# Motherhood and Female Labor Supply in the Developing World <br> Evidence from Infertility Shocks 

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#### Abstract

We introduce a new instrument for family size, infertility, to investigate the causal relationship between children and female labor force participation. Infertility mimics an experiment where nature assigns an upper bound for family size, independent of a woman's background. This new instrument allows us to investigate the differential labor supply without restrictions on initial family size. Using the Demographic and Health Surveys from 26 developing countries we show that OLS estimates are biased upward. We find that the presence of children affects neither the likelihood of work nor its intensity, but impacts the type of work a woman pursues.


## I. Introduction

The increasing representation of women in the global labor force has been one of the most remarkable labor market trends of recent times. Never before have so many women been economically active: Worldwide, the female labor force was 1.2 billion women in 2003 (International Labor Office 2004). The last 40 years also have witnessed a dramatic global decline in fertility in the developing world. According to the United Nations, the total fertility rate in the less-developed regions of the world is 2.75 children per woman in 2005-2010, down from 5.41 children per women in 1970-75 (United Nations, World Population Prospects 2007). A neg-

[^0]ative correlation between the presence of children and female participation in the labor force is well established both over time and across countries and many scholars have posited a causal relationship between these series. ${ }^{1}$

However, the interpretation of the relationship between family size and mother's work is complicated by the endogeneity of fertility. The number of children a woman has is a choice variable that is influenced by her labor force participation. Additionally, unobserved factors are likely to influence both fertility and female labor force participation. For instance, more independent women may choose to have fewer (or no) children and these women may also be overrepresented in the labor force. Thus, the observed negative relationship between children and labor force participation could be biased. The amount of bias may vary by stages of development as the necessity of female employment decreases with economic development. Thus, if poorer women have fewer choices, OLS estimates for this group should suffer from less bias.

To clearly identify the relationship between children and labor force participation an exogenous source of variation in family size is needed. In this paper, we focus on a new source of variation in family size based on biological events. In particular, we use infertility shocks as an instrument for family size to identify the causal effect of children on female labor force participation. Clearly, infertility affects the number of children a woman can have. In our sample, infertile or subfecund women report 1.2 fewer children than their fertile counterparts. In addition, we show that background characteristics of women appear to be unrelated to infertility. Thus, an indicator variable for the infertility status of women of childbearing age is a plausible instrument for family size.

Previous strategies that use twinning (Rosenzweig and Wolpin 1980) or sex composition of the first two children (Angrist and Evans 1998) as instruments for family size restrict the sample to women who have had at least one or two births, respectively. Papers that use these approaches are limited to the effect of having an additional child on female labor force participation conditional on reaching a specific parity. An advantage of our instrument, in contrast, is the possibility to identify the causal effect of children on female labor force participation irrespective of the number of existing children, and therefore it represents a broader sample of women. In addition we estimate a different Local Average Treatment Effect (LATE) than the one associated with sex mix or twinning. The women whose family size is altered by the sex composition of their children or a twin birth are women with low underlying desired fertility, whereas infertility "removes" children from women with high underlying desired fertility.

The use of biological events such as infertility to understand the causal role of family size on female labor force participation is understudied in the developing

[^1]world, although the use of these types of events is prevalent in the United States. ${ }^{2}$ Furthermore, most of the existing empirical work in developing countries focuses on a particular country or region of the world. For example, Cruces and Galiani (2007) use the sex composition of the first two children to identify the relationship between family size and labor force participation in Mexico and Argentina. They find that women who are induced to have a third child, out of a desire for a balanced sex mix of their offspring, are less likely to participate in the paid labor force. The use of sex composition has been criticized by Schultz (2008) and others because the gender of children could have a wealth effect in countries where dowries are required. Cáceres-Delpiano (2009) is the only paper we are aware of that contains a global sample of women and exogenous variation in family size at the micro level. He exploits the variation in family size due to twinning on the first birth, which causes some families (those with low desired fertility) to have larger families than they would otherwise desire. He finds heterogeneous effects depending on the parity affected and the margin of analysis. ${ }^{3}$

Agüero and Marks (2008) focus on Latin America and find no causal relationship between family size and female labor force participation. We employ a similar strategy to this paper but improve it along several dimensions. First, our work reflects a more globally representative sample, including 26 countries as opposed to six. This allows us to investigate heterogeneous responses by the level of development, which is not possible using Latin American countries alone. We also use a broader set of outcome variables. Agüero and Marks (2008) study participation in paid work only, while in this paper we explore overall participation (paid and unpaid) as well as the intensity of work. We also consider carefully the possibility of measurement error in the self-reported infertility indicator and present a wider set of robustness tests. Finally, unlike Agüero and Marks (2008), we also investigate the effect of family size on the labor force participation of younger women (aged 35 or younger) and find a significant negative effect in their paid labor force participation.

Our data set includes 90,965 women in low-income and mid-income countries. OLS estimates suggest an additional child reduces the overall (paid or unpaid) female labor force participation. Our main finding is that, after using the infertility instrument, children have a much smaller causal effect on the overall labor force participation of women. This finding is robust to the inclusion of health indicators, alternative definitions of family size, and when the outcome explores work intensity. However, the results suggest that children affect negatively the likelihood of working for cash for young women and also for those living in very low-income countries.

[^2]
## II. Infertility, Data and Methods

## A. Data

The medical literature defines infertility (or subfecundity) as the failure to conceive after a year of regular intercourse without contraception. Infertility can be further broken down into primary infertility, which describes women who have never been able to conceive a pregnancy, and secondary infertility, describing those who have had at least one successful pregnancy, but have not been able to achieve another.

In this paper we use cross-sectional data from the third round of the Demographic and Health Surveys (DHS III) collected between 1994 and 1999. The DHS are standardized nationally representative household surveys in developing countries. Women answered questions about their employment status, birth history, current and future contraceptive use, fertility preferences, and their socioeconomic and marital status. In some countries, anthropometric measures including height and weight were collected.

The DHS III allow us to identify self-reported infertility in two ways from two separate questions asked to all women. The first way is when women mentioned subfertility or infertility as their reason for not currently using contraceptives (Infertile 1). The second way is when nonsterilized women responded that they are unable to have more children when asked about their desire for future children (Infertile 2). Our main infertility indicator is the union of these two measures: $\max =\{$ Infertility 1, Infertility 2\}. We define labor force participation as a variable that takes the value of one if a woman reported working at all (paid or unpaid) in the last 12 months and zero otherwise. As additional outcome variables we will narrow the definition of work. The first classifies a woman as working if she was paid in cash in the last 12 months, and the second if she works year round.

Not all countries with a DHS III could be included in the analysis. To be included a survey had to meet the following criteria: (1) the survey had to include questions that were used to identify infertile women and these questions needed to be asked to the entire sample of women; (2) the infertility questions needed to include infertile, subfertile, subfecund, or unable to get pregnant as a standalone response; (3) the survey had to contain information about participation in the labor force and intensity of work in a consistent manner; (4) the data had to be publicly available. Our final sample contains 27 surveys representing 26 countries. In some specifications, we will split the sample by income levels (mid-income versus low-income) to investigate whether there are heterogeneous effects by stage of development. Appendix Table A1 contains additional information about the countries included in our sample.

## B. Sample Construction

Our main sample contains 90,965 women between the ages of 20 and 44 . We exclude from the sample students, women with missing labor force information, and women who have never had a sexual encounter. In keeping with the medical definition of infertility, we can only identify infertility for nonsterilized women who are not currently taking contraceptives; those who do not take contraceptives constitute more
than 60 percent of the women surveyed in the DHS III. ${ }^{4}$ Columns 1 and 2 of Table 1 compare our sample to an equivalent sample that includes women who are actively controlling their fertility (denoted by "Full Sample"). Women who actively control their fertility are more likely to live in urban areas and are more educated. However, the two groups of women are similar in age and have similar family sizes and work behavior.

Summary statistics for our sample are contained in Column 2 of Table 1. We also present information divided by income level of the country according to the World Bank's World Development Indicators. The average woman in our sample is 31 years old with 2.3 children, and 6 percent report being infertile. ${ }^{5} 64$ percent of our sample worked either for pay or in kind in the last 12 months, 49 percent of our sample worked for pay and 40 percent worked year round. Women in the more developed countries are much more educated, more likely to reside in an urban area, desire much smaller families, have smaller families, and are more likely to report infertility.

## C. Estimation Strategy

To assess the causal effect of family size on mothers' labor force participation, we use infertility as an IV for number of children in the household.

For the sample described above, the main specification is given by Model 1:

$$
\begin{equation*}
L F P_{i}=\alpha+\beta K_{i}+\sum_{j} \gamma_{j} A G E_{j i}+X_{i}^{\prime} \delta+e_{i} \tag{1}
\end{equation*}
$$

where $L F P_{i}$ is equal to one if the $i$ th woman is in the labor force and zero otherwise. The key variable $K_{i}$ captures the number/presence of children living at home. Thus, $\beta$ is the parameter of interest. Because there is a nonlinear relationship between age and infertility for women, we include a flexible functional form of age by including 24 binary variables (one per age in years and indexed by $j$ ) in all specifications. Vector $X_{i}$ includes survey fixed effects in Model 1. Model 2 adds to Model 1 other variables that may influence female labor force participation, such as education, age and education interactions, age at first intercourse, marital status, age at first marriage, and spouse's education. Model 3 contains all of the variables in Model 2 plus an indicator of health status. Finally, in all specifications the standard errors are clustered at subnational levels (for example, departments, provinces, or districts depending on the country).

OLS estimates of $\beta$ are likely to be biased due to unobserved variables in $e_{i}$. The direction of bias is given by two elements: the relationship between the omitted variable and the outcome variable $\left(L F P_{i}\right)$, and its relationship with the variable of interest $\left(K_{i}\right)$. In particular, consider the case where female autonomy influences labor force participation. If autonomy correlates positively with the outcome variable and negatively with the number of children, excluding this variable from Equation 1 biases the OLS estimates upward since part of the estimated effect of children on labor force participation can be attributed to the lack of female female autonomy.

[^3]Table 1
Descriptive Statistics

| Variable | Full Sample ${ }^{\text {a }}$ | Our Sample | Midincome countries ${ }^{\text {b }}$ | Lowincome countries ${ }^{\text {b }}$ |
| :---: | :---: | :---: | :---: | :---: |
| Infertile (1 or 2) | 0.04 | 0.06 | 0.08 | 0.05 |
|  | (0.19) | (0.24) | (0.27) | (0.22) |
| Infertile 1 | 0.02 | 0.03 | 0.05 | 0.03 |
|  | (0.14) | (0.18) | (0.21) | (0.16) |
| Infertile 2 | 0.02 | 0.03 | 0.04 | 0.03 |
|  | (0.15) | (0.18) | (0.20) | (0.17) |
| Number of children | 2.45 | 2.34 | 2.17 | 2.43 |
|  | (1.81) | (1.90) | (1.88) | (1.90) |
| Has children | 0.87 | 0.83 | 0.81 | 0.84 |
|  | (0.34) | (0.38) | (0.39) | (0.37) |
| Daughter preference ${ }^{\text {c }}$ | 0.17 | 0.16 | 0.18 | 0.15 |
|  | (0.37) | (0.37) | (0.38) | (0.35) |
| Number of children desired | 4.22 | 4.87 | 2.96 | 6.00 |
|  | (4.22) | (2.99) | (1.69) | (3.01) |
| Worked in the last 12 months | 0.64 | 0.64 | 0.59 | 0.67 |
|  | (0.48) | (0.48) | (0.49) | (0.47) |
| Worked for a wage | 0.52 | 0.49 | 0.51 | 0.47 |
|  | (0.50) | (0.50) | (0.50) | (0.50) |
| Worked full year | 0.43 | 0.40 | 0.45 | 0.38 |
|  | (0.50) | (0.49) | (0.50) | (0.48) |
| Age | 31.09 | 30.57 | 31.32 | 30.16 |
|  | (6.84) | (6.97) | (7.08) | (6.88) |
| Rural childhood ${ }^{\text {d }}$ | 0.46 | 0.52 | 0.40 | 0.59 |
|  | (0.50) | (0.50) | (0.49) | (0.49) |
| Above primary education | 0.40 | 0.28 | 0.50 | 0.16 |
|  | (0.49) | (0.45) | (0.50) | (0.37) |
| Age at first intercourse ${ }^{\text {e }}$ | 17.54 | 17.13 | 18.59 | 16.25 |
|  | (3.56) | (3.58) | (3.98) | (2.98) |
| Currently married | 0.89 | 0.88 | 0.83 | 0.90 |
|  | (0.31) | (0.33) | (0.37) | (0.30) |
| Age when first married | 18.54 | 18.03 | 19.84 | 17.09 |
|  | (4.17) | (4.19) | (4.55) | (3.66) |
| Urban | 0.49 | 0.39 | 0.58 | 0.28 |
|  | (0.50) | (0.49) | (0.49) | (0.45) |
| Visited clinic in last 12 months ${ }^{\text {f }}$ | 0.52 | 0.46 | 0.50 | 0.44 |
|  | (0.50) | (0.50) | (0.50) | (0.50) |
| Observations | 149,539 | 90,965 | 32,474 | 58,491 |

[^4]We will use infertility to instrument for $K_{i}$ in Equation 1 to address the endogeneity concern.

## III. Estimation Issues

## A. Endogenous Instruments

Fertility is highly heterogeneous across couples. Determining the sources of such heterogeneity among couples remains a challenge for fertility research (Weinberg and Dunson 2000). Below, we summarize some findings from the medical literature.

It is well established that infertility increases with a woman's age (see Dunson, Baird, and Colombo 2004 and Buck et al. 1997). However, the medical literature is not in agreement about what other factors, if any, influence infertility. For example, there is some evidence suggesting that previous use of birth control, as well as indicators of poor health such as sexually transmitted diseases, smoking, drinking, extreme body mass index (BMI), and miscarriages are associated with infertility (see Augood, Duckitt, and Templeton 1998; Gesink Law et al. 2007; Grodstein, Goldman, and Cramer 1994; Hassan and Killick 2005). However, the majority of this evidence is suspect as it comes from couples recruited for prospective studies (Negro-Vilar 1993). That is, couples experiencing fertility problems are recruited for a study and their observable characteristics are then correlated with their time to pregnancy. ${ }^{6}$ This type of study design has been shown to produce spurious associations (Juul, Keiding, and Tvede 2000).

There is evidence that infertility appears to be independent of the background characteristics of infertile women. For example, variables such as father's social class and parity have been shown to be unrelated to observed heterogeneity in fertility (Joffe and Barnes 2000). In an article summarizing the epidemiological literature regarding the role of lifestyle factors (cigarette smoking, alcohol and caffeine consumption, exercise, BMI, and drug use) on female infertility, Buck et al. (1997) conclude that " $[\mathrm{f}]$ ew risk factors have been assessed or identified for secondary infertility." Also, education, occupation, and race have been shown to be unrelated with impaired fecundity (Wilcox and Mosher 1993; Chandra 1994) using U.S. data from the National Survey of Family Growth.

We present new evidence that infertility is not correlated with "predetermined" or background characteristics of women. Infertility is a valid instrument if it is unrelated to omitted variables that influence labor force participation. Table 2 presents evidence on the validity of our instrument. This table reports coefficient estimates for our fertility measures from a series of regressions (indexed by $V_{i}$ ) that, in addition to fertility status, control for age nonparametrically as follows:

$$
\begin{equation*}
V_{i}=\theta_{1} \text { Infertile }_{i}+\theta_{2}\left(1-\text { Infertile }_{i}\right)+\sum_{j} \rho_{j} A G E_{j i}+\eta_{i} \tag{2}
\end{equation*}
$$

[^5]Table 2
Women's Characteristics by Fertility Status

|  | Women's Characteristics ( $V_{i}$ ) | Infertile ( $q_{1}$ ) | Fertile ( $q_{2}$ ) | Test $q_{1}-q_{2}=0$ |
| :---: | :---: | :---: | :---: | :---: |
| Panel A | Number of children | $\begin{gathered} -0.347 \\ (0.041) \end{gathered}$ | $\begin{gathered} 0.960 \\ (0.055) \end{gathered}$ | $\begin{gathered} -1.307 * * \\ {[-25.99]} \end{gathered}$ |
|  | Has children | $\begin{gathered} 0.439 \\ (0.024) \end{gathered}$ | $\begin{gathered} 0.669 \\ (0.020) \end{gathered}$ | $\begin{gathered} -0.230^{* *} \\ {[-18.79]} \end{gathered}$ |
| Panel B | Worked in the last 12 months | $\begin{gathered} 0.542 \\ (0.032) \end{gathered}$ | $\begin{gathered} 0.532 \\ (0.036) \end{gathered}$ | $\begin{array}{r} 0.009 \\ {[0.70]} \end{array}$ |
|  | Worked for a wage | $\begin{gathered} 0.399 \\ (0.028) \end{gathered}$ | $\begin{gathered} 0.350 \\ (0.032) \end{gathered}$ | $\begin{aligned} & 0.048 * * \\ & {[4.50]} \end{aligned}$ |
|  | Worked full year | $\begin{gathered} 0.300 \\ (0.032) \end{gathered}$ | $\begin{gathered} 0.271 \\ (0.036) \end{gathered}$ | $\begin{gathered} 0.029 \\ {[1.71]} \end{gathered}$ |
| Panel C | Number of siblings ${ }^{\text {a }}$ | $\begin{gathered} 5.60 \\ (0.263) \end{gathered}$ | $\begin{gathered} 5.51 \\ (0.217) \end{gathered}$ | $\begin{gathered} 0.090 \\ {[0.74]} \end{gathered}$ |
| Panel D | Age at first intercourse ${ }^{\text {b }}$ | $\begin{aligned} & 15.90 \\ & (0.259) \end{aligned}$ | $\begin{aligned} & 15.86 \\ & (0.151) \end{aligned}$ | $\begin{gathered} 0.044 \\ {[0.20]} \end{gathered}$ |
|  | Month of birth | $\begin{aligned} & 6.15 \\ & (0.188) \end{aligned}$ | $\begin{aligned} & 6.07 \\ & (0.220) \end{aligned}$ | $\begin{gathered} 0.080 \\ {[0.91]} \end{gathered}$ |
|  | Birth order ${ }^{\text {a }}$ | $\begin{aligned} & 2.72 \\ & (0.104) \end{aligned}$ | $\begin{gathered} 2.65 \\ (0.093) \end{gathered}$ | $\begin{gathered} 0.067 \\ {[1.18]} \end{gathered}$ |
|  | Rural childhood ${ }^{\text {c }}$ | $\begin{gathered} 0.549 \\ (0.047) \end{gathered}$ | $\begin{gathered} 0.645 \\ (0.042) \end{gathered}$ | $\begin{gathered} -0.095^{* *} \\ {[-4.72]} \end{gathered}$ |
|  | Height (in cm) ${ }^{\text {d }}$ | $\begin{aligned} & 156.1 \\ & (0.957) \end{aligned}$ | $\begin{gathered} 155.5 \\ (1.28) \end{gathered}$ | $\begin{array}{r} 0.625 \\ {[1.20]} \end{array}$ |
|  | Daughter preference ${ }^{\text {e }}$ | $\begin{gathered} 0.167 \\ (0.021) \end{gathered}$ | $\begin{gathered} 0.152 \\ (0.018) \end{gathered}$ | $\begin{array}{r} 0.015 \\ {[1.12]} \end{array}$ |
|  | Number of children desired | $\begin{gathered} 4.21 \\ (0.466) \end{gathered}$ | $\begin{gathered} 4.88 \\ (0.540) \end{gathered}$ | $\begin{gathered} -0.662 * * \\ {[-3.57]} \end{gathered}$ |
| Panel E | BMI overweight ${ }^{\text {d }}$ | $\begin{gathered} 0.165 \\ (0.028) \end{gathered}$ | $\begin{gathered} 0.195 \\ (0.032) \end{gathered}$ | $\begin{gathered} -0.030^{*} \\ {[-1.79]} \end{gathered}$ |
|  | BMI obese ${ }^{\text {d }}$ | $\begin{gathered} 0.069 \\ (0.019) \end{gathered}$ | $\begin{gathered} 0.033 \\ (0.008) \end{gathered}$ | $\begin{aligned} & 0.036 * * \\ & {[2.83]} \end{aligned}$ |
|  | Ever had a miscarriage ${ }^{\text {f }}$ | $\begin{gathered} 0.116 \\ (0.016) \end{gathered}$ | $\begin{gathered} 0.110 \\ (0.014) \end{gathered}$ | $\begin{gathered} 0.006 \\ {[0.30]} \end{gathered}$ |
|  | Visited clinic in last 12 months ${ }^{\text {g }}$ | $\begin{gathered} 0.449 \\ (0.036) \end{gathered}$ | $\begin{gathered} 0.444 \\ (0.029) \end{gathered}$ | $\begin{array}{r} 0.005 \\ {[0.32]} \end{array}$ |
| Panel F | Currently married | $\begin{gathered} 0.852 \\ (0.020) \end{gathered}$ | $\begin{gathered} 0.836 \\ (0.023) \end{gathered}$ | $\begin{gathered} 0.015 \\ {[0.90]} \end{gathered}$ |
|  | Above primary education | $\begin{gathered} 0.312 \\ (0.056) \end{gathered}$ | $\begin{gathered} 0.226 \\ (0.049) \end{gathered}$ | $\begin{aligned} & 0.086^{* *} \\ & {[3.42]} \end{aligned}$ |

(continued)

Table 2 (continued)

| Women's Characteristics $\left(V_{i}\right)$ | Infertile $\left(q_{1}\right)$ | Fertile $\left(q_{2}\right)$ | Test $q_{1}-q_{2}=0$ |
| :--- | :---: | :---: | :---: |
| Urban | 0.435 | 0.325 | $0.109 * *$ |
| Age at first marriage | $(0.043)$ | $(0.034)$ | $[5.97]$ |
|  | 16.55 | 16.13 | 0.427 |
|  | $(0.300)$ | $(0.301)$ | $[1.61]$ |

Notes: * denotes significance at 10 percent, ${ }^{* *}$ at 5 percent, and ${ }^{* * *}$ at 1 percent. Standard errors in parenthesis and $t$-statistics in brackets.
a. Missing for Bolivia (1998), Colombia, Dominican Republic, Ghana, Guatemala, Kazakhstan, Comoros, Kyrgyz Republic, Nicaragua, Niger, Nepal, and Uzbekistan.
b. Missing for Nicaragua/excludes inconsistent and don't know.
c. Missing for Guinea, Madagascar, and Philippines.
d. Only available for Colombia, Dominican Republic, Kazakhstan, and Kyrgyz Republic.
e. Missing for Comoros and South Africa (only numeric answers included).
f. Missing for Benin, Bolivia (1998), Guinea, Comoros, Mozambique, and Zambia.
g. Missing for Ghana.

In essence, these regressions ask, conditional on age, whether infertile women are different from their fertile counterparts $\left(\theta_{1}-\theta_{2}=0\right)$ in terms of observable characteristics. We control for age as the average fertile woman in the sample is 30.2, while the average infertile woman is 35.9 and different cohorts of women have different background traits. ${ }^{7}$

Panel A of Table 2 confirms that infertile women have significantly fewer children than their fertile counterparts. On average, after conditioning on age, infertile women have 1.3 fewer children. Additionally, infertile women are 23 percentage points more likely to be childless than their fertile counterparts. Panel B presents a preliminary look at our main findings. While infertile women have fewer children than fertile women, there appears to be no difference in their likelihood of working in the last 12 months or in the odds of their working year round. Infertile women do appear to be more likely to work for a wage.

If there is a genetic component to fertility problems, then women for whom biology restricts their family size will have the additional advantage of having fewer siblings and will have had access to more parental resources via a quality quantity tradeoff. If this is the case, any increase in labor force participation associated with fewer children may be due to the fact that women who bear fewer children had higher "quality" upbringings.

The data do not allow us to directly measure the intergeneration correlation of fertility problems. However, for a subsample of the countries in our analysis we can investigate if infertile women, on average, come from smaller families than their fertile counterparts. These results can be seen in Panel C of Table 2. The data suggest

[^6]that infertile women have as many siblings as their fertile counterparts. Additional evidence from twin studies (Christensen et al. 1998) suggests that the genetic component of infertility is minimal.

Panel D of Table 2 demonstrates that for many important background variables, infertile women mirror their fertile counterparts. They became sexually active at the same age, they were born in similar times of the year, they have similar numbers of older siblings, and they are the same height. When asked about their gender preference over offspring there is no difference by fertility status. Infertility thus mimics an experiment in which nature assigns to each woman an upper bound for the number of children, independent of background.

We find weak evidence in Panel E that current health is related to fertility. Infertile women are more likely to report being obese, but they are no more likely to have paid a recent visit to a health clinic or to have miscarried. When we turn in Panel F to current attributes, which may be decided after the onset of infertility, we find little difference in marriage behavior. However, infertile women are more educated than their fertile counterparts and they are more likely to reside in urban areas. Given the difference in contemporaneous traits between fertile and infertile women, the main model will condition on marital status, education, and current location. However, our results are not sensitive to the inclusion of these additional controls, nor to the inclusion of health indicators.

## IV. Results

## A. Main Findings

Identification of the IV model requires a strong correlation between our measure of infertility and family size. Figure 1a shows a visual representation of the first-stage results. The dotted line displays the average number of children for fertile women by age. The solid line represents the analog for infertile women. For all ages, it is clear that infertility is highly correlated with the number of children a woman has. Given the strength of our $F$-statistics (see Tables 3, 4, and 6-9, below), there is no evidence of weak instruments in our first-stage results.

A key result of our paper can be seen in Figure 1b, which shows the proportion of women who worked (paid cash or not) in the 12 months prior to the surveys by their infertility status and age. While infertility has a strong association with family size, the raw data in Figure 1b show that there are not systematic differences in female labor force participation by infertility status. The Wald estimate (short-dash line) computes the ratio of the difference in labor force participation over the difference in the number of children by infertility status and age. The Wald estimates show that there is no association between changes in family size, brought by infertility, and female labor force participation. In Table 3 we present the regression counterparts to this graphical representation.

Column 1 in Table 3 presents the OLS estimate, which suggests that each additional child decreases labor force participation by 2.4 percentage points. ${ }^{8}$ Column 2

[^7]

Figure 1a
Average Number of Children by Infertility Status
contains the corresponding 2SLS estimate. It suggests that the effect of children on labor force participation, using the variation in the number of children that comes through the infertility channel, is much smaller than the OLS estimates would suggest. This is a key result of the paper. This finding, of little to no casual impact of children on female labor force participation, is similar to the evidence from twin studies for married women in the United States (Bronars and Grogger 1994; Jacobsen et al. 1999). The 2SLS point estimate ( -0.006 ) is close to zero and statistically insignificant, suggesting that the OLS parameter was biased. This is consistent with the case where unobserved variables, such as female autonomy, religiosity, or career ambition, are important factors driving both female employment and smaller family size.

To further support that infertility is a valid instrument, it is instructive to consider several potential threats to the validity of a causal interpretation. For instance, educated women are more likely to report infertility and educated women are more likely to work. If infertility is unrelated to other determinates of labor force participation, then their inclusion should not alter our findings. In Model 2 (Columns 3 and 4) we add controls for education, age and education interactions, marital status,

[^8]

Figure 1b
Labor Force Participation by Infertility Status and Wald Estimates
Notes: Author's calculation based on the DHS III of countries selected. The 95 percent bootstrapped confidence intervals were calculated based on 1,000 replications.

Table 3
Number of Children and Labor Force Participation of Women

|  | Model 1 |  |  | Model 2 |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Dependent variable: Women <br> worked in the last 12 months <br> $(=1)$ | OLS | 2SLS | OLS | 2SLS |  |
|  | $(1)$ | $(2)$ | $(3)$ | $(4)$ |  |
|  |  |  |  |  |  |
| Number of children | $-0.024^{* * *}$ | -0.006 | $-0.017 * * *$ | -0.005 |  |
|  | $[0.002]$ | $[0.008]$ | $[0.002]$ | $[0.007]$ |  |
| Observations | 90,965 | 90,965 | 90,965 | 90,965 |  |
| $F$-statistic (first stage) |  | 814.2 |  | 853.9 |  |

[^9]spouse's education, age at first intercourse, and current location to the controls for age and survey contained in Model 1. While the coefficient on the OLS decreases in magnitude, the IV estimates are unchanged by the inclusion of these additional control variables.

We are particularly concerned that infertility is proxying for poor health and that poor health could directly influence labor force participation, invalidating our identification strategy. As cited above, some medical literature suggests a relationship between poor health and infertility. However, Field and Ambrus (2008) review more than 60 biomedical studies and conclude that the onset of menarche is not associated with socioeconomic status indicators, including malnutrition. This is consistent with the review of Behrman and Deolalikar (1988) and Behrman, Deolalikar, and Wolfe (1988), in which they explore the effect of nutrition on fertility. They report that studies that have systematically reviewed the evidence show little support for the nutrition-fertility link. Also, the authors point out that the few attempts to estimate the birth production function are likely to suffer from problems of reverse causality and endogeneity (Behrman and Deolalikar 1988, p. 690).

A consistent set of health indicators is not available for the full sample. For a small subsample of countries, we have anthropometric measures of health (height and categorical body mass index indicators for underweight, overweight, and obese). The results when reestimating Model 2 for the subsample that contains anthropometric measures are shown in Panel A of Table 4. In Panel B, for a different but larger set of countries, we have information on whether women visited a health clinic in the last year. Panel C contains results for an alternative sample for which we have information on miscarriages.
The first two columns of Table 4 show the results estimating Model 2 for the restricted subsample of surveys that contain the specific health-related variables in each panel. In all cases, there is a clear negative correlation between the number of children and female labor force participation, but it disappears when using 2SLS. This is analogous to the result in Table 3. For example, in Panel A, for the subsample that contains height and BMI indicators, the OLS estimate suggests that each child reduces labor supply by five percentage points, while the 2SLS estimate is a statistically insignificant 0.006 . Columns 3 and 4 add the corresponding health-related variables to the subsamples, and the parameter estimates are very close to estimates in Columns 1 and 2. The 2SLS parameters in Column 4 are small, insignificant, and sometimes positive. The results of this table suggest that omitted health factors are not contaminating our estimates. ${ }^{9}$ To preserve sample size, for the remainder of the analysis we will focus on Model 2 (which excludes health controls), although all findings are robust to the other model specifications described here.

[^10]Table 4
Number of Children and Labor Force Participation with Health Controls

|  | Model 2 |  | With health variables |  |
| :--- | :---: | :---: | :---: | :---: |
| Dependent variable: Woman | OLS | 2SLS | OLS | 2SLS |
| worked in the last 12 months | $(1)$ | (2) | (3) | (4) |


| Panel A: Includes controls for women's height | and body mass index indicators ${ }^{\mathrm{a}}$ |  |  |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Number of children | $-0.050^{* * *}$ | 0.006 | $-0.049^{* * *}$ | 0.007 |  |  |  |  |
|  | $[0.005]$ | $[0.015]$ | $[0.005]$ | $[0.015]$ |  |  |  |  |
| Observations | 9,867 | 9,867 | 9,867 | 9,867 |  |  |  |  |
| $F$-statistic (first stage) |  |  |  |  |  | 121.0 |  | 123.3 |
| Panel B: Includes controls for visited a health | clinic ${ }^{\mathrm{b}}$ |  |  |  |  |  |  |  |
| Number of children | $-0.025^{* * *}$ | -0.004 | $-0.025^{* * *}$ | -0.005 |  |  |  |  |
|  | $[0.003]$ | $[0.008]$ | $[0.003]$ | $[0.008]$ |  |  |  |  |
| Observations | 87,842 | 87,842 | 87,842 | 87,842 |  |  |  |  |
| $F$-statistic (first stage) |  | 756.0 |  | 751.6 |  |  |  |  |
| Panel C: Includes controls for miscarriages ${ }^{\mathrm{c}}$ |  |  |  |  |  |  |  |  |
| Number of children | $-0.029^{* * *}$ | -0.001 | $-0.029 * * *$ | -0.001 |  |  |  |  |
|  | $[0.003]$ | $[0.010]$ | $[0.003]$ | $[0.010]$ |  |  |  |  |
| Observations | 59,537 | 59,537 | 59,537 | 59,537 |  |  |  |  |
| $F$-statistic (first stage) |  | 509.7 |  | 508.5 |  |  |  |  |

Notes: Robust standard errors (in brackets) are clustered at the subnational level. * denotes significance at 10 percent, ${ }^{* *}$ at 5 percent, and ${ }^{* * *}$ at 1 percent. Model 2: educational attainment, age, age and education interactions, age at first intercourse, marital status, age at first marriage, and spouse's education and survey fixed effects. The additional controls added for Columns 3 and 4 are described in each panel. We include underweight, overweight, and obese as BMI indicators. The 2SLS instrument for children at home using the union of the infertility measures. The $F$-statistic refers to the first-stage results.
a. Anthropometric information only available for Colombia, Dominican Republic, Kazakhstan, and Kyrgyz Republic.
b. Missing for Nigeria.
c. Missing for Benin, Bolivia (1998), Guinea, Comoros, Mozambique, and Zambia.

## B. Accuracy of Self-Reported Measures of Infertility

A potential concern with the data used in this paper is measurement error due to the self-reported nature of the infertility measure. ${ }^{10}$ Women may be ignorant of their true underlying fertility, they may fear revealing fertility problems to a survey taker, they may have idiosyncratic definitions of infertility, or they may be medically infertile but offer other reasons for why they are not using contraception or do not desire additional children. The two survey questions used to classify women as infertile are asked to the full sample of women, yet about two-thirds of the women

[^11]who report infertility in one question provide a response other than infertility to the other survey question. For instance, some women who state that they cannot have additional children, when asked about their desire for future children offer early menopause, ignorance about birth control, or infrequent sexual activity as the main reason for not using contraceptives.

We acknowledge the possible presence of measurement error in the reports of infertility; however, we do not expect measurement error to present a significant problem for two reasons. First, classical measurement error in an instrumental variable does not bias the estimated impact of children on labor force participation. ${ }^{11}$ Second, self-reported measures of infertility are strongly correlated with biologically measured infertility. In a widely cited paper, Cates, Farley, and Rowe (1985) conducted a worldwide study to provide a standard approach to compare infertility patterns in couples in 33 medical centers from 25 countries. From 1979 to 1984 the study compiled information about couples seeking medical evaluation for their selfreported infertility problem. They show that the percentage of couples who selfreported infertility and became pregnant at some point during the investigation ranged from 16 percent in Asia and Africa, to 12 percent for developed areas in Europe and Australia. This suggests a relatively low level of measurement error in the self-reported infertility rates of couples. ${ }^{12}$

An additional concern is that women who answered the questionnaire in the presence of family members (especially male family members) might misrepresent their fertility status. For example, women with low bargaining power within the marriage could choose to report themselves as infertile when the husband is present as a way to hide the real reason for their lack of contraceptive use. These reasons could include the husband or other adults prohibiting the woman's use of family planning methods. To investigate this concern, we employ the strategy described in Equation 2 with the presences of male family members as the variable of interest. In Table 5 we show that there is no systematic relationship between the presence of a husband, the presence of other males, or the presence of other family members at the time of the survey and the likelihood self-reported infertility. This evidence allows us to reject another type of possible misclassification of our infertility measures.

Finally, Table 6 explores the robustness of our main finding to various specifications of the instrument. The first column replicates the 2SLS results in Table 3,

[^12]Table 5
Presence of other Adults and Fertility Status

| Additional characteristics $\left(V_{i}\right)$ | Infertile $\left(q_{1}\right)$ | Fertile $\left(q_{2}\right)$ | Test $q_{1}-q_{2}=0$ |
| :--- | :---: | :---: | :---: |
| Husband was present during interview $^{\mathrm{a}}$ | 0.017 | 0.018 | -0.001 |
|  | $(0.005)$ | $(0.004)$ | $[0.003]$ |
| Other males were present during interview | 0.037 | 0.041 | -0.005 |
|  | $(0.011)$ | $(0.010)$ | $[0.004]$ |
| Adults were present during interview | 0.120 | 0.135 | $-0.015^{*}$ |
|  | $(0.025)$ | $(0.023)$ | $[0.008]$ |

Notes: * denotes significance at 10 percent, ** at 5 percent, and $*^{* *}$ at 1 percent. Standard errors in parenthesis and $t$-statistics in brackets.
a. For married women only.
where a woman is classified as infertile if she responds in the affirmative to either of the infertility questions ( $\max =\{$ Infert1,Infert 2$\}$ ). Columns 2 and 3 present 2SLS estimates where each of the survey questions that solicit information about infertility is used as the sole instrument. The 2SLS estimates are nearly identical across these alternative definitions of infertility. The first-stage results reveal that an affirmative answer to either of the infertility questions is associated with significantly smaller family sizes.

We next attempt to improve upon our classification of infertility. In this specification the instrumental variable is assigned the values of zero, one, or two in correspondence to the number of survey questions in which a woman revealed herself to be infertile (Infert3 $=$ Infert + Infert 2 ). The underlying assumption is that a woman who in two separate instances stated that she is infertile is more likely to be truly infertile. After conditioning on age, the average woman who identified as infertile on both survey questions has 2.2 fewer children than her fertile counterpart. Column 4 contains the 2SLS results with the Infert3 instrument. The estimated causal impact of children on mothers' labor force participation is almost identical to the results shown in the base specification Column 1. Under the assumption that this coding of the instrument reduces the amount of measurement error in the definition of infertility, this analysis suggests that misclassification in the infertility variable does not bias the 2SLS estimates.

In Column 5, we exploit having two separate survey questions that identify infertility and use each as an instrument to run a Hansen $J$-test for overindentifying restrictions. Column 6 adds Infert4 $=$ Infert1 Infert2 to the set of instruments for the Hansen $J$-test. Both GMM models again confirm our main finding of no causal impact of children on female labor force participation. The estimated $J$ statistics cannot reject the joint null hypothesis of correct model specification and orthogonally of the instruments.
Table 6
Robustness of the Results to Alternative Definitions of the Instrument

| Method: Instrument | $\begin{gathered} \text { 2SLS: max } \\ \{\text { Infert1, Infert } 2\} \end{gathered}$ <br> (1) | 2SLS: Infertl <br> (2) | 2SLS: Infert2 <br> (3) | 2SLS: Infert $3^{\text {a }}$ <br> (4) | GMM: Infert1, Infert 2 <br> (5) | GMM: Infert1, Infert2, Infert $4^{\text {b }}$ <br> (6) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Number of children | -0.005 | -0.004 | -0.005 | $-0.004$ | -0.004 | -0.004 |
|  | [0.007] | [0.008] | [0.010] | [0.007] | [0.007] | [0.007] |
| Observations | 90,965 | 90,965 | 90,965 | 90,965 | 90,965 | 90,965 |
| First stage | - 1.212 | -1.354 | - 1.162 | - 1.046 |  |  |
| $F$-statistic | 853.92 | 920.39 | 420.92 | 965.98 | 558.73 | 418.48 |
| Hansen $J$ ( $p$-value) |  |  |  |  | 0.893 | 0.950 |

[^13]Table 7
Alternative Definitions of Children

|  | OLS <br> $(1)$ | 2SLS <br> $(2)$ | First Stage <br> $(3)$ | Observations |
| :--- | :---: | :---: | :---: | :---: |
| (1) Number of children at | $-0.017 * * *$ | -0.005 | -1.212 | 90,965 |
| home | $[0.002]$ | $[0.007]$ | $(853.9)$ |  |
| (2) Number of children | $-0.039 * * *$ | -0.011 | -0.549 | 90,965 |
| younger than six | $[0.004]$ | $[0.016]$ | $(469.3)$ |  |
| (3) Presence of children | $-0.047 * * *$ | -0.027 | -0.223 | 90,965 |
|  | $[0.007]$ | $[0.041]$ | $(717.4)$ |  |

Notes: Presence of children is a binary variable equal to one if family size is at least one and zero otherwise. Each cell in Columns $1-3$ represents separate regressions. * denotes significance at 10 percent, ** at 5 percent, and *** $^{*}$ at 1 percent. All regressions include the control variables listed in Model 2. Robust standard error clustered at subnational levels in squared bracket and $F$-statistics for the first stage are in parenthesis.

## C. Alternative Definitions of Family Size

In Table 7 we consider alternative definitions of family size. The first row reproduces the main results from Table 3. In this case, family size is defined as the total number of children at home irrespective of their age. The second row uses the number of children younger than the age of six as the measure of family size. As the care of young children is more intensive, the presence of young children may have a larger impact on mothers' labor force participation. Here, the OLS is larger than in the first case. An additional child younger than six is associated with a decline in labor force participation of four percentage points, as opposed to 1.7 points when considering all children. Similarly, in absolute value, the 2SLS parameter is larger for children younger than six $(-0.011)$ than for all children $(-0.005)$. In both cases, these estimates are not statistically different from zero and are substantially smaller than their OLS counterparts.

As mentioned before, the existing literature is restricted to a sample of women who have had at least one child. Since infertility, unlike twinning or the sex composition of the offspring, can impact all women, we are able to investigate the difference between childless women and women with children as opposed to imposing a linear relationship between number of children and female labor force participation. Our conclusion, of little to no causal effect of family size on female labor force participation, may be an artifact of the specification of a linear model, which masks sizable marginal family effects. It is possible that the disruption to female labor force participation is concentrated in the first child and that the impact of additional children is negligible. Mogstad and Wiswall (2009) argue that relaxing linearity restrictions in IV estimation of models is an important addition to empirical research, particularly when theory suggests the possibility of nonlinear causal effects.

Table 8
Heterogeneous Effects

|  | OLS <br> (1) | $\begin{gathered} \text { 2SLS } \\ \text { (2) } \end{gathered}$ | First Stage (3) | Observations |
| :---: | :---: | :---: | :---: | :---: |
| Panel A: Full sample <br> (1) Number of children | $\begin{gathered} -0.017 * * * \\ {[0.002]} \end{gathered}$ | $\begin{gathered} -0.005 \\ {[0.007]} \end{gathered}$ | $\begin{aligned} & -1.212 \\ & (853.9) \end{aligned}$ | 90,965 |
| Panel B: Younger than 35 <br> (2) Number of Children | $\begin{gathered} -0.021 * * * \\ {[0.003]} \end{gathered}$ | $\begin{gathered} -0.010 \\ {[0.012]} \end{gathered}$ | $\begin{aligned} & -1.122 \\ & (435.7) \end{aligned}$ | 65,890 |
| Panel C: Middle-income c <br> (3) Number of Children | $\begin{aligned} & \text { thries }{ }^{\mathrm{a}} \\ & -0.032 * * * \\ & {[0.003]} \end{aligned}$ | $\begin{gathered} -0.003 \\ {[0.014]} \end{gathered}$ | $\begin{aligned} & -1.088 \\ & (387.8) \end{aligned}$ | 32,474 |
| Panel D: Low-income coun <br> (4) Number of children | $\begin{aligned} & \mathrm{ies}^{\mathrm{a}} \\ & \quad-0.009 * * * \\ & {[0.002]} \end{aligned}$ | $\begin{gathered} -0.010 \\ {[0.008]} \end{gathered}$ | $\begin{aligned} & -1.292 \\ & (524.5) \end{aligned}$ | 58,491 |

Notes: Each cell in Columns 1-3 represents separate regressions. * denotes significance at 10 percent; ** at 5 percent, and ${ }^{* * *}$ at 1 percent. All regressions include the control variables listed in Model 2 . Robust standard error clustered at subnational levels in squared bracket and F-statistics for the first stage are in parenthesis.
a. Countries are classified according to the World Bank's World Development Indicators.

Row 3 of Table 7 shows the results where the main independent variable is an indicator variable that takes a one if a woman has any children, and is a zero if she is childless. Being a mother as opposed to being childless is associated with a 4.7 percentage point reduction in work behavior (Column 1, Row 3). Consistent with our previous results, the 2SLS estimates suggest that the observed association overestimated the true relation in a way that is consistent with the existence of unobserved variables, such as autonomy or career ambition, affecting both the labor supply and the decision to become a mother. The 2SLS estimate suggests a lower effect (in absolute value) and it is not statistically different from zero. ${ }^{13}$ Regardless of how we define family size, our results indicate that children do not have a causal effect on the likelihood that a woman enters the labor force.

## D. Heterogeneous Impacts

In Table 8, we explore the robustness of our findings to particular subsamples of the data. Panel A reproduces the full-sample baseline results. We first investigate whether an effect could be observed for younger women for whom the age of

[^14]children will be significantly lower. In Panel B we restrict the sample to those aged 35 or younger, as in Angrist and Evans (1998). We find a negative correlation between family size and labor participation for younger women. In absolute terms, the effect is slightly larger for this subsample ( 2.1 percentage points) compared to the full sample (1.7). As before, the 2SLS parameter (Column 2) suggests an overestimation of the OLS ( -1 percentage point). However, consistent with our previous findings, we cannot reject the null hypothesis of a zero effect for this subsample either.

Panels C and D of Table 8 present results from our preferred specification for the subsample of women at different levels of economic development as defined in Appendix Table A1. The switch in economic development from primarily agricultural societies to more industrialized societies generates a U-shaped relationship between female labor force participation and per capita income (Goldin 1995 and Mammen and Paxson 2000). As countries enter intermediate levels of development, working women must leave their homes and enter a labor market with few promother labor policies, where they may face social stigma. As such, we may expect the relationship between family size and mothers' labor force participation to be different for women in nations at differing stages of development. In particular, if in agricultural societies there exist few opportunities to work outside of the home, while in more developed countries formal work reflects more of a choice, then we might expect the OLS estimates to be more upward biased in the more developed countries.

Panel C of Table 8 presents results for the subsample of women in middle-income countries. The OLS estimates suggest that each additional child reduced labor force participation by 3.2 percentage points (Column 1), while the 2SLS estimates are very small and not statistically different from zero. This suggests that the observed correlation between family size and labor force participation for women in midincome countries is spurious and reflects the fact that certain types of women select into work.

Panel D of Table 8 shows the results for women in the lower-income countries. The OLS estimates are smaller in magnitude compared to the middle-income country estimates. However, for this subsample we cannot reject the hypothesis that the OLS and IV estimates are the same. For women in the least developed countries, work is mainly agricultural and there is little selection into the labor force. As we will show later, for this subsample of women, children appear to have a small but significant impact on paid labor force participation. We also have explored dividing the sample by the mother's level of education (completed at least primary education). The results mirror our findings for the middle- and low-income countries (not shown for brevity). The OLS results suggest that children have a greater impact on the labor force behavior of more educated women. The 2SLS results show that children have no causal impact on the labor force behavior of more educated mothers. For less-educated women, additional children appear to have a small impact on work behavior.

## E. Alternative Definitions of Work

Finally, Table 9 investigates the robustness of our findings to different definitions of labor force participation. In Panel A we show the results that classify a woman as

Table 9
Alternative Definitions of Labor Force Participation

|  | OLS | 2SLS |  |  |
| :---: | :---: | :---: | :---: | :---: |
| $(1)$ | First Stage <br> $(3)$ | Observations |  |  |
| Panel A: Work Intensity |  |  |  |  |
| (1) Full sample | $-0.014^{* * *}$ | 0.001 | -1.211 | 90,943 |
|  | $[0.002]$ | $[0.007]$ | $(852.8)$ |  |
| (2) Women younger | $-0.017 * * *$ | 0.000 | -1.122 | 65,869 |
| than 35 | $[0.003]$ | $[0.011]$ | $(435.6)$ |  |
| (3) Mid-income | $-0.029 * * *$ | 0.005 | -1.088 | 32,452 |
| countries | $[0.003]$ | $[0.012]$ | $(387.0)$ |  |
| (4) Low-income | $-0.008^{* * *}$ | -0.003 | -1.292 | 58,491 |
| countries | $[0.002]$ | $[0.009]$ | $(524.5)$ |  |
| Panel B: Paid Cash |  |  |  |  |
| (5) Full Sample | $-0.020^{* * *}$ | -0.009 | -1.216 | 87,210 |
|  | $[0.002]$ | $[0.008]$ | $(839.1)$ |  |
| (6) Women younger | $-0.025 * * *$ | $-0.018 *$ | -1.141 | 63,145 |
| than 35 | $[0.003]$ | $[0.011]$ | $(466.8)$ |  |
| (7) Mid-income | $-0.034^{* * *}$ | 0.000 | -1.088 | 32,456 |
| countries | $[0.002]$ | $[0.015]$ | $(388.1)$ |  |
| (8) Low-income | $-0.012^{* * *}$ | $-0.021 * * *$ | -1.306 | 54,754 |
| countries | $[0.002]$ | $[0.008]$ | $(516.9)$ |  |

Notes: Work Intensity is a binary variable equal to one if the women worked year around and zero otherwise. Paid Cash is a binary variable that equals one if the women worked for cash and zero otherwise. Each cell in Columns $1-3$ represents a separate regression. * denotes significance at 10 percent; ** at 5 percent, and $* * *$ at 1 percent. Observations with missing outcome variables were dropped. All regressions include the control variables listed in Model 2. Robust standard error clustered at subnational levels in squared bracket and F -statistics for the first stage are in parenthesis.
working only if she works (either paid or not) for the full year. As before, children appear to have no casual impact on the likelihood that a woman works year round as the IV estimate is sometimes positive and very close to zero. This pattern is also observed for younger women and in both medium- and low-income countries.

In Panel B we reproduce the same regressions as in Panel A, but applied to the definition of work as limited to any paid work, as opposed to unpaid agricultural work or work that is paid in kind, in the last 12 months. Limiting our focus to work for which cash changes hands generates interesting new results. First, for the full sample we find that an additional child does not seem to affect the participation of women (Row 5, Column 2). This result is consistent with the findings of Agüero
and Marks (2008) in a paper limited to Latin American countries. However, when restricting the sample for younger women (Row 6) we observe a statistically significant effect in the 2SLS. An extra child diminishes paid employment by almost two percentage points. Second, for middle-income countries there is no causal effect. This is consistent with our previous discussion and with results for Latin American countries from Agüero and Marks (2008). However, an additional child reduces the paid employment of women in low-income countries by 2.1 percentage points. In these countries, our estimates suggest that a woman with 2.4 children (the sample average) has a 5 percent lower probability of paid employment compared to her childless counterpart, all other things being equal. In less-developed countries there appear to be real challenges balancing motherhood with participation in the formal labor market.

## V. Conclusions

Our paper investigates the relationship between children and labor force participation for a sample of women representing 26 countries in the developing world. We employ a new identification strategy (infertility) in which "nature" prevents some women from obtaining their desired fertility levels. We present evidence that supports the use of infertility as an instrument for child bearing. First, infertility is highly correlated with family size. In our sample, infertile women have 1.2 fewer children then their fertile counterparts. Second, we show that our classification of infertility is unrelated to a woman's observed background traits. Third, we argue that measurement error in self-reported infertility does not bias our identification strategy.

We find, for the subpopulation of women who are not actively controlling their fertility, that children have a much smaller causal effect on mothers' overall labor force participation than the OLS estimates predict. These findings are robust to the inclusion of additional controls and alternative classifications of number of children. Additionally, these findings are robust to different definitions of labor force participation (worked last year, worked for pay, and worked year round). However, our results suggest that having children is indeed a barrier for participation in the paid labor force for younger women and mothers in poorer countries. Thus, motherhood affects the type of work a woman pursues by decreasing the likelihood of paid work.

Our results for middle-income countries contrast Cruces and Galiani (2007) and Caceres-Delpiano (2008), who find that women in the developing world who are induced to have an additional child, out of a desire for a balanced sex mix of their children or because of twinning, are less likely to participate in the labor force. Our results could differ because we are identifying effects for a different subpopulation. ${ }^{14}$ A recent paper by Ebenstein (2009) argues that, in the presence of heterogeneous treatment effects, the LATE may differ from the average treatment effect when those

[^15]influenced by the instrument (the compliers) are not representative of the overall population. The women whose family size is altered by the sex composition of their children or a twin birth are women with low underlying desired fertility, whereas infertility removes children from women with high underlying desired fertility. Given that family planning programs tend to target women with high desired fertility, our estimates are arguably more relevant from a public policy vantage point.

While we find that, for younger women and those living in poorer countries, participation into paid labor is slightly reduced by having children, our main findings provide little evidence to support the belief that the global rise in female labor force participation can be directly attributed to declining family sizes. A common factor such as the empowerment of women could be driving both trends. Thus, our results suggest that policies focusing solely on family planning are unlikely to greatly increase female labor force participation.

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Appendix
Statistics by Survey
Table A1 Summary Statistics by Sur

| Survey year | Infertile <br> (1 or 2) | Number of <br> children | Worked last <br> 12 months | Above primary <br> school | Observations |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Country |  |  |  |  |  |  |
| Mid-income countries |  |  |  |  |  |  |
| Bolivia | 1994 | 0.08 | 2.71 | 0.68 | 0.35 | 3,039 |
| Bolivia | 1998 | 0.06 | 2.74 | 0.58 | 0.36 | 3,696 |
| Brazil | 1996 | 0.15 | 1.67 | 0.62 | 0.49 | 2,065 |
| Columbia | 1995 | 0.12 | 1.55 | 0.67 | 0.52 | 2,377 |
| Dominican Republic | 1996 | 0.12 | 1.60 | 0.50 | 0.36 | 2,029 |
| Guatemala | 1999 | 0.05 | 3.39 | 0.27 | 0.08 | 2,555 |
| Kazakhstan | 1995 | 0.09 | 1.68 | 0.78 | 0.99 | 1,036 |
| Peru | 1996 | 0.07 | 2.40 | 0.67 | 0.46 | 7,255 |
| Philippines | 1998 | 0.04 | 2.60 | 0.54 | 0.59 | 4,125 |
| Uzbekistan | 1996 | 0.08 | 1.97 | 0.62 | 1.00 | 1,184 |
| South Africa | 1998 | 0.06 | 1.64 | 0.38 | 0.53 | 3,113 |

Table A1 (continued)

|  | Survey year | Infertile <br> (1 or 2) | Number of <br> children | Worked last <br> 12 months | Above primary <br> school | Observations |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Country |  |  |  |  |  |  |
| Low-income countries | 1996 | 0.06 | 2.46 | 0.93 | 0.05 | 3,190 |
| Benin | 1995 | 0.06 | 2.06 | 0.86 | 0.10 | 3,444 |
| Central Africa Republic | 1997 | 0.05 | 2.77 | 0.62 | 0.04 | 4,742 |
| Chad | 1996 | 0.03 | 2.92 | 0.46 | 0.13 | 1,231 |
| Comoros | 1999 | 0.06 | 2.01 | 0.86 | 0.40 | 2,720 |
| Ghana | 1999 | 0.04 | 2.45 | 0.85 | 0.06 | 4,320 |
| Guinea | 1998 | 0.05 | 2.58 | 0.59 | 0.23 | 3,301 |
| Kenya | 1997 | 0.15 | 1.93 | 0.55 | 1.00 | 1,071 |
| Kyrgyz Republic | 1997 | 0.04 | 2.39 | 0.83 | 0.22 | 3,737 |
| Madagascar | 1996 | 0.02 | 2.69 | 0.56 | 0.04 | 6,338 |
| Mali | 1997 | 0.05 | 2.17 | 0.66 | 0.03 | 5,427 |
| Mozambique | 1998 | 0.05 | 2.67 | 0.43 | 0.30 | 3,673 |
| Nicaragua | 1998 | 0.03 | 2.54 | 0.61 | 0.05 | 4,648 |
| Niger | 1999 | 0.04 | 2.53 | 0.53 | 0.25 | 4,384 |
| Nigeria | 1997 | 0.10 | 2.28 | 0.56 | 0.22 | 4,135 |
| Zambia | 1994 | 0.09 | 2.11 | 0.53 | 0.30 | 2,130 |
| Zimbabwe |  |  |  |  |  |  |

Note: Countries are classified according to the World Bank's World Development Indicators.


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[^1]:    1. For instance, Bloom et al. (2009) in a sample of 97 countries finds that an increase in the total fertility rate (TFR) of one corresponds to around a three percent reduction in female labor force participation. Average TFR in this sample is 4.34 , so this suggests that the average woman has a reduction of labor force participation of 13 percent due to children. See also Kögel (2004) for an analysis of Organization of Economic Cooperation and Development countries and Schultz (2008) for a critical review of the literature.
[^2]:    2. Recent quasi-experimental research in the United States suggests that declining fertility rates can explain little to none of the changes in women's labor force participation over time. See Bronars and Grogger (1994) and Jacobsen et al. (1999) (twinning); Hotz, McElroy, and Sanders (2005) (miscarriage); Angrist and Evans (1998) (the sex composition of the first two children). Our identification strategy most closely resembles Cristia (2008), who investigates differences in employment outcomes by child status for women seeking help to become pregnant. He finds a large negative short-run impact on female labor force participation owing to the birth of a first child. Our approach is also similar to Li (2005), who exploits miscarriages and finds a negative and significant impact on labor supply.
    3. In the context of developing countries, the low birth weight of twins (compared to singletons) could add pressure to infant mortality. This could have a direct effect on labor participation decisions violating the exclusion restriction.
[^3]:    4. 69 percent of women not using contraceptives at the time of the survey had never used them.
    5. A meta-analysis of papers using population surveys suggests that the 12 -month infertility rate in lessdeveloped countries is between 6.9 and 9.3 percent (Boivin et al. 2007).
[^4]:    a. Full sample adds back in women who were currently taking contraceptives.
    b. Countries are classified according to the World Bank's World Development Indicators.
    c. Missing for Comoros and South Africa (only numeric answers included).
    d. Missing for Guinea, Madagascar, and Philippines.
    e. Missing for Nicaragua/excludes inconsistent and don't know.
    f. Missing for Ghana.

[^5]:    6. For example, if women who drink take more risks than women who do not drink, then drinkers with higher (natural) fertility may have had all of their desired pregnancies through unintended conception, leaving only the relatively subfertile drinkers to be at risk of self-identified difficulties in conception and to be included in the sample (see Weinberg and Dunson 2000, Tielemans et al. 2002).
[^6]:    7. Regressions (not shown) that exclude the age controls produce qualitatively similar results, except that when one does not condition on age it appears as if infertile women work more than fertile women (which is expected since working increases with age, see Figure 1b). Additionally, infertile women are more likely to report a miscarriage.
[^7]:    8. It is difficult to compare our results with the existing literature because of the different universe and
[^8]:    the definition of the variable of interest. For example, Cruces and Galiani (2007) find that increasing the family size from two to more than two negatively affects the labor force participation by nine percentage points (using data from Mexico and Argentina). The corresponding estimate in the United States in 1990 is 15 percentage points (Angrist and Evans 1998). Cáceres-Delpiano (2008) finds that an additional child decreases labor supply by 2.8 percentage points for a sample of women in developing countries with at least one child.

[^9]:    Note: Robust standard errors (in brackets) are clustered at the subnational level. * denotes significance at 10 percent, $*^{* *}$ at 5 percent, and $* * *^{*}$ at 1 percent. Model 1 includes women's age and survey fixed effects. Model 2 adds to Model 1 education, age and education interactions, age at first intercourse, marital status, age at first marriage, and spouse's education. The 2SLS instrument for children at home using the union of the infertility measures. The $F$-statistic refers to the first-stage results.

[^10]:    9. A possible concern could be that the onset of infertility negatively impacts mental health and poor mental health could reduce labor force participation. The DHS does not contain mental health indicators. However, their omission from the estimating equation would lead to upward biased coefficients for family size if the above relationship holds. Because we find no effect of family size on labor force participation, we can rule out a positive effect. Thus, our main conclusion still holds.
[^11]:    10. It is possible that our measure of infertility is idiosyncratic due to self-reporting. However, other papers using biological variables as instruments also used self-reported information, which could be idiosyncratic as well. For example, the age of menarche in Field and Ambrus (2008) and the incidence of miscarriage in Li (2005) are self-reported measures.
[^12]:    11. Mathematically we have a system of the following two equations $K_{i}=X^{\prime} \pi_{10}+\pi_{11} Z_{i}+\xi_{1 i}$ and $L F P_{i}=X^{\prime} \pi_{20}+\pi_{21} Z_{i}+\xi_{2 i}$ where the parameter $\pi_{11}$ captures the first-stage effects of $Z_{i}$ on $K_{i}$, adjusting for covariates. The parameter $\pi_{21}$ captures the reduced form effects of $Z_{i}$ on $L F P_{i}$ adjusting for the same covariates. The covariate-adjusted IV estimator is the sample analog of the ratio $\pi_{21} / \pi_{11}=\rho$ (Angrist and Pischke 2008). Suppose that $Z_{i}$ is measured with classical measurement error such that $Z_{j}=Z_{j}^{*}+e_{j}$ where $Z_{j}^{*}$ is true infertility. Under classical measurement error the following is true $\operatorname{plim}\left(\hat{\pi}_{21}\right)=\lambda \pi_{21}$ and $\operatorname{plim}\left(\hat{\pi}_{22}\right)=\lambda \pi_{22}$ where $\lambda$ is the reliability ratio which is equal to $\operatorname{Cov}\left(Z_{i}, Z_{i}^{*}\right) / \operatorname{Var}\left(Z_{i}\right)$ Wooldridge (2001). The $\lambda$ s cancel out and the covariate-adjusted IV estimator remains $\rho$, and thus is unbiased.
    12. Recent papers on the prevalence of infertility in developing countries rely on self-reported data (see Boivin et al. 2007 and de Kok 2007). Walraven et al. (2001) is an exception. In rural Gambia, women were asked about infertility problems by a fieldworker and by a female gynecologist. The authors found that respondents reported a higher prevalence of infertility to gynecologists ( 14 percent instead of 10 percent). However, the authors do not report the proportion of cases where self-reported infertility (in either case) compares to test results from medical examination.
[^13]:    Notes: * denotes significance at 10 percent, ${ }^{* *}$ at 5 percent, and ${ }^{* * *}$ at 1 percent. All regressions include the control variables listed in Model 2 . Robust standard error clustered at subnational levels in squared bracket. The $F$-statistic refers to the first-stage result. The $G M M$ specifications use each infertility measure separately. The Hansen- $J$ shows the $p$-value for the test for overidentifying restrictions.
    a. Infert 3 is equal to Infert1 + Infert2.
    b. Infert4 is the product of Infertl and Infert 2 .

[^14]:    13. We observe a similar pattern when considering the intensity of work as measure by working full-year versus part-year. However, it appears as if motherhood reduces the likelihood of paid work by five percentage points from a base of 0.48 .
[^15]:    14. The difference in findings is not due to the fact that we are able to include childless women in our sample while the other empirical strategies are restricted to women with children. If we restrict our analysis to the subsamples of at least one child (twinning) or at least two children (sex mix) the overall results remain unaltered.
