or More Births			
	Firstborns		1995 Census		1983 Census		1995 Census		1983 Census		1995 Census	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
No. of children	-.0008 (.0005)	-.0025 (.0006)	.0008 (.0003)	.0005 (.0004)	-.0012 (.0006)	-.0032 (.0006)	-.0004 (.0004)	-.0006 (.0004)	-.0015 (.0006)	-.0034 (.0007)	.0002 (.0004)	.0005 (.0005)
Subject = boy	.0003 (.0020)	-.0002 (.0020)	-.0034 (.0012)	-.0034 (.0012)	.0004 (.0023)	-.0001 (.0022)	-.0032 (.0014)	-.0031 (.0014)	-.0001 (.0026)	-.0002 (.0026)	-.0052 (.0016)	-.0052 (.0016)
Twins	.0108 (.0118)	.0110 (.0117)	-.0037 (.0062)	-.0037 (.0062)	-.0062 (.0112)	-.0054 (.0112)	-.0080 (.0068)	-.0083 (.0068)	-.0138 (.0134)	-.0131 (.0133)	-.0002 (.0076)	.0001 (.0076)
Same sex	-.0013 (.0020)	-.0014 (.0019)	-.0007 (.0012)	-.0008 (.0012)	-.0011 (.0026)	-.0010 (.0026)	.0014 (.0016)	.0014 (.0016)	.0000 (.0030)	.0003 (.0030)	-.0003 (.0018)	-.0002 (.0018)
All boys	.0009 (.0022)	.0002 (.0022)	-.0030 (.0014)	-.0031 (.0014)	.0016 (.0032)	.0015 (.0032)	-.0007 (.0020)	-.0007 (.0020)	.0009 (.0037)	.0007 (.0037)	-.0018 (.0022)	-.0019 (.0022)
All girls	-.0028 (.0023)	-.0023 (.0023)	.0022 (.0014)	.0022 (.0014)	-.0039 (.0035)	-.0037 (.0035)	.0034 (.0022)	.0034 (.0022)	-.0011 (.0041)	-.0002 (.0041)	.0016 (.0024)	.0018 (.0024)
Boy at last birth (mixed sex in earlier births)0030 (.0026)	.0030 (.0026)	-.0015 (.0016)	-.0016 (.0016)	.0024 (.0030)	.0019 (.0030)	.0010 (.0018)	.0010 (.0018)
Full controls	No	Yes	No	Yes	No	Yes	No	Yes	No	Yes	No	Yes
N	135,568		392,504		102,349		283,573		75,272		235,421	

NOTE.—Effects of family size and the instruments on the probability of appearing in the census data are shown. Each cell reports estimates from a separate regression. Samples include all Jewish individuals registered at the population registry who were alive at the census date. Estimates reported in even columns come from regressions that control for indicators for age at census date, missing month of birth, mother's age at census date, mother's age at first birth, mother's age at immigration (when relevant), and father's and mother's place of birth (i.e., full controls). Robust standard errors are reported in parentheses.

Table A2
Reduced Form Effects in No-First-Stage Samples

	Twins2 Instrument		Boy12 Instrument	
	Mother's Age at First Birth < 21 (1)	Spacing between First and Second Birth < 2 (2)	Mother's Age at First Birth < 21 (4)	Spacing between First and Second Birth < 2 (5)
		Asia-Africa Ethnicity (3)		Asia-Africa Ethnicity (6)
Effect on family size: ^a				
No. of children	.067 (.126)	.132 (.156)	.017 (.033)	-.033 (.047)
Effect on outcomes of firstborns: [†]				
Schooling:				
Highest grade completed	.094 (.136)	-.162 (.147)	.049 (.041)	.027 (.053)
Years of schooling ≥ 12	.023 (.031)	-.012 (.038)	.008 (.008)	.008 (.010)
Some college (age ≥ 24)	.042 (.045)	-.092 (.045)	.011 (.008)	.000 (.012)
College graduate (age ≥ 24)	-.029 (.032)	-.084 (.031)	.003 (.007)	.002 (.010)
Labor market outcomes (age ≥ 22):				
Worked during the year	.011 (.035)	-.004 (.049)	-.002 (.007)	-.003 (.010)
Hours worked last week	.370 (2.013)	-.451 (2.973)	.059 (.456)	.119 (.652)
Monthly earnings (in 1995 shekels)	-420.5 (341.7)	-283.1 (284.8)	-9.0 (83.2)	-61.0 (95.3)
ln(earnings) for full-time workers	-.020 (.098)	-.059 (.115)	.003 (.016)	-.007 (.022)
Marriage and fertility:				
Married on census day	.025 (.027)	.020 (.035)	.006 (.006)	.009 (.008)
Married by age 21 (age ≥ 21)	-.034 (.034)	-.061 (.041)	.001 (.006)	.010 (.008)
Any children	.008 (.038)	.027 (.059)

NOTE. — The sample includes firstborns from families with two or more births. Regression estimates are from models that include the control variables specified in table 3. Robust standard errors are reported in parentheses.
^aFirst-stage effects of twins2 and boy12 on number of children for different subsamples.
[†]Reduced-form effects of twins2 and boy12 on the outcomes listed below.

Table A3
Pooled First Stage

	Two or More Births		Three or More Births		Four or More Births		Five or More Births	
	Firstborns (1)	Firstborns (2)	First and Second Borns (3)	Firstborns (4)	First, Second, and Third Borns (5)	Firstborns (6)	First, Second, Third, and Fourth Borns (7)	
Twins	.637 (.057)	.594 (.045)	.698 (.049)	.879 (.114)	.998 (.110)	.629 (.232)	.714 (.162)	
Twins × Asia-Africa	-.482 (.105)	-.138 (.094)	-.228 (.086)	-.135 (.148)	-.291 (.137)	-.052 (.268)	-.159 (.196)	
All girls	.090 (.017)	.102 (.032)	.079 (.027)	.158 (.083)	.127 (.068)	.147 (.212)	.242 (.161)	
All girls × Asia-Africa	.052 (.032)	.165 (.051)	.212 (.043)	.202 (.105)	.212 (.085)	.173 (.245)	.123 (.180)	
All boys	.061 (.017)	.103 (.029)	.077 (.025)	.096 (.073)	.102 (.058)	.489 (.218)	.307 (.155)	
All boys × Asia-Africa	-.045 (.031)	-.011 (.047)	-.015 (.041)	.067 (.095)	.007 (.076)	-.535 (.247)	-.336 (.176)	
Boy at last birth (mixed sex in earlier births)	...	-.048 (.023)	-.038 (.019)	-.024 (.036)	-.024 (.029)	.039 (.067)	.031 (.048)	
Boy at last birth × Asia-Africa (mixed sex in earlier births)	...	-.057 (.035)	-.067 (.030)	-.085 (.045)	-.087 (.036)	-.112 (.075)	-.126 (.054)	
Subject = boy	.013 (.017)	.018 (.023)	.006 (.015)	.021 (.045)	.025 (.024)	-.023 (.090)	.002 (.039)	
Subject = boy × Asia-Africa	.007 (.031)	-.005 (.035)	.016 (.022)	.005 (.055)	.010 (.028)	.054 (.100)	.018 (.043)	
Asia-Africa	.242 (.024)	.189 (.032)	.103 (.026)	-.001 (.056)	-.054 (.041)	-.276 (.111)	-.253 (.072)	

NOTE.—First-stage effects on number of children using the full set of instruments are shown. Regression estimates are from models that include the control variables specified in table 3. Regressions for cols. 4–7 also include controls for twins and sex composition at lower-parity births. Regressions for columns 3, 5, and 7 also include controls for birth order and spacing. Robust standard errors are reported in parentheses. Standard errors in columns 3, 5, and 7 are clustered by mother's ID.

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