6.1 Introduction

Each year, roughly five percent of teenagers give birth in the United States, a level that is considerably higher than that in any other developed country (United Nations 2006). This point-in-time statistic masks higher cumulative rates of childbearing as women pass through their teen years. As we show subsequently, in the United States between 7 and 10 percent of women will give birth before the age of eighteen and roughly 20 percent will give birth before the age of twenty.

Concern is often expressed regarding the potential harm that teen childbearing imposes on the mother, the child, and potentially to society more broadly. The National Campaign to Prevent Teen Pregnancy (2007) has summarized many of the statistics that are often used to support arguments about the potential pitfalls associated with teen childbearing. They highlight the fact that women who give birth as teens tend to subsequently have lower educational attainment and higher rates of welfare receipt. Their children are more likely to be born with low birth weight and have weaker performance in school. Although it is difficult to determine the extent to which the teen birth is the causal reason for these poor outcomes, these relationships

Melissa S. Kearney is an assistant professor of economics at the University of Maryland and a faculty research fellow of the National Bureau of Economic Research. Phillip B. Levine is the Class of 1919 Professor at Wellesley College and a research associate of the National Bureau of Economic Research.

This paper was prepared for the conference meeting, “An Economics Perspective on the Problems of Disadvantaged Youth.” The authors thank conference participants for very helpful comments and Dubravka Colic and Rebecca Vichniac for very helpful research assistance. This research was funded by the Annie E. Casey Foundation. We thank them for their support but acknowledge that the findings and conclusions presented in this report are those of the authors alone, and do not necessarily reflect the opinions of the Foundation.
are both sufficiently strong and alarming that they receive a great deal of attention.

If early childbearing is associated with poor outcomes for both mothers and their children, then why do women give birth at such an early age? Public discussions directed at answering this question have focused on a number of potential explanations: the incentives of the welfare system, poor labor market outcomes for teens, lack of access to affordable contraception, poor parental and peer influences, and socioeconomic disadvantage, among others. In this chapter we focus on the last potential contributor.

Socioeconomic disadvantage can lead to early childbearing through a number of different mechanisms. The poor may lack the resources available to know about the different opportunities available to them or to take advantage of those opportunities. This could hinder their ability to make optimal choices regarding contraceptive use, educational attainment, labor market training, and the like. Alternatively, those at the bottom of the economic ladder may have given up hope of improving their economic conditions or those of their offspring. Schools and/or labor market conditions in their communities may be so weak that staying in school and avoiding early motherhood might not be seen as offering any material benefit. In addition, some evidence suggests that those who grow up in disadvantaged situations have a stronger “taste” for children. Edin and Kafalas (2005) argue that “the daily stresses of an impoverished adolescence . . . breed a deep sense of need for something positive to ‘look to’” (205).

Our goal in this chapter is to examine the empirical relationship between socioeconomic disadvantage and rates of early childbearing. We begin by exploring past research in different disciplines that posit factors that may lead to early childbearing, focusing on the role that socioeconomic disadvantage may play. We then extend our literature review to discuss related empirical research that may inform our discussion. We continue our analysis by offering our own empirical exercises. First, we use micro-level data from the Panel Study of Income Dynamics (PSID) to provide a descriptive analysis of the relationship between socioeconomic disadvantage and early childbearing. Second, we aggregate Vital Statistics microdata from 1968 through 2003 to conduct a cohort-based analysis of the relationship between rates of socioeconomic disadvantage of a birth cohort and the cohort’s subsequent early childbearing experiences. We proxy for disadvantage at birth with four alternate factors, all based on the mother’s characteristics: having been born

1. To be clear, we do not investigate what it is about socioeconomic disadvantage that leads young women to have children before the age of eighteen or twenty. Our empirical approach does not allow us to separately identify which aspects of socioeconomic disadvantage—such as poor schools, peer influences, living arrangements, or lack of optimism about future labor market opportunities—are driving this relationship. Yet, as we clarify in our literature review, the state of knowledge regarding the broader relationship is sufficiently limited that we can make a substantial contribution focusing on that alone.
to a mother with a low level of education, to an unmarried mother, or to a
ten or minor mother.

Our cohort-based analysis allows us to answer the following hypotheti-
cal question: if we reduce the rate of socioeconomic disadvantage among
a birth cohort of women, then what impact does that have on their subse-
quent rate of early childbearing? Asked differently, to what extent is early
childbearing driven by socioeconomic disadvantage and its associated envi-
ronmental factors? We know from previous studies that women who grow
up “disadvantaged” are much more likely to give birth as teens. Our PSID
analysis confirms this strong correlation at the individual level. Our cohort
level analysis implies an even tighter intergenerational correlation between
rates of background disadvantage and early childbearing. But, when our
analysis econometrically controls for fixed state and year of birth effects in
the model to account for cultural and other differences across cohorts, the
relationship between rates of disadvantage and early childbearing is found
to be quite modest. For example, the elasticity of early childbearing rates
by age eighteen with respect to the probability of being born to a mother
under age eighteen is only 0.05. This suggests that broader, societal forces
are far more important in determining rates of early childbearing than rates
of socioeconomic disadvantage per se.

The remainder of the chapter proceeds as follows. Section 6.2 presents
some theoretical considerations that are designed to help think about what
we might expect regarding the relationship between socioeconomic disad-
vantage and early childbearing. Section 6.3 describes the results of past
research that may help guide our thinking. In section 6.4, we present our
analysis of PSID data, providing a descriptive analysis of this relationship.
Section 6.5 reports the details of our cohort-based analysis using aggregated
Vital Statistics natality data. We conclude in section 6.6.

6.2 Theoretical Considerations

Noneconomists typically attribute early childbearing to be the result of
myriad influences that affect a youth’s development and fall outside the
control of a rational decision-making process (e.g., Brooks-Gunn and
Brooks-Gunn and Furstenberg (1989) consider five perspectives on ado-
lescent sexual behavior: (a) biological perspectives; (b) parental influences;
(c) peer influences; (d) academic perspectives; and (e) social cognitive per-
spectives. In stark contrast to the economic model of rational decision mak-
ing, the authors note that “most teens do not consciously plan to become
sexually active, and they often do not foresee their first sexual experience. As
such, it frequently is not experienced as a decision but rather as something
that ‘happened,’” citing Chilman (1983, 251) on this last point.

A focus on biological perspectives emphasizes the role of hormonal factors
in driving the onset of sexual activity. In their consideration of parental influences the authors highlight research suggesting that teens who have good communication with their parents, teens who have feelings of connectedness and supportiveness with their parents, and teens with relatively more parental supervision tend to have later onset of intercourse. In discussing peer influences, the authors cite work suggesting that perceptions about what is normative in one’s peer group are more strongly associated with sexual behavior than the actual behavior of one’s peers. But the authors suggest that the presumed effects of parental and peer influences on teenage sexual behavior are stronger than the available research evidence indicates. A focus on academic perspectives emphasizes the observation that teenagers with lower academic success or aspirations are more likely to have sex as teenagers. And finally, their consideration of social cognitive abilities raises questions about the ability to integrate domain-specific knowledge into a coherent system (i.e., into an understanding of where babies come from). Other social cognitive processes that the authors point to as relevant and in need of greater research understanding include self-definitions, self-efficacy, and social comparisons; that is, how a teenage girl determines what it means to be a mature woman.

A more recent article by Brooks-Gunn and Paikoff (1997) moves even further from the traditional economic approach to the issue by suggesting that the study of adolescent sexuality must consider not only behaviors, but also feelings. They write that though insightful, the framework that they and others have used to consider adolescent sexual behavior, namely in the contexts of family, peer, neighborhood, biological, and cognitive characteristics, has been limiting. They propose four key topics that need to be explored in order to understand adolescent sexuality: (a) sexual well-being and developmental transitions; (b) the gendered nature of sexuality; (c) decision making and sexuality, and (d) the meaning of sexuality to youth.

Economists generally do not attempt to model parental behavior or the effect of feelings directly. But that does not mean that we ignore that these factors are potentially very important, perhaps crucial, to determining whether a young woman will engage in sexual activity and give birth as a teenager. Indeed, it is the correlation of these “other” factors with observable characteristics such as childhood poverty and growing up in a single-mother household that leads us to worry that the empirical associations between such background characteristics and early childbearing cannot be

2. The authors are quick to note that social and contextual effects will interact with the onset of hormonal changes: “So while very early sexual initiations may be in part hormonally mediated, by the time that behavior is normative, social factors may account for sexual initiation” (251). They cite research indicating that initiation of sexual behavior is highly associated with what is perceived as normative in one’s peer group. They further purport that racial differences in the initiation of intercourse prior to puberty speak to the importance of social and contextual factors on sexual behavior.
interpreted as causal. For example, if single mothers tend to be the type of women who would be less likely to supervise or communicate with their adolescent daughters regardless of marital status, then it is not the fact of being born to a single mother per se that leads daughters from single-parent homes to have relatively higher rates of early childbearing.

Work in other social sciences on this topic has tended to group the theoretical linkages between background characteristics and teen nonmarital childbearing into four categories, as helpfully summarized by An, Haveman, and Wolfe (1993): (a) the lifestyle characteristics of the parents; (b) information and network effects; (c) stressful childhood events; and (d) a utility maximization perspective. The first perspective emphasizes the intergenerational transmission of a culture of “welfare dependence.” The idea is that a girl growing up in a mother-only family where welfare receipt is the norm will develop preferences and behaviors that lead her to repeat such a lifestyle for herself. The second perspective holds that girls who grow up in poverty or without connections to the labor market will be more likely to engage in early childbearing than girls who grow up with economic resources and connections to a world that engenders career or educational ambitions. The third perspective is borne from research in sociology suggesting that stressful and unsettling events during childhood or early youth, such as changes in family structure, may lead to feelings of insecurity in young women. These types of feelings might lead a young girl to desire a baby or family of her own and therefore give birth as a teenager.

The fourth perspective, emphasizing a rational choice framework, tends to be the approach taken by economists who have written in this area (see, e.g., Leibowitz, Eisen, and Chow 1986; Duncan and Hoffman 1990; Lundberg and Plotnick 1995). This literature builds on the seminal work of Becker (1960) and Ward and Butz (1980) by modeling fertility as a decision-making process determined by economic factors. While Becker (1960) and Ward and Butz (1980) focus on the fertility of married women, their insight can easily be applied to the decision facing unmarried teenage women. Duncan and Hoffman (1990) is an early example of a study modeling the choice to have a nonmarital birth as a rational comparison of the income outcomes associated with the choice, modeled by these authors as welfare benefits versus expected income returns from career and a higher probability of marriage.

Recent work by behavioral economists offers some important modifications that may help synthesize a rational choice model with the perspectives of other social scientists. Work in this field argues that the rational-choice model is inaccurate in some systematic and important ways. A key insight for the issue of early childbearing is that when modeling decisions to undertake actions that involve immediate gratification and future costs—such as sex or smoking—otherwise rational individuals might exhibit “present-biased preferences” (e.g., O’Donahue and Rabin 1999; Laibson 1994). Such preferences are characterized by “excessive myopia,” whereby individuals
put additional weight on the present period relative to all future periods. When we talk about the actions of teens, such a model implies that teens might engage in too much risky behavior (say, unprotected sex) because they attach too little weight to their well-being as adults. In other words, they overly discount long-term consequences relative to short-term gratification. Economists tend to be uncomfortable normatively declaring that people’s behaviors are not in their best interest; but if individuals make decisions based on present-biased preferences, that is tantamount to them making decisions about present actions that they will regret later in life. Such preferences might also be considered a reflection of self-control problems. If an otherwise rational decision maker who is considering the costs and benefits of early childbearing has such present-biased preferences, or has self-control problems, then she might make decisions that lead to early childbearing, even if it is not in her long-term best interest.

O’Donoghue and Rabin (2001) review insights and issues raised by behavioral economists and psychologists that are relevant to modeling risky behavior by adolescents, including behaviors leading to unprotected sex and potentially early childbearing. Their reading of the behavioral/psychological evidence is that adolescents are similar to adults in terms of their ability to carry out the decision-making process but, importantly, that youths are less able to recognize the consequences associated with given actions. If minors are not fully capable of predicting the consequences of early childbearing—either for themselves or their children—they might initiate childbearing earlier than would be optimal from their own fully-informed perspective.

6.3 Relevant Empirical Literatures

Our analysis of the impact of socioeconomic disadvantage on early childbearing is related to several empirical literatures. Some are directly on point while others address different, albeit related, questions and can help inform our discussion and analysis. This section describes each of these related areas and what we can learn from them regarding the relationship between socioeconomic disadvantage and early childbearing.

6.3.1 The Costs of Teenage Childbearing

There is a fairly large literature in economics examining the consequences of giving birth at an early age. This literature has separately focused on the costs to the teen mother herself in the form of inferior subsequent outcomes (e.g., lower completed schooling and earned income) and the costs to her offspring (e.g., lower cognitive ability, higher rates of child poverty, and, potentially, early childbearing). Recent analyses have focused on methods designed to separately identify the role that early childbearing plays in altering these outcomes, abstracting from all of the other differences that exist between women who give birth at an early age and those who delay child-
bearing. Analysts have used approaches including examining differences in later-life outcomes between siblings who did and did not have a teen birth (Geronimus and Korenman 1992 and Geronimus, Korenman, and Hillemeier 1994); between teens who gave birth to those who were pregnant but miscarried (Hotz, McElroy, and Sanders 2005); and between teens who gave birth to twins to those who gave birth to singletons (Grogger and Bronars 1993). The motivation for these approaches is to find a relevant comparison group for teenage mothers whose average outcomes might reasonably be considered a proxy for what the average teen mother would have experienced had she not given birth before the age of eighteen or twenty. These studies generally find that the inferior outcomes observed for women who give birth as teens are largely due to underlying unobserved heterogeneity, as opposed to the teen birth itself.3

It is important to recognize that these studies are in some sense examining the reverse causal relationship from the one we address in this chapter. The studies previously described are designed to identify the impact of early childbearing on subsequent rates of disadvantage. Our analysis attempts to identify the link between early childhood disadvantage and the propensity to give birth at an early age. In fact, the studies described deliberately hold constant measures of childhood disadvantage in order to isolate the impact of teen childbearing from any confounding influence of disadvantage. This is precisely not what we try to do with our empirical analysis described following.

6.3.2 The Determinants of Teenage Childbearing

Two distinct literatures in economics focus directly on the determinants of teen childbearing. One focuses on the impact of policies and social conditions in a woman’s state of residence and the other examines the impact of personal characteristics. The policy-related literature is methodologically stronger, applying quasi-experimental methods designed to identify causal estimates of a policy’s impact on teen childbearing. Moffitt (1992, 1998, 2003) reviews the extensive literature on the effects of welfare policy on the incidence of female-headed households.4 Economists have also explored the role that abortion policy, and particularly parental involvement laws and Medicaid funding restrictions, play in teen fertility behavior (e.g., Levine, Trainor, and Zimmerman 1996; and Levine 2003). Kearney and Levine (2007) examine the cost of contraception, finding that publicly-funded

3. Ribar (1994) and Klepinger, Lundberg, and Plotnick (1999) use the age at menarche as an instrument for teen pregnancy, based on the observation that an earlier age at menarche leads to more years at risk of becoming pregnant. Ribar (1994) finds that this approach eliminates any negative effect of teen births on high school completion; Klepinger, Lundberg, and Plotnick (1999) find a negative causal effect.
4. Female-headed households are not necessarily headed by women who initiated childbearing as a teen, but there is a strong correlation between the two.
family planning can significantly reduce teen birth rates. The role of labor market conditions also has been examined (e.g., Dehejia and Lleras-Muney 2004; Levine 2001). Despite the methodological advantages of this work, the focus on specific policies and their impact on early childbearing is different than our goal of examining the role of socioeconomic disadvantage. But, these literatures do help guide our choice of other factors that are important to control for in our model.

The literature that is perhaps most closely aligned with our goals examines the impact of personal background characteristics on early childbearing outcomes. Haveman and Wolfe (1995) review a large number of studies in this area, highlighting the key contributions in this area of Duncan and Hoffman (1990), An, Haveman, and Wolfe (1993), and Lundberg and Plotnick (1995). An important contribution since this review is Duncan et al. (1998). These studies examine the relationship between factors related to economic disadvantage (including family income, parental education, and family structure) and the likelihood of giving birth at a young age and/or outside a marital union. Given that there are almost surely unobserved factors correlated with both an individual’s observed characteristics and her propensity for a teen or nonmarital birth, these individual-level studies are limited in their ability to identify a causal link between disadvantage per se and early childbearing.\(^5\) Two recent studies exploit arguably exogenous variation in female educational attainment in order to identify the causal link between education and teen childbearing (Black, Devereux, and Salvanes 2004; McCrary and Royer 2006).\(^6\) These studies are somewhat further removed from the focus of our analysis, as they concentrate on a woman’s own level of education. We are interested in the environment into which a young girl is born and are therefore interested in relating her mother’s level of education to her own likelihood of giving birth as a teen.

6.3.3 The Intergenerational Transmission of Income

As we describe following, our empirical analysis relates early childbearing as the outcome of interest to childhood disadvantage, which we define as being born to a mother with certain characteristics. When we use being born to a young mother as our measure of disadvantage and examine its links to early childbearing, we are conducting an exercise very similar in spirit to that considered in the literatures on intergenerational transmission of personal characteristics like income, education, and health. (For a

---

5. The methodological considerations of these studies tend to focus more on issues such as correlated errors in joint decision processes that include teen childbearing and welfare receipt, rather than on finding exogenous variation in measures of economic disadvantage.

review of these literatures, see Solon [1999].) These literatures examine the extent to which a parent’s characteristic is transmitted to his or her child. The earlier economics literature on this question almost always focused on the intergenerational transmission of income from fathers to sons. In a regression framework where son’s income is the dependent variable, the coefficient on father’s income represents the intergenerational correlation. The methodological issue frequently grappled with in this literature is how to appropriately measure income (say, by using multiple years of earnings) in order to reduce attenuation bias due to measurement error. More recent studies of intergenerational income correlation have considered the relationship between father’s or family income and daughter’s earnings or family income. These studies have addressed conceptual issues such as assortative mating (Chadwick and Solon 2002) and the transmission of gender preferences for work (Altonji and Dunn 2000).

More recent work in this area has focused on trying to identify the mechanisms for intergenerational correlations, asking whether the intergenerational correlation in income is attributable to an intergenerational correlation in health or education, for instance (e.g., Currie and Moretti 2007; Black, Devereaux, and Salvanes 2005). If one views part of our analysis as the intergenerational correlation in early childbearing, then this may be thought of as one of the mechanisms generating an intergenerational correlation in income.

6.4 Evidence from the Panel Study of Income Dynamics (PSID)

We begin our exploration of the empirical linkages between socioeconomic disadvantage and early childbearing with an examination of data from the Panel Study of Income Dynamics (PSID). The PSID is a longitudinal survey of a representative sample of U.S. individuals and their households. It was conducted annually from 1968 to 1997 and biannually since then. We study the cohort of women age twenty to thirty-five in the 2003 survey and observe particular circumstances of their births using information contained in the 1968 through 1983 survey files. We focus on five measures of disadvantage at birth: being born to a mother of age less than twenty or less than eighteen, being born to a single mother, being born to a mother who has not completed high school, and being born into a family whose income is at or below the federal USDA poverty threshold. We also consider two measures of socioeconomic disadvantage during adolescence:

7. In our work on early childbearing, measurement issues are less of a concern, particularly in our analysis of Vital Statistics data.
8. This would depend upon the extent to which early childbearing is causally related to low income. As we described earlier, this proposition is not perfectly clear on the basis of past research.
not living with married parents (or stepparents) at age fifteen and living in poverty at age fifteen.\footnote{In our exploration of linkages between background factors and early childbearing in the PSID, we do attempt to be as exhaustive in our set of variables considered as the PSID analyses of either An, Haveman, and Wolfe (1993) or Duncan et al. (1998), which look at much larger sets of demographic characteristics for earlier cohorts of young women. We merely look to the PSID for descriptive purposes. In the attempt to uncover causal relationships, we rely primarily on a cohort-based analysis of vital statistics natality data.}

Our PSID sample consists of 1,797 women age twenty to thirty-five observed in the 2003 survey and back to the year of their birth. Table 6.1 reports rates of disadvantage and rates of teen childbearing by disadvantage factor. Among our sample of women, 24 percent gave birth before age twenty.\footnote{One potentially confusing finding in these data is that 24 percent of women gave birth to children before age twenty, but only 14 percent of them were born to mothers under twenty years old. The reason for the discrepancy is that not all of the women in our PSID sample are firstborn children. So, the number of women born to teen mothers understates the number of women in that older cohort who actually had a \textit{first} birth during their teen years.} The rates of teen childbearing are dramatically higher for women who were born with each of our four measures of disadvantage. Forty-six percent of women born to teen mothers give birth as teenagers themselves; 43 percent of women are born to an unmarried mother; 44 percent of women are born to a mother with less than a high school degree (including teenagers); and 49 percent of women are born into poverty. Similar increases in the likelihood of giving birth by age eighteen are also observed for women from disadvantaged backgrounds.

One interesting finding in table 6.1 is that the heightened propensity to give birth as a teen relative to the full sample is about the same for any of the disadvantage factors considered. In particular, being born to a teen mother, an unmarried mother, or a less-educated mother has about the same impact on rates of early childbearing as does being born to a poor mother, which is perhaps the measure that is most closely associated with the notion of childhood disadvantage. This is important to keep in mind when we move to our cohort-based analysis, where poverty status at birth is something that we are unable to measure.

We also use these data to estimate raw intergenerational correlations in teen childbearing in a manner consistent with past research on intergenerational correlations in economic outcomes. In particular, we run simple regressions of own teen childbearing behavior on an indicator variable for whether each woman was born to a teen mother. We conduct an analogous exercise for births before age eighteen. It is important to recognize that these simple models are only designed to identify correlations; results should not necessarily be interpreted as causal.

The results of this exercise are reported in the top panel of table 6.2. They indicate that women who were born to teen mothers are 25 percentage points more likely to go on to give birth as teens themselves; given that the mean
The rate of teen childbearing is 24 percent, this is roughly a doubling of the odds. The results for births by age eighteen are not statistically significant, likely due to the very small number of children born to mothers under age eighteen in the sample.

The remainder of table 6.2 reports the results of multivariate regression specifications where the dependent variable is defined as giving birth before the age of twenty and giving birth before the age of eighteen, and combinations of measures of disadvantage are included as explanatory variables. The results from these models can be interpreted as descriptive only, but they establish a form of “horse race” between disadvantage factors that provide at least one gauge of their relative importance. The six measures of socioeconomic disadvantage examined include the following: born to a teen mother, born to an unmarried mother, born to a mother with less than a high school degree, born into poverty, not living with married parents at age fifteen, and living in poverty at age fifteen. The coefficient estimates from these specifications imply statistically significant and substantial increases in rates of early childbearing among young women associated with most of these measures of disadvantage, even after controlling for the other correlated measures of disadvantage.

It is important to keep in mind that all of the results reported so far are purely descriptive in nature, representing correlations, and are not designed to tease out causal findings. The next section of this chapter will move more in that direction.

### Table 6.1: Rates of early childbearing by disadvantage factor

<table>
<thead>
<tr>
<th>Disadvantage Factor</th>
<th>% with disadvantage</th>
<th>% gave birth before age 20</th>
<th>% gave birth before age 18</th>
</tr>
</thead>
<tbody>
<tr>
<td>All</td>
<td>—</td>
<td>0.24</td>
<td>0.12</td>
</tr>
<tr>
<td>Born to mother less than age 20</td>
<td>0.14</td>
<td>0.46</td>
<td>0.26</td>
</tr>
<tr>
<td>(n = 1,797)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Born to mother less than age 18</td>
<td>0.04</td>
<td>0.43</td>
<td>0.24</td>
</tr>
<tr>
<td>(n = 1,797)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Born to unmarried mother</td>
<td>0.28</td>
<td>0.45</td>
<td>0.23</td>
</tr>
<tr>
<td>(n = 1,743)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Born to mother with less than HS degree</td>
<td>0.28</td>
<td>0.44</td>
<td>0.26</td>
</tr>
<tr>
<td>(n = 1,266)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Born into poverty</td>
<td>0.13</td>
<td>0.49</td>
<td>0.26</td>
</tr>
<tr>
<td>(n = 1,611)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Not living with married parents at age 15</td>
<td>0.45</td>
<td>0.39</td>
<td>0.21</td>
</tr>
<tr>
<td>(n = 1,412)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Living in poverty at age 15</td>
<td>0.04</td>
<td>0.53</td>
<td>0.38</td>
</tr>
<tr>
<td>(n = 1,553)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Notes:** The sample is comprised of women age twenty to thirty-five in the 2003 PSID. Estimates are similar when we use a uniform sample size across measures.
In this section of the chapter, we use Vital Statistics birth data to explore the relationship between early childbearing among a birth cohort compared to the share of the cohort that was born to young mothers. One advantage that a cohort-based approach has compared to an analysis of a woman’s own childbearing experiences relative to her mother is that a lot of the unobserved heterogeneity across women is averaged out. This is a property that will assist our analysis using these data, as described in more detail subsequently.

### 6.5.1 Data Description

We now turn to an exploration of data from the Vital Statistics Natality Detail Files between 1968 and 2003. These data represent individual records on births that took place in the United States. We use data from 1968 to

<table>
<thead>
<tr>
<th>Dependent variable</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gave birth by age 20</td>
<td>0.250</td>
<td>—</td>
<td>0.043</td>
<td>—</td>
</tr>
<tr>
<td>Gave birth by age 18</td>
<td>(0.041)</td>
<td>(0.061)</td>
<td>(0.067)</td>
<td>(0.067)</td>
</tr>
</tbody>
</table>

**Notes:** Estimates are the results from linear probability models that include no other covariates besides those listed.

### 6.5 Analysis of Vital Statistics Birth Data

In this section of the chapter, we use Vital Statistics birth data to explore the relationship between early childbearing among a birth cohort compared to the share of the cohort that was born to young mothers. One advantage that a cohort-based approach has compared to an analysis of a woman’s own childbearing experiences relative to her mother is that a lot of the unobserved heterogeneity across women is averaged out. This is a property that will assist our analysis using these data, as described in more detail subsequently.

**Table 6.2 Relationship between early childbearing and measures of disadvantage**

<table>
<thead>
<tr>
<th>Dependent variable</th>
<th>Gave birth by age 20</th>
<th>Gave birth by age 18</th>
</tr>
</thead>
<tbody>
<tr>
<td>Born to mom &lt; 20 (col. [1] and [2]) or 18 (col. [3] and [4])</td>
<td>0.250</td>
<td>—</td>
</tr>
<tr>
<td></td>
<td>(0.041)</td>
<td>(0.061)</td>
</tr>
<tr>
<td>Born to mom &lt; 20 (col. [1] and [2]) or 18 (col. [3] and [4])</td>
<td>0.014</td>
<td>—</td>
</tr>
<tr>
<td></td>
<td>(0.045)</td>
<td>(0.049)</td>
</tr>
<tr>
<td>Born to single mom</td>
<td>0.186</td>
<td>0.092</td>
</tr>
<tr>
<td></td>
<td>(0.031)</td>
<td>(0.037)</td>
</tr>
<tr>
<td>Born to mom &lt; HS grad</td>
<td>0.198</td>
<td>0.191</td>
</tr>
<tr>
<td></td>
<td>(0.030)</td>
<td>(0.033)</td>
</tr>
<tr>
<td>Born into poverty</td>
<td>0.075</td>
<td>0.112</td>
</tr>
<tr>
<td></td>
<td>(0.041)</td>
<td>(0.028)</td>
</tr>
<tr>
<td>Age 15—Not living w/ married parents</td>
<td>—</td>
<td>0.112</td>
</tr>
<tr>
<td></td>
<td>(0.028)</td>
<td>(0.032)</td>
</tr>
<tr>
<td>Age 15—Living in poverty</td>
<td>—</td>
<td>0.093</td>
</tr>
<tr>
<td></td>
<td>(0.068)</td>
<td>(0.054)</td>
</tr>
<tr>
<td>Constant</td>
<td>0.138</td>
<td>0.112</td>
</tr>
<tr>
<td></td>
<td>(0.014)</td>
<td>(0.017)</td>
</tr>
<tr>
<td>Sample size</td>
<td>1,213</td>
<td>1,022</td>
</tr>
<tr>
<td>Adjusted $R^2$</td>
<td>0.116</td>
<td>0.117</td>
</tr>
</tbody>
</table>

11. These data are first available starting in 1968 and 2003 is the last year currently available. From 1985 onward, these data represent a complete count of births. Prior to 1972, births were sampled at a 50 percent rate nationwide. In the intervening period, some states sampled at a 50 percent rate and others included all births. In our analysis, we applied appropriate weights to provide estimates of all births.
1986 to identify the number of women born in each state and year along with the proportion of those births that can be classified as “disadvantaged.” The alternative measures of disadvantage for those women born in these years include having a mother who is a high school dropout, unmarried, or under age eighteen or twenty at the time of giving birth. These data provide us with rates of “disadvantage” for the women born in these birth cohorts and also provide us with a denominator for a measure of the rate of early childbearing that these women subsequently experience.

To get the numerator for this early childbearing statistic, we use data from the 1980 through 2003 Vital Statistics file to tally births born to women less than age eighteen or age twenty. These data allow us to assign every birth that takes place in the United States to the mothers’ state and year of birth. From these data, we tally all first births that occur to women less than age eighteen or twenty from that state/year of birth cohort. Dividing this count by the size of that cohort provides a measure of the rate of early childbearing.

Consider, for instance, the 1970 birth cohort from New York. We use the 1970 Vital Statistics natality file to tally the total number of females born in that state and year, which becomes the denominator for our early childbearing statistic. That data file also allows us to identify the number born to teen/minor mothers, to unmarried mothers, and to mothers with less than a high school degree. We use this information to construct the rates of “disadvantage” in this cohort. To calculate the numerator for our early childbearing statistic, we sum the number of first births in 1983 to thirteen-year-old mothers born in New York, the number of first births in 1984 to fourteen-year-old mothers born in New York, and so on through the 1989 file, when the 1970 birth cohort would have been age nineteen. That sum represents the number of girls born in New York in 1970 who gave birth by age twenty.

The ability to link births to mothers’ birth cohorts is crucial for our purposes. In effect, we are linking three generations: we look at birth records in a given year and identify the age—and birth cohort—of the mother. We then look to the birth records for that cohort of teenage mothers and identify the characteristics of their mothers. The completeness of Vital Statistics birth data is high.

12. We also experimented with a measure of the poverty rate, but decided not to include it in our analysis because of data limitations. Poverty rates by state and year are not available for the birth cohorts in our sample. Instead, we tried using county level data from the 1970 Census, attaching to each birth cohort the poverty rate that existed in the relevant county of birth. As the geographic composition of births changed over time within a state, this measure would provide some within state variation in poverty. Unfortunately, we found that this variation was insufficient to provide robust parameter estimates.

13. When we consider births before age twenty, we can only use birth cohorts through 1984.

14. For a very small number of births, this information is missing. These births are not included in the analysis.
records substantially reduces measurement error in our estimated rates of early childbearing. Nevertheless, the data are not perfect. First, our identification of a teen mother’s birth cohort relies on the reporting of a mother’s age on the birth record; the natality data does not report the exact year in which the mother was born. We simply subtract the mother’s age from the year she gave birth to determine her birth year, so there may be a misclassification by up to one year in the mother’s birth cohort associated with an early birth.\textsuperscript{15}

Another important limitation of these data is that information on mother’s education and marital status is not complete in the years in which we are measuring the size of birth cohorts and recording their rates of disadvantage. First, data on maternal education is not available at all for the 1968 birth cohort and they exist for only a subset of states for the years between 1969 and 1979. To balance the panel when we use this variable, we include in our analysis just the thirty-six states for which these data are available in all years. Similarly, direct information on marital status of the mother is only available in all years for thirty-seven states; we focus our attention on just those states in analyses regarding marital status.\textsuperscript{16}

6.5.2 Descriptive Analysis

Figures 6.1 through 6.4 provide a description of these data, separately considering the conditions at birth for these cohorts of women along with their subsequent rates of early childbearing. Figure 6.1 displays trends in the percentage of each birth cohort that exhibit each form of disadvantage. The first conclusion that one can draw from this figure is that there are very distinct trends over time that differ across measures of disadvantage. The percentage of birth cohorts that are born to less-educated mothers (defined as not having completed high school) has fallen rather consistently from slightly more than 31.7 percent among the 1969 birth cohort to 20 percent for the 1986 birth cohort. The fraction of a birth cohort born to a mother less than age twenty has likewise fallen from a high of 19.7 percent among the 1973 birth cohort to 12.5 percent among the 1986 birth cohort. On the other hand, the fraction of each birth cohort born to an unmarried mother has risen continuously over this period (and beyond). In the youngest birth cohort we study, nearly one in four women (23.1 percent) were born to unmarried mothers, compared to roughly one in ten among the oldest birth cohort. These secular changes over time, driven by other social

\textsuperscript{15} Another minor limitation of these data is that births to women who were born in the United States but gave birth in another country would not be captured in these data. It is our impression that this is a very infrequent event and we ignore it here.

\textsuperscript{16} The set of states with missing information on maternal education is the following: AL, AR, CA, CT, DE, DC, FL, GA, ID, MD, NM, OR, PA, TX, and WA. The set of states with inadequate data on marital status is: CA, CT, GA, ID, MD, MA, MT, NM, NY, OH, VT, MI, NV, and TX.
forces (cohort fixed effects), are the type of variation in the data that needs to be held constant in an analysis trying to identify the causal impacts of disadvantage.

Similar variation exists in these measures of disadvantage across states at a point in time. Figure 6.2 displays one representative measure, the percentage of a birth cohort born to teen mothers, across states. Among the 1969 birth cohort, over 20 percent were born to teen mothers in Mississippi and Alabama, but less than half that percentage in Massachusetts and Minnesota. Clearly, at least a portion of this gap may be attributable to different social customs that exist across these states. This is apparent in the fact that most states that have rates of births to teen mothers in 1969 also have relatively high rates in 1983. Again, these long-standing differences in state attitudes toward fertility (state fixed effects) need to be held constant if we are going to attempt to identify the causal impact of disadvantage.

It is interesting to note in figure 6.2 that the change between 1969 and 1983 within states in the percentage of a birth cohort born to teen mothers varies considerably across states. This rate fell in every state over the fourteen-year period, but it fell by more in some states than others. For instance, Rhode Island and New Mexico experienced a trivial drop, but the percentage of a birth cohort born to teen mothers fell by roughly one-third in Washington and Kansas. Although it is not clear that the variation in changes in measures of birth cohort disadvantage—such as being born to a teen mother—

![Fig. 6.1 Trends in conditions at birth](image-url)
over time between states is purely exogenous to preferences/tastes for early childbearing (as we will discuss later), focusing on these within-state, across-birth-cohort changes as our identifying source of variation removes the confounding influences of secular changes that occurred over time nationally and long-standing cross-sectional variation across states.

Figures 6.3 and 6.4 provide analogous information about the patterns in early childbearing for women in these birth cohorts. Figure 6.3 displays the aggregate trend over time. It shows that the percentage of women giving birth either before age eighteen or before age twenty rose somewhat among women born for the early 1970s birth cohorts, but then fell subsequently. This pattern roughly corresponds to the aggregate trends in annual teen childbearing rates, which spiked in the late 1980s and early 1990s. Figure 6.4 displays cross-sectional variation in the percentage of women giving birth by age twenty (similar patterns exist for births by age eighteen). Again, there is substantial cross-sectional variation in early childbearing rates. Among women in the 1969 birth cohort, about 30 percent in Mississippi and the District of Columbia gave birth by age twenty. The comparable figure for women born in Massachusetts and Connecticut is closer to 10 percent. For women born in most states, the rate of early childbearing fell somewhat between the 1969 and 1983 birth cohorts. Importantly, there is considerable

17. No data is available for births by age nineteen beyond the 1984 birth cohort because 2003 is the most current Vital Statistics data available.
**Fig. 6.3**  Trends in rates of early childbearing

**Fig. 6.4**  Cross-sectional variation in percentage of birth cohorts giving birth by age 20
variation across states in the extent of the reduction in rates of early child-
bearing between 1969 and 1983.

6.5.3 Econometric Specification

Our empirical approach relates the variation in state-year birth cohort
teen childbearing rates to the variation in state-year birth cohort rates of
disadvantage. The thought experiment that corresponds to this econometric
exercise is the following: if we could reduce the amount of socioeconomic
disadvantage among a birth cohort of women, what impact would that have
on their rates of early childbearing down the road? If that impact were large,
we would infer that early childbearing is in large part a consequence of being
born into socioeconomic disadvantage, as so measured. If that impact were
small in magnitude, however, we would conclude that teen childbearing is
driven only in small measure by childhood disadvantage.

In our more formal econometric discussion, we develop our thinking
regarding the models that we estimate and report, highlighting their strengths
and limitations as well as appropriate interpretations. Let us begin by con-
sidering the following cross-sectional, individual-level model. The outcome
variable is a binary indicator for early childbearing (EC). It is modeled as a
function of some measure of economic disadvantage in the individual's past
(D) and other personal characteristics (X1) for individual i.

\[
EC_i = \beta_0 + \beta_1 D_i + \Gamma_i X_{1i} + \nu_i.
\]

This equation is a simplification of the approaches reviewed in Haveman and
Wolfe (1995), emphasizing the cross-sectional nature of the variation in the
data available. The estimate of coefficient \( \beta_1 \) is interpretable as the relation-
ship between the disadvantage measure and early childbearing, conditional
on the control variables included in the \( X \)-vectors.

The main limitation of equation (1) is that there are likely to be factors
that are related to both the measure of disadvantage and the early child-
bearing outcome that are not controlled for and may even be very difficult
or impossible for the researcher to observe. As such, they lead to omitted
variable bias that precludes assigning a causal interpretation to \( \beta_1 \). Suppose,
for instance, that a woman's own values lead her to have a child early and
to instill those values in her daughters, leading them to have children early
as well. If those values are related to lower socioeconomic outcomes, then
we may erroneously attribute the relationship in early childbearing across
generations to economic disadvantage.

We can move closer to identifying a causal relationship between \( D \) and \( EC \)
by controlling for state and cohort fixed effects in the model. For instance,
early childbearing outcomes for women who reached age twenty in 1996 are
likely to be different than their counterparts who reached age twenty in 1986,
due to differences in the cultural climates in the years in which they were
developing values and making relevant choices. Likewise, there are cultural
differences across communities that exert influence on individual decisions and behaviors. We thus augment the model as follows:

\[ EC_{ics} = \beta_0 + \beta_1 D_{ics} + \Gamma_1 X_{ics} + \gamma_c + \gamma_s + \eta_{ics}. \]

In this specification, each individual \(i\) is identified by her “community” (state in this example, denoted by \(s\)) and period (or cohort—denoted by \(c\)). This approach eliminates two possible forms of unobservable heterogeneity. To the extent that rates of disadvantage vary by cohort and by state, and to the extent that time and place matter to early childbearing decisions, the inclusion of these controls in the model will alleviate some of the omitted variable bias.

We can go one step further in removing individual-level heterogeneity by estimating this model at the aggregate level. This has the effect of averaging out differences across individuals within cohorts/states. There is sure to be a great deal of variation in individual values that may be related to disadvantage and early childbearing decisions and this variation is greatly reduced when aggregated. So, instead of estimating equation (2) with individual-level data, we estimate this relationship at the level of a state/year birth cohort:

\[ EC_{cs} = \beta_0 + \beta_1 D_{cs} + \Gamma_1 X_{ics} + \gamma_c + \gamma_s + \eta_{ics}. \]

This differs from equation (2) in the subscripts, indicating the aggregated nature of the data within cohorts. Each variable is now the average over individuals within the state/birth cohort cell.

We control for a set of \(X\) variables that varies at the level of state/year birth cohort. At the cohort level, rates of childbearing will vary with the average demographic composition of a cohort. We explicitly control for marital status, age, racial/ethnic composition, and level of educational attainment among women fifteen to forty-four in the state/year cohort. Any remaining demographic differences are captured with cohort fixed effects, to the extent that those differences are geographically uniform at a point in time. The vector of control variables in equation (3) also includes a set of variables that are designed to measure environmental conditions around the time that early childbearing decisions are being made. The specific variables we include are the unemployment rate, to capture economic conditions, and an extensive set of policy variables, including abortion restrictions, welfare benefit levels and reform indicators, and State Children’s Health Insurance Program (SCHIP) implementation.18 Further details regarding the specific

18. Specifically, these variables include indicators for Temporary Assistance for Needy Families (TANF) implementation, the presence of a welfare family cap, maximum Aid to Families with Dependent Children (AFDC)/TANF benefits for a family of three, SCHIP implementation, Medicaid coverage of abortion, abortion parental notification requirements, and abortion delay rules. We also control for cohort size, although the results are virtually unchanged when cohort size is excluded from the model.
variables used and their sources are available in Kearney and Levine (2007). We simplify the measurement of these variables by including their values at the time each birth cohort was age seventeen or age nineteen in specifications in which our measure of early childbearing is births by age eighteen and births by age twenty, respectively.

How should we interpret the results of estimating equation (3)? As stated previously, the thought experiment captured by this cohort specification is the following: if we were to “shock” the amount of socioeconomic disadvantage \((D)\) of a birth cohort of women, what impact would that have on cohort rates of early childbearing \((EC)\)? Two points of comparison and contrast with equation (2) are important. First, equation (2) relates a particular background characteristic, say being born in poverty, and early childbearing outcomes. As discussed, the existence of omitted variable bias in such equations almost surely make a causal interpretation inappropriate. We have not completely surmounted this issue of omitted variable bias with our cohort specification. There might be remaining unobserved differences across state/year birth cohorts that correlate with both our measures of disadvantage and early childbearing outcomes. For example, it is possible that values change within states across birth cohorts in a way that is correlated with both economic disadvantage and early childbearing. As we note following, a comparison of results from the estimation of various specifications of equation (3) suggest that these possibilities are probably not that important.19

The second important issue relevant to the interpretation of the coefficient, \(\beta_1\), is that when estimating this equation at the cohort level, \(\beta_1\) incorporates spillover or peer effects. When we “shock” the amount of disadvantage in a state/year birth cohort, that might have a general effect on sexual behavior and fertility outcomes even for those women whose own background characteristic is not changed. Consider the possibility that the tastes and values of girls born to married women are to some extent influenced by the proportion of girls in their cohort born to unmarried mothers. We remain agnostic as to how such spillover or peer effects operate in this context.

In general, we cannot fully characterize the complexities of the mechanism by which socioeconomic disadvantage may affect early childbearing.

19. One methodological approach that would, in principle, be useful to eliminate this remaining source of unobservable heterogeneity is instrumental variables. To implement this approach, one would need instruments that are correlated with our measures of disadvantage without being correlated with the residual in early childbearing. The difficulty in identifying such an instrument is that any variable that shifts the characteristics of a birth cohort, like the fraction born to teen mothers, is likely to alter other characteristics of that cohort, like its values/tastes. For instance, we experimented with using variation in abortion legalization across states and over time in the early 1970s, like Ananat et al. (2004). The problem with this is that if abortion legalization changed the selection process determining who is born (like the percentage to teen mothers), it is likely also to affect attributes of that birth cohort other than their socioeconomic disadvantage (like their tastes/values). Without any other obvious suitable instruments, we have chosen not to pursue this estimation strategy.
When we define $D$ in equation (3) as, say, the proportion of a state/year birth cohort born to a teen mother, we are estimating the relationship between the prevalence of this factor and cohort rates of early childbearing. This does not identify the isolated impact of teen childbearing itself; when we “shock” teen childbearing, we necessarily shock all associated factors that are not directly controlled for in the model, such as the values of the women, their parenting styles, their career aspirations, their relationship formations, and so forth. So, we do not purport to get inside the black box of what it is about a particular characteristic that leads to changes in teen childbearing propensities; we claim only to empirically estimate the broader relationship.

6.5.4 Results

The results of our analysis using Vital Statistics data are reported in tables 6.3 and 6.4. The left panel of these tables uses data from the 1968 through 1986 birth cohorts and considers a dependent variable measuring their rate of early childbearing, defined as giving birth before age eighteen. The mean value of this early childbearing rate across states and birth cohorts is 8.4 percent. The right panel uses data from the 1968 through 1984 birth cohorts and is comparable, but defines early childbearing to be that occurring before age twenty; the mean of this rate is 20.1 percent.

The first row of table 6.3 presents sample means for each measure of disadvantage at birth for the relevant subset of birth cohorts. These data indicate that about 26 percent of the women in these birth cohorts were born to mothers who had not yet completed high school, about 16 percent to mothers who were not yet married, over 16 percent to mothers who were teens, and over 6 percent to mothers who were minors.

The first panel of regression results in table 6.3 reflects estimates from simple regressions where the dependent variable is the relevant rate of early childbearing for the birth cohort and the sole independent variable is the rate of economic disadvantage at birth among that state/year female birth cohort. These results show a very strong intergenerational linkage between disadvantage at birth and subsequent rates of early childbearing. These figures indicate that a birth cohort with an additional 1 percentage point higher rate of children in it born to minors is associated with about a 1 percentage point higher rate of childbearing as minors themselves. A similar result is obtained for births by age twenty as a function of the proportion of the cohort born to teen mothers.

This estimated relationship of a nearly one-to-one correspondence in rates of teen childbearing across generations stands in contrast to the estimated relationship at the individual level in the PSID. Recall that in that analysis we found that having been born to a teen mother increases one’s likelihood of having a teen birth by 25 percentage points. This comparison suggests that early childbearing is much more tightly linked across generations in the aggregate. We interpret these findings as suggesting that the
### Table 6.3  Estimates of the impact of alternative measures of economic disadvantage on early childbearing

<table>
<thead>
<tr>
<th>Measure of disadvantage</th>
<th>Mother with low education</th>
<th>Unmarried mother</th>
<th>Born to teen</th>
<th>Born to minor</th>
</tr>
</thead>
<tbody>
<tr>
<td>Proportion w/ disadvantage</td>
<td>0.257</td>
<td>0.162</td>
<td>0.163</td>
<td>0.061</td>
</tr>
<tr>
<td>Coefficient/standard error</td>
<td>0.261</td>
<td>0.184</td>
<td>0.550</td>
<td>1.097</td>
</tr>
<tr>
<td>(0.021)</td>
<td>(0.040)</td>
<td>(0.033)</td>
<td>(0.059)</td>
<td></td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.635</td>
<td>0.168</td>
<td>0.746</td>
<td>0.776</td>
</tr>
<tr>
<td>No other covariates</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Coefficient/standard error</td>
<td>0.106</td>
<td>0.060</td>
<td>0.122</td>
<td>0.250</td>
</tr>
<tr>
<td>(0.022)</td>
<td>(0.053)</td>
<td>(0.063)</td>
<td>(0.114)</td>
<td></td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.982</td>
<td>0.968</td>
<td>0.974</td>
<td>0.974</td>
</tr>
<tr>
<td>State and birth cohort fixed effects</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Coefficient/standard error</td>
<td>0.137</td>
<td>0.057</td>
<td>0.129</td>
<td>0.272</td>
</tr>
<tr>
<td>(0.022)</td>
<td>(0.040)</td>
<td>(0.056)</td>
<td>(0.095)</td>
<td></td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.983</td>
<td>0.974</td>
<td>0.98</td>
<td>0.98</td>
</tr>
<tr>
<td>State and birth cohort fixed effects along with additional policy and demographic control variables</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Coefficient/standard error</td>
<td>0.161</td>
<td>0.155</td>
<td>0.187</td>
<td>0.342</td>
</tr>
<tr>
<td>(0.036)</td>
<td>(0.065)</td>
<td>(0.074)</td>
<td>(0.128)</td>
<td></td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.986</td>
<td>0.985</td>
<td>0.985</td>
<td>0.986</td>
</tr>
<tr>
<td>Number of observations</td>
<td>612</td>
<td>666</td>
<td>918</td>
<td>918</td>
</tr>
</tbody>
</table>

Notes: Additional policy and demographic control variables include the unemployment rate, indicator variables for the implementation of a welfare waiver (pre-TANF) or TANF, and the implementation of a welfare family cap, log maximum welfare benefits for a family of three, an indicator for SCHIP implementation, abortion policy indicators including the presence of a Medicaid funding restriction, parental notification law and mandatory delay law, and aggregate demographic characteristics of women of childbearing age, including the percent married, percent white, percent Hispanic, and the percent who are high school dropouts, high school graduated, and have attended some college. All these variables are measured in the year the cohort turned age eighteen (left panel) or twenty (right panel).
culture or environment that supports teen childbearing is more deterministic of early childbearing outcomes than the specific influence of the individual’s mother and her childbearing experiences. This is an example of what we referred to earlier as spillover or peer effects that may lead to differences in results based on aggregate data as opposed to microdata.

The remainder of table 6.3 provides additional evidence that cultural factors play a very important role in explaining early childbearing. In the second panel of the table, we report ordinary least squares (OLS) estimates in models that are augmented by state and birth cohort fixed effects, but with no other covariates. In these models, the estimated coefficient on the relevant measure of disadvantage is greatly attenuated compared to the results from the simple regression models. Although considerably smaller in magnitude, many of the estimates still are statistically significant, at least at the 10 percent level.

The coefficient on the proportion of a state/year birth cohort born to a minor in a model where the dependent variable is defined as giving birth before age eighteen falls from 1.097 (standard error of 0.059) to 0.250 (standard error of 0.114). This estimate implies that a cohort with a 1 percentage point higher rate of being born to a minor (a 16 percent increase from a 6.1 percent base) will, net of state and birth year fixed effects, have a roughly 0.25 percentage point higher rate of childbearing as minors themselves. With a mean rate of childbearing as minors of 8.4 percent, this is a very small impact for a sizeable intervention.

Similarly, the estimated coefficient on the proportion of a state/year birth cohort born to mothers without a high school degree (base rate of 25.7 percent) in the model focusing on births by age eighteen falls from 0.261 (standard error of 0.021) to 0.106 (standard error of 0.022). This means that if the rate of less educated mothers fell from 25.7 percent to, say, 20.7 percent, then the rate of early childbearing by age eighteen would fall by about half a percentage point. Again, given that the mean rate of childbearing before age eighteen is 8.4 percent, this is not a very large impact for a substantial reduction in the rate of socioeconomic disadvantage.

Table 6.4 replicates the analysis in table 6.3, reporting models of the form described by equation (3), but in log-log form so that the coefficients can

---

20 One reason that we would expect the intergenerational correlation in early childbearing to be smaller in microdata is that outcomes are determined by probabilities even if behavior is changed with certainty. As an extreme example to clarify the point, suppose that all children born to young mothers are willing to engage in unprotected sexual intercourse at a young age themselves. Yet because of the uncertainties associated with finding a partner, having sex, getting pregnant, and carrying the pregnancy to term, one would not expect all of these women to have children at an early age themselves. One could imagine a larger impact in aggregate data if communities in which more women became mothers at an early age change the social norms of behavior for all teens. In that case, the hypothesized changes in behavior among more young women could lead to larger changes in outcomes even after factoring in all the relevant probabilities.
Table 6.4  Elasticity estimates of the impact of alternative measures of economic disadvantage on early childbearing

<table>
<thead>
<tr>
<th>Measure of disadvantage</th>
<th>Mother with low education</th>
<th>Unmarried mother</th>
<th>Born to teen</th>
<th>Born to minor</th>
<th>Mother with low education</th>
<th>Unmarried mother</th>
<th>Born to teen</th>
<th>Born to minor</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Coefficient/standard error</td>
<td>0.852</td>
<td>0.256</td>
<td>1.115</td>
<td>0.861</td>
<td>0.623</td>
<td>0.227</td>
<td>0.915</td>
<td>0.669</td>
</tr>
<tr>
<td></td>
<td>(0.055)</td>
<td>(0.076)</td>
<td>(0.059)</td>
<td>(0.041)</td>
<td>(0.050)</td>
<td>(0.063)</td>
<td>(0.052)</td>
<td>(0.040)</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.646</td>
<td>0.095</td>
<td>0.78</td>
<td>0.799</td>
<td>0.557</td>
<td>0.127</td>
<td>0.803</td>
<td>0.792</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>State and birth cohort fixed effects</td>
<td>0.252</td>
<td>0.103</td>
<td>0.101</td>
<td>0.049</td>
<td>0.196</td>
<td>0.000</td>
<td>0.124</td>
<td>0.054</td>
</tr>
<tr>
<td></td>
<td>(0.120)</td>
<td>(0.067)</td>
<td>(0.100)</td>
<td>(0.082)</td>
<td>(0.075)</td>
<td>(0.041)</td>
<td>(0.055)</td>
<td>(0.042)</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.983</td>
<td>0.981</td>
<td>0.979</td>
<td>0.979</td>
<td>0.988</td>
<td>0.985</td>
<td>0.985</td>
<td>0.985</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>State and birth cohort fixed effects along with additional policy and demographic control variables</td>
<td>0.282</td>
<td>0.052</td>
<td>0.129</td>
<td>0.054</td>
<td>0.229</td>
<td>0.002</td>
<td>0.140</td>
<td>0.065</td>
</tr>
<tr>
<td></td>
<td>(0.088)</td>
<td>(0.055)</td>
<td>(0.092)</td>
<td>(0.063)</td>
<td>(0.056)</td>
<td>(0.036)</td>
<td>(0.046)</td>
<td>(0.029)</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.987</td>
<td>0.985</td>
<td>0.984</td>
<td>0.984</td>
<td>0.99</td>
<td>0.987</td>
<td>0.987</td>
<td>0.987</td>
</tr>
<tr>
<td>Number of observations</td>
<td>612</td>
<td>666</td>
<td>918</td>
<td>918</td>
<td>576</td>
<td>629</td>
<td>867</td>
<td>867</td>
</tr>
</tbody>
</table>

Notes: All measures of disadvantage are included in the right-hand side as the log of the proportion in the cohort with each disadvantage. Standard errors are clustered at the state level. Additional policy and demographic control variables include the unemployment rate, indicator variables for the implementation of a welfare waiver (pre-TANF) or TANF, and the implementation of a welfare family cap, log maximum welfare benefits for a family of three, an indicator for SCHIP implementation, abortion policy indicators including the presence of a Medicaid funding restriction, parental notification law and mandatory delay law, and aggregate demographic characteristics of women of childbearing age, including the percent married, percent white, percent Hispanic, and the percent who are high school dropouts, high school graduated, and have attended some college. All these variables are measured in the year the cohort turned age eighteen (left panel) or twenty (right panel). Standard errors are clustered at the state level. Regressions are weighted by cohort size.
be interpreted as elasticities. Once we include fixed effects, we see that a 10 percent decrease in the proportion born to a mom with less education than a high school degree is associated with a 2.5 percent reduction in the rate of childbearing by age eighteen. The analogous decrease in the proportion born to a teen mom is also associated with a decrease of approximately 2 percent. These are the two largest elasticities that we observe and they are not that large. Other elasticities are considerably smaller. For instance, a 10 percent reduction in the proportion of a cohort born to minors only reduces the rate of childbearing by age eighteen in that cohort by about 0.5 percent.

It is interesting to note that once the model controls for state and birth year fixed effects, there no longer appears to be a statistically significant relationship between the percentage of a birth cohort born to unmarried mothers and childbearing by age eighteen (it still has a statistically significant impact on childbearing by age twenty in levels, but not in logs). Nevertheless, this may indicate that mother’s marital status may not be as good an indicator of socioeconomic disadvantage as mother’s age or educational attainment. One might even expect that it would become an even poorer measure of disadvantage at birth as we go forward since the last birth cohort we are analyzing here is from 1986. As nonmarital childbearing continues its steady growth, it is reasonable to assert that it has become less selective on economic disadvantage.

The remaining panel of tables 6.3 and 6.4 includes a large array of variables controlling for differences in population characteristics (age/race/educational attainment/marital status) in each state/year at the time each birth cohort reached age seventeen or nineteen as well as differences in relevant policies (abortion, welfare, Medicaid) in place and labor market conditions at those times. As we discussed earlier, including birth cohort and state fixed effects helps control for important elements of heterogeneity that may introduce bias, but they do not solve the problem. In particular, omitted variables that reflect differences across cohorts in different locations may still result in bias. Although we recognize the possibility of unobservable factors that have this feature, we incorporate these additional variables here as an attempt to reduce the problem.

The results reported in the bottom panel of the tables are very similar to those reported in the middle panel, from models that include no additional covariates besides the state and birth year fixed effects. Among the additional variables included, the unemployment rate at about the time that early childbearing would take place is estimated to be positively related to early childbearing. Similarly, when the population of women fifteen to forty-four in a state/year is comprised of more high school dropouts, the rate of early childbearing at that time in that location tends to be higher. Other than that, all other factors tend to be statistically insignificant. The fact that adding these observable factors that vary by birth cohort/state had so little impact on the disadvantage coefficient estimates may provide a glimmer of hope
that much of the unobservable heterogeneity has been eliminated. Of course, it would be imprudent to rely on this proposition too heavily.

6.6 Conclusions

This chapter has addressed the relationship between socioeconomic disadvantage and early childbearing. After presenting a review of relevant theoretical and empirical literatures from economics as well as other disciplines, we provided a descriptive analysis from the PSID of the relationship between socioeconomic disadvantage and early childbearing at the individual level. Confirming what many previous studies have shown, we find that growing up disadvantaged is associated with substantially higher rates of teen childbearing.

The main empirical contribution of this chapter is a cohort-based analysis of the relationship between rates of socioeconomic disadvantage among women at birth and their subsequent rates of early childbearing. This analysis is conducted at the level of a state and year female birth cohort. We initially use these cohort-based data to estimate an intergenerational correlation in early childbearing, relating the percentage of a birth cohort that gives birth at a young age to the percentage of the cohort born to young mothers. Unlike the intergenerational transmission of early childbearing propensities between mothers and daughters, the intergenerational cohort-level correlation includes the impact of peer and spillover effects generated by a shared culture or environment. The results of our analyses suggest that the correlation of early childbearing across generations is much stronger in the aggregate than at the individual level. This suggests that community characteristics and the culture of teen childbearing may be more important than whether or not one’s own mother gave birth at an early age.

We obtain similar results when we take advantage of the panel nature of the cohort-based data and estimate the relationship between early childbearing and measures of disadvantage at birth. With these data, we can econometrically capture cultural/environmental differences that are long-standing in nature across states as well as geographically uniform changes that take place over time. When these elements are accounted for with state and year fixed effects, the estimated relationship between disadvantage at birth and subsequent early childbearing is greatly attenuated. This suggests that the observed relationship is almost entirely driven by broader changes in social conditions. For each of our four measures of socioeconomic disadvantage, our estimates imply that a 10 percent reduction in the proportion of a cohort with that particular proxy characteristic would lead to a decline of less than about 2.5 percent in the proportion who give birth by age eighteen or age twenty. Our results lead us to conclude that the impact of a fairly large shock to socioeconomic disadvantage would have only a modest impact
on rates of early childbearing. Other broader societal forces seem to play a larger role in determining early childbearing rates.

References


———. 2001. Risky behavior among youths: Some issues from behavioral econom-


