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ABSTRACT

Although the gender wage gap in the U.S. has narrowed, women's career trajectories diverge from men's after the birth of children, suggesting a potential role for family-friendly policies. We provide new evidence on employer provision of these policies. Using the American Time Use Survey, we find that women are less likely than men to have access to any employer-provided paid leave and this differential is entirely explained by part-time status. Using the NLSY97, we find that young women are more likely to have access to specifically designated paid parental leave, even in part-time jobs. Both datasets show insignificant gender differentials in access to employer-subsidized child care and access to scheduling flexibility. We conclude with a discussion of policy implications

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Ann P. Bartel Graduate School of Business Columbia University 3022 Broadway, 623 Uris Hall New York, NY 10027 and NBER apb2@columbia.edu Fifty-five years after the passage of the Equal Pay Act, gender differences in the labor market remain. The gender gap in pay has narrowed but not closed, and female participation rates seem to have plateaued. Indeed, the U.S. now lags behind many other advanced industrial countries on measures of gender equality in the workplace. In particular, substantial gender gaps in labor force participation and wages exist, especially for mothers. The larger gender gaps faced by mothers suggest that family-friendly employer policies, such as paid leave, childcare, and work scheduling flexibility, could play a potentially consequential role.

We fill an important gap in the literature by considering whether there are gender differentials in access to employer-provided paid leave, childcare, and work scheduling flexibility. We focus on employer-provided benefits for two main reasons: first, in the U.S., employers are the major source of these types of benefits; and second, relatively little information exists in the literature on gender differences in access to such benefits. We also discuss current public policy provisions and the role that new or expanded public policies might play.

Using two nationally representative datasets, we find that women are less likely than men to have access to employer-provided paid leave and this differential is entirely explained by the fact that women are more likely to work part-time. Young women are found to be more likely to have access to specifically designated paid parental leave even in part-time jobs. Women and men are equally unlikely to have access to childcare and scheduling flexibility. We find substantial heterogeneity by educational level: while women at all levels of education are more likely to have access to paid parental leave than similar men, highly educated women are less likely to have access to employer-provided childcare or scheduling flexibility than comparable

men. These results have important implications for policymakers, which we discuss in the concluding section.

Recent trends and developments

Equal employment opportunity policies

U.S. equal employment opportunity (EEO) policies were established with the Equal Pay Act of 1963 and Title VII of the 1964 Civil Rights Act ("EEOC Notice 915.002" 1997). The Equal Pay Act prohibits unequal pay on the basis of sex for equal work. Title VII makes it illegal for employers to discriminate against employees or job applicants on the basis of gender, sexual orientation, race, religion, age, nationality, disability, and other characteristics.

Since the 1960s, when EEO became law, the participation rate of women in the labor market increased steadily, until peaking in 2000 at 60.7 percent (Goldin 2014; Black, Schanzenbach, and Breitwieser 2017). Labor force participation particularly increased among women under age 35 (Goldin and Mitchell 2017). This may be partly attributable to the affirmative action policies that came about after EEO: reviewing the literature, Harry Holzer and David Neumark (2000; 2006) present descriptive evidence suggesting that firms using affirmative action practices have more women apply and be hired for open positions.

In addition, the gender wage gap steadily decreased in the latter part of the twentieth century. Francine Blau and Lawrence Kahn (2017) show that, while female wages have dramatically increased since the 1960s, they have not reached parity with men's, as progress largely stagnated since 2000. Blau and Kahn's (2017) decomposition analysis finds that both individual- and firm-level characteristics account for gender wage differences. In 2010, occupation and industry accounted for the majority of the explained gender wage gap; the worker's race, experience, and region also contributed. The authors conclude that it is plausible

that EEO policies were responsible for the decline in the gap, but the evidence does not definitively indicate a causal effect.

One important factor in the narrowing of the gender wage gap is the new lifecycle of women's employment. The average age at first birth has increased with recent cohorts, resulting in decreases in labor force participation later in the lifecycle as women leave the labor force to care for children, a trend that Claudia Goldin and Joshua Mitchell (2017) call the "sagging middle." This mid-life retreat from work has two important implications. First, as the age of first birth has shifted later, employment has increased for younger women (Goldin and Mitchell 2017). As is discussed below, younger women and men experience greater parity in the labor market prior to childbirth, indicating that EEO policies have the desired effect before women have children but are potentially less effective once workers become parents.¹ Second, because women are giving birth later, they have greater attachment to the labor force, take less leave time after birth, and reenter the workforce faster (Dey 2014; Goldin and Mitchell 2017). And yet, though women in their twenties and thirties are out of the labor force for fewer years and work more steadily throughout their lives (Blau and Winkler 2017; Goldin and Mitchell 2017), they do not experience parity in labor force participation or wages later in life (Black, Schanzenbach, and Breitwieser 2017; Chung et al. 2017).

¹ Marianne Bertrand, Claudia Goldin, and Lawrence Katz (2010) find that recent male and female MBA graduates from the University of Chicago begin their careers with almost identical earnings. Five years after graduation, men earn 30 log points more than women; 10-16 years after graduation, this gap increases to nearly 60 log points. One of the primary reasons for this divergence is that women experience more career interruptions. Goldin and Katz (2016) show that this pattern is not inevitable. They explore the narrowing gender gap among pharmacists, finding that technological and retail changes have increased substitutability among pharmacists, which, in turn, reduced the wage penalty for part-time work and the gender wage gap.

This new lifecycle of women's employment and the narrowing of the gap in labor force participation are linked to a closing of the education gap (Dey 2014; Goldin 2014; Blau and Kahn 2017). The average woman today has a higher level of education than the average man, and a greater share of women hold advanced degrees (Blau and Kahn 2017). This increase in women's attainment has ushered in greater labor force participation and experience (Goldin 2014).

There is substantial heterogeneity in labor force participation rates among subgroups of women, particularly for mothers. For men and women without children, labor force participation has nearly reached parity; but for mothers and fathers, a wide gap still exists (Weeden, Cha, and Bucca 2016). This gap among parents is exacerbated by gender differences in labor force exits after the birth of a child. In the year following childbirth, women's likelihood of employment decreases by 30 to 40 percent (Kuziemko et al. 2018). Women also stay out of the labor force for longer than men; this differential is narrowing, due to women taking shorter leaves (Dey 2014).

Progress on narrowing the gender gap in pay has also been uneven, particularly for workers with children. The motherhood wage penalty, whereby having children is linked to lower wages for women, is well documented in survey data (Waldfogel 1997; Budig and England 2001; Pal and Waldfogel, 2016; Blau and Kahn 2017) and in experiments (Correll, Benard, and Paik 2007). Recent estimates indicate that, on average, childless women earn 87 percent of the wages of similar childless men, while mothers earn 75 percent of the wages of similar fathers (Chung et al. 2017). The motherhood wage penalty may explain wage divergence across the lifecycle: while men and women begin their careers with roughly equivalent wages, the gender wage gap widens as cohorts age (Goldin 2014; Chung et al. 2017).

The motherhood wage penalty varies greatly across subgroups of women. Mothers in low-skill and low-income jobs bear the bulk of the wage penalty (Blau and Winkler 2017; Budig and Hodges 2014; Budig 2014). Non-married mothers and African American mothers also face a greater wage gap than do married, white, or Hispanic mothers (Pal and Waldfogel 2016).

It is difficult to determine whether the relationship between motherhood and lower wages is causal, or whether it results from the selection of women with lower wages into motherhood. The timing of the emergence of a prominent gender wage gap—after age thirty-five, when many workers are married and begin to have children—indicates a plausible causal story (Budig 2014). Blau and Kahn (2017) posit several reasons to expect the motherhood wage penalty to be causal. First, without paid parental leave, women may be more likely to leave their employers upon the birth of a child; they may exit the labor force or join an employer with more family-friendly policies. Second, if firms expect women to leave upon childbirth, they may refrain from hiring or investing in women. Finally, having children may change worker behavior in ways difficult to measure, such as decreased productivity or work scheduling restrictions. It is likely, therefore, that the motherhood penalty occurs due to issues that arise after childbirth (Goldin 2014; Blau and Kahn 2017), indicating that EEO policies are not sufficient to close the gap.

In summary, while EEO policies have likely played a role in narrowing gender gaps, progress in recent years has stalled. Though men and women begin their careers on similar ground, their trajectories diverge after having children. This may be due in large part to the fact that the time demands associated with having children are largely borne by women, which can decrease labor force participation and wages (Blau and Kahn 2017).

Family-friendly policies

As compared to EEO policies, where the U.S. has been a leader, in the domain of familyfriendly policies the U.S. lags behind its peers. While EEO policies have likely helped narrow overall gender wage and employment gaps, these policies do not address gaps for women with children, which are now the major drivers of overall gender gaps. Family-friendly policies which would help women maintain labor force attachment after having children—are needed to continue the foundation set by EEO policies.

Evidence from peer countries indicates that family-friendly policies may help reduce gender gaps. Blau and Kahn (2013) find that while the gender gap in labor force participation in the U.S. fell from 19.4 percentage points in 1990 to 14.1 percentage points in 2010, it fell much more substantially—on average, from 26.9 percentage points to 13.0 percentage points—in twenty-one other countries over the same period. Applying average family-friendly policies in other countries to the U.S., they estimate that family-friendly policies would have increased U.S. women's labor force participation rate from 75.2 percent to 82 percent during this time period. Further, while other nations have experienced similar economic and technological challenges as the U.S., our peer countries have not experienced stalled female labor force participation (Black, Schanzenbach, and Breitwieser 2017). Taken together, this evidence suggests that strengthening family-friendly policies may help close gender gaps for mothers that EEO policies have not yet touched. Michelle Budig, Joya Misra, and Irene Boeckmann (2015), using data from twenty-two nations, find that the most effective policies to keep mothers in the labor force after childbirth are mid-length leaves and childcare, and that both policies reduce the motherhood wage penalty.²

Given the potentially important role that family-friendly policies might play in reducing the gender gaps in the U.S., we focus on three key family-friendly policies: paid family leave, childcare, and work scheduling flexibility.

Paid family leave

The U.S. is the only industrialized country that does not provide paid and job-protected leave for new parents. Qualified employees may take up to twelve weeks of job-protected leave under the Family and Medical Leave Act (FMLA), though only approximately 60 percent of workers are eligible (Klerman, Daley, and Pozniak 2012). Paid family leave (PFL) has garnered increased attention in recent years, as California, New Jersey, Rhode Island, and, most recently, New York, have implemented these policies; Washington, D.C., Massachusetts, and Washington state will be implementing these policies in the coming years. These policies augment longstanding Temporary Disability Insurance (TDI) programs through which mothers in 5 states (California, Hawaii, New Jersey, New York, Rhode Island) can take some paid leave during pregnancy and post-childbirth. The effects of PFL are theoretically ambiguous, as they depend on how mothers and employers react to leave availability and leave-taking (Rossin-Slater 2017).

California's PFL is the most extensively studied U.S. program, as it dates back the furthest, to 2004. Research generally finds positive outcomes for children and parents (Milkman and Appelbaum 2013; Bartel et al. 2014; Stearns 2015; Bedard and Rossin-Slater 2016; Boushey

² It is important to note that such policies may not fully eliminate gaps, if gendered roles and attitudes persist (Kleven, Landais, and Søgaard 2018).

2016; Rossin-Slater 2017). Specifically, with regard to labor market outcomes, the evidence from California suggests that PFL increases the likelihood that mothers return to work in the nine to twelve months following a birth by 18 percentage points, increases weeks and hours worked by mothers when children are one and two years old (Baum and Ruhm 2016)[,] and increases weekly work hours for mothers of children under the age of three by 10 to 17 percent (Rossin-Slater, Ruhm, and Waldfogel 2013). One potential mechanism is that PFL may keep women in the workforce who otherwise would have dropped out after having a child (Rossin-Slater 2017). PFL is therefore a promising strategy to boost the U.S.'s sagging labor force participation for mid-career women, potentially decreasing the gender gaps in labor force participation and wages.

Childcare

Childcare is a critical component of family-friendly policy, especially for mothers of young children. As compared to parental leave, childcare may be particularly relevant in reducing gender gaps because it allows mothers to continue in employment without taking time off and/or reducing work hours, and because it affects a longer portion of their working life (Olivetti and Petrongolo 2017).. However, childcare is largely seen as a private responsibility in the U.S. (Craig and Mullan 2010; Chaudry et al. 2017). While subsidies are available for low-income families, only about 15 percent of eligible families receive such assistance (Chaudry et al. 2017). Tax credits are available to low- and middle-income families but are not widely used due to cumbersome program rules (Chaudry et al. 2017). Publicly funded childcare centers, preschools, and pre-kindergartens serve a small share of preschool age children.

Access to high-quality and affordable childcare has the potential to improve a variety of parent and child outcomes, including parental attachment to the labor market and wages (Usdansky and Wolf 2008; Ha and Miller 2015; Chaudry et al. 2017). International evidence

indicates that greater spending on childcare is associated with increased female employment and decreased gender wage gaps (Olivetti and Petrongolo 2017). Recent evidence from Washington, D.C. shows that providing free preschool for three- and four-year-olds increases maternal labor force participation by 10 percentage points (Malik 2018). Affordability is particularly crucial. Considerable evidence indicates that childcare costs affect women's wages and labor force participation (Herbst 2010; Ahn 2012; Ha and Miller 2015), especially for low-income families. In 2011, families who used childcare spent an average of 7 percent of income on care; low-income families spent 30 percent or more (Laughlin 2013). Yet current U.S. spending on childcare is low. The U.S. spends approximately 2 percent of GDP on childcare, and less than 0.2 percent of GDP on programs for young children (Chaudry et al. 2017).

Because public provision is limited, employer-provided childcare could potentially be a promising solution. While employer-supported childcare would have direct costs, it could also promote worker productivity and commitment (Hipp, Morrissey, and Warner 2017) and reduce absences and interruptions (Usdansky and Wolf 2008). Yet most American employers do not provide any support for childcare: the National Compensation Survey indicates that only 11 percent of workers have access to employer-provided childcare, a proportion that has held roughly constant for the past two decades (U.S. Department of Labor 2017). Little is known about whether women are less likely to have access to this benefit than their male peers.

Work scheduling

Finally, policies that affect work hours and schedules, such as the right to request part time or flexible hours and advance work schedules, can help women stay in the labor force after having children by providing an avenue by which to reconcile responsibilities to an employer and the family. These policies may be beneficial to women across the income distribution. For salaried workers, scheduling flexibility policies are associated both with an increase in female labor force participation after childbirth (Herr and Wolfram 2012) and an increase in female wages (Weeden 2005). These policies may be particularly beneficial for women in low-wage jobs, as they tend to have the least flexibility and may be penalized by being assigned fewer hours if they request it (Goldin 2014; Boushey 2016).

Although there is less research to date on these policies than other types of familyfriendly policies, recent studies have brought work scheduling practices to the forefront of public attention and have identified three primary types of unpredictable scheduling: lack of advance notice, last-minute scheduling changes, and changes in weekly work schedules (Henly and Lambert 2014). Recent results from the first randomized controlled trial of an intervention targeting unpredictable scheduling are promising (Williams et al. 2018). The intervention, conducted at Gap, Inc. stores in San Francisco and Chicago from November 2015 to August 2016, addressed unpredictable work scheduling by requiring stores to provide two-weeks advance notice of schedules, end the use of "just-in-time" scheduling, allow swapping of shifts among employees through an app, improve consistency of schedules from week to week, and guarantee some workers a minimum number of hours per week, among other practices. Results show that these practices increased median sales by 7 percent, increased worker productivity by 5 percent, and were generally welcomed both by employees and managers.

National surveys suggest that 6 percent of the workforce has flexibility in location of work and 49 percent has flexibility in hours or days worked (U.S. Department of Labor 2017). But the specific types of flexibility measured vary considerably, and little evidence exists as to gender differences in access to flexible work arrangements.

Access to Family-Friendly Benefits

Data and methods

We use national survey data from two datasets to provide new evidence on access to employer-provided paid leave, childcare, and work scheduling flexibility in the U.S. and consider whether there are gender differentials in such access. The first dataset, the American Time Use Survey (ATUS) 2011 Leave Module, is a nationally representative sample of the working-age population aged fifteen to eighty-five years. Critical for our purposes, the Leave Module contains detailed measures of whether the respondent receives paid leave from his or her employer, including whether he/she is able to take paid leave for vacation, errands or personal reasons, own illness or medical care, a family member's illness or medical care, childcare, eldercare, or the birth or adoption of a child. While the ATUS provides important details about whether and how workers can use paid leave, it does not measure whether workers have access to paid parental leave specifically set aside for new mothers and fathers.

We therefore turn to the National Longitudinal Survey of Youth 1997 (NLSY97), which has measures of paid leave specifically to be used by new parents. The NLSY97 is a nationally representative sample of persons who were ages twelve to sixteen in 1997; we use data from 2011, 2013, and 2015, when the respondents are ages twenty-six to thirty-six. Thus, the NLSY97 not only provides measures not captured in the ATUS, but it also provides a large sample of working adults in prime childbearing years. We also use the NLSY97 to examine access to employer-provided or subsidized childcare; the ATUS does not measure childcare benefits.

Finally, we return to the ATUS to study access to work scheduling flexibility. The Leave Module asks respondents whether they can vary their work schedules in lieu of using leave time. Respondents indicate whether they can vary work scheduling by the day of the week, hours worked in a day, or work location. This measure captures flexibility that workers may utilize when they would otherwise have to use paid or unpaid leave, but it does not capture whether they have access to more regular flexibility. We therefore use the NLSY97 to examine whether workers report a regularly accessible flexible work schedule. Last, we use the ATUS to study working from home, which is another method for achieving flexibility. To do this, we examine the number of minutes that respondents report working while at home in a given day.

For ease of interpretation, ordinary least squares (OLS) linear probability models are estimated to predict access to family-friendly policies (results from probit models are similar). We restrict our analysis of the ATUS sample to prime-age workers ages twenty-five to fifty-four old; as noted earlier, the NLSY97 sample contains only respondents ages twenty-six to thirty-six. Our primary variable of interest is gender. We also include a set of individual and employer control variables. Individual covariates include race and ethnicity (white non-Hispanic, black non-Hispanic, Hispanic, or other non-Hispanic), age, age-squared, education (high school degree or less, some college, or college degree or more), relationship status (married, cohabiting, or single), household composition (no children, youngest child ages zero to five, or youngest child ages six to seventeen), and citizenship status (native-born citizen, foreign-born citizen, or foreign-born noncitizen). Employer covariates include industry (thirteen dummies using the 2007 Census Industry Classification system), occupation (six dummies using the 2010 Census Occupation Classification system), sector (government, private non-profit, or private for-profit), and union status. We also include controls for geographic area (state fixed effects in the ATUS; region fixed effects in the NLSY97). As the NLSY97 includes observations from multiple years, we include year fixed effects in these models.

We estimate three sets of models. The first includes only controls for individual characteristics (plus the relevant geographic fixed effects and, in the NLSY97, year effects). The

second adds potentially endogenous controls for employer characteristics, and the third adds a control for part-time status, which is also potentially endogenous. As such, we estimate the following equations:

$$y_{is} = \alpha + \beta_1 female_{is} + \beta_2 X_{is} + \beta_3 S_s + \varepsilon_{is} \quad (1a)$$

$$y_{is} = \alpha + \beta_1 female_{is} + \beta_2 X_{is} + \beta_3 S_s + \beta_4 Z_{is} + \varepsilon_{is} \quad (2a)$$

$$y_{is} = \alpha + \beta_1 female_{is} + \beta_2 X_{is} + \beta_3 S_s + \beta_4 Z_{is} + \beta_5 W_{is} + \varepsilon_{is} \quad (3a)$$

$$y_{irt} = \alpha + \beta_1 female_{irt} + \beta_2 X_{irt} + \beta_3 R_r + \beta_4 T_t + \varepsilon_{irt} \quad (1b)$$

$$y_{irt} = \alpha + \beta_1 female_{irt} + \beta_2 X_{irt} + \beta_3 R_r + \beta_4 T_t + \beta_5 Z_{irt} + \varepsilon_{irt} \quad (2b)$$

$$y_{irt} = \alpha + \beta_1 female_{irt} + \beta_2 X_{irt} + \beta_3 R_r + \beta_4 T_t + \beta_5 Z_{irt} + \varepsilon_{irt} \quad (2b)$$

Equations 1a, 2a and, 3a measure gender differences in access to paid leave and scheduling flexibility, as well as minutes spent working from home in a given day, in the ATUS, and equations 1b, 2b, and 3b measure gender differences in access to paid parental leave, employer-provided or subsidized childcare, and regular scheduling flexibility in the NLSY97. In equations 1a, 2a, and 3a, *y* is access to a family-friendly policy for worker *i* living in state *s*, *female* is a dichotomous indicator set to one (zero) for female (male) workers, *X* is a vector of individual covariates, *S* is a vector of state dummy variables, *Z* is a vector of employer covariates, and *W* is a dummy variable for part-time status. In equations 1b, 2b, and 3b, *y* is access to a family-friendly policy for worker *i* living in region *r* in year *t*, *female* is a dichotomous indicator set to one (zero) for female (male) workers, *R* and *T* are vectors of region and year dummy variables *Z* is a vector of employer covariates, and *W* is a dummy variable for part-time status.

Primary results

Table 1 presents unadjusted means for access to family-friendly policies by gender in the ATUS and NLSY97. While the majority of respondents of both genders report access to some type of paid leave, a significantly smaller share of female respondents (4.3 percentage points fewer than men) have paid leave (panel A). In the ATUS, there are no significant gender differentials in access to paid leave for infant bonding but in the NLSY97 (panel B), we find that significantly more women have access to paid parental leave. Turning to employer-provided/subsidized childcare, results from the NLSY97 indicate that fewer than one in ten workers receive this benefit with no gender difference in access (panel B). Finally, ATUS results in panel A shows no significant gender difference in daily minutes spent working from home. However, results for the NLSY97 measure of regular access to scheduling flexibility (panel B) indicate a significantly greater share of women have access to this benefit.

[Insert table 1]

The gender differences observed in Table 1 could be explained by a sorting model in which workers sort into firms based on characteristics such as wages, family-friendly policies, and the availability of part-time hours. Male and female workers may value these characteristics differently, both because of personal preferences and because of the constraints they face, such as requiring time for caregiving. Women with children, or women who anticipate having children, may be more likely to sort into firms that accommodate their preferences and constraints, even if doing so requires a trade-off of pay and other benefits. For example, a mother with young children may prefer a job in which she can work part-time, though this may mean she does not have access to paid time off.

Tables 2 and 3 provide evidence consistent with this type of sorting. Table 2 shows that women, especially those with children, are significantly more likely to work part time than men. Table 3 considers whether part-time jobs provide fewer family-friendly benefits than full-time jobs. With the exception of scheduling flexibility for the NLSY97 sample, all the part-time/full-time differentials in family friendly policies are significant. Taken together, Tables 2 and 3 demonstrate that women are more likely to engage in part-time work and, in so doing, they forego access to family friendly benefits. With regard to wages, we find no significant hourly wage differential for part-time vs. full-time workers in the NLSY97; as expected, there is a significant difference in annual hours worked.

[Insert tables 2 and 3]

The results in tables 2 and 3 suggest that part-time status may explain why table 1 shows gender differentials in access to some family-friendly benefits. To more fully explore whether part-time status alone, or other control variables, explain the gender differentials, we next turn to the OLS regression results. Table 4 reports results using ATUS data, with panel A showing results for access to any paid leave by prime-age workers, panel B showing access to scheduling flexibility in lieu of leave, and panel C showing time spent working from home on a given day³. When specified only with individual controls and state fixed effects (column 1), and when adding employer controls (column 2), women are around 5 percentage points less likely to have access to any paid leave than men. Column 3 adds a control for part-time status, which essentially brings the gender differential to zero. This result indicates that, while women have

³ We also estimated all ATUS models with a sample mirroring the ages of the NLSY97 sample. Results for this younger sample (not shown but available on request) are similar in magnitude, direction, and significance.

less access to paid leave, this differential is driven by part-time workers, who are 37.7 percentage points less likely to have access to paid leave than full-time workers (see appendix table A1).

Table 5 uses the ATUS to study gender differentials by allowed type of paid leave use.⁴ Specifications 1 and 2 (panels A and B) indicate that women are significantly less likely (5 to 7 percentage points) to have access to leave that can be used for vacation, personal time, their own illness, or someone else's illness. Once we account for part-time status (panel C), these differences are no longer significant. Nor do we see significant gender differences in being able to use paid leave for infant bonding. To explicitly examine paid parental leave more closely, we turn to the primary regression results for the NLSY97, as shown in panel A of table 6 (full results shown in appendix table A2). Across all three specifications, these results indicate that, among workers of prime childbearing age, women are up to 7.5 percentage points more likely to have access to paid parental leave than similar men even when we control for part-time status.

[Insert tables 4, 5, and 6]

Using the NLSY97, access to employer-provided childcare for workers of prime childbearing age is studied in Panel B of table 6. Specification 1 shows no significant gender differential, specification 2 indicates that women are significantly less likely (1.6 percentage points) to have access to childcare than similar men, and, as in table 4, the addition of a control for part-time status in specification 3 yields an insignificant gender difference (part-time workers are 4 to 5 percentage points less likely to have access to employer-provided childcare than fulltime workers; table A3).

 $^{^4}$ Due to missing data, our sample varies according to the outcome measure used. When conducting the same analysis with a consistent sample (n=2,403), our results are similar in magnitude, direction, and significance.

Gender differentials in access to scheduling flexibility in lieu of taking leave are studied using both datasets. Using the ATUS, Panel B of table 4 indicates no significant gender differences in access to any flexibility in lieu of leave nor in access to particular types of flexibility (see appendix table A4 for full estimates of table 2, panel B; see appendix table A5 for estimates by type of leave). Using the NLSY97, estimates in panel C of table 6 examine gender differentials in regular access to scheduling flexibility (see appendix table A6 for full estimates). While specification 1 suggests that women are 3.6 percentage points more likely to have access to such flexibility than men with similar personal characteristics, specifications 2 and 3 indicate no significant gender differential. Taken together, these results show there are no gender difference in access to scheduling flexibility, whether it is in lieu of leave or more regularly available. Finally, using the ATUS in panel C of Table 4, we find no significant gender differences in minutes spent working from home on a given day, suggesting that women are not disproportionately likely to work from home instead of using other forms of scheduling flexibility.

Education-stratified results

Educational attainment influences the types of firms and jobs into which workers sort. As such, in supplemental models we examined results stratified by level of educational attainment (high school degree or less, some college but no degree, and a college degree or more). These regressions (shown in table 7) include state fixed effects, individual controls, employer controls and part-time status. In no case are women significantly less likely to have access to any leave than their male counterparts, and in the NLSY97 women in all education groups are significantly more likely to have access to paid parental leave. When considering childcare, the most highly-educated women are 5 percentage points less likely to have access to employer-provided or

subsidized childcare than comparable men. Finally, our results indicate that women with a college degree or more are less likely to have access to regular flexibility than similar men, though women with less educational attainment do not appear to face these differentials.

Policy Implications

Our analysis shows that while women have less access to any employer-provided paid leave than men, this is entirely explained by part-time status. It seems that women are sorting into part-time jobs to have more time for caregiving and in so doing are foregoing income and access to paid time off. However, we also found that young women, even those in part-time jobs, are more likely than men to have access to specifically designated paid parental leave. Furthermore, we found insignificant gender differentials in access to employer-provided or subsidized childcare, and access to scheduling flexibility. Thus, the problem with access to family-friendly policies is not gender differences but rather overall low levels of access to such policies for both women and men.

There are two main mechanisms by which public policy could expand coverage. One is to mandate that employers provide such coverage. Employer mandates are ill-suited to many types of family-friendly policies, as they can lead to discrimination against workers with children (Gruber 1994), particularly women, and they can impose high costs on employers, particularly on small businesses (Mathur et al. 2017). Employer mandates however may be the only alternative when the policies involve workplace practices such as scheduling.

The second mechanism—to provide such coverage through public provision—is more appropriate than employer mandates in situations with both high costs and the potential for discrimination. Family-friendly policies often can be funded through a small payroll tax on all workers and/or employers, thereby distributing the cost of coverage across workers rather than

burdening the specific employers whose employees take leave. In addition, universal provision mitigates the potential for worker selection into jobs with family-friendly benefits and employer discrimination against these workers, although discrimination could still occur if employers face other costs and believe that particular groups of workers will be more likely to use the benefits.

With these considerations in mind, we now turn to a discussion of whether and how family friendly policies might be expanded to address the shortfalls in coverage that we found.

Access to paid family leave

We found that young women are significantly more likely to have access to paid parental leave than comparable men. Yet overall levels of access to paid parental leave are low, with estimates indicating that slightly less than half (NLSY97) to slightly more than half (ATUS) of all workers have access to this benefit. While mandating employers to provide paid family leave receives strong public support (Horowitz et al. 2017), this may not be the appropriate mechanism by which to provide paid family leave, as it would likely be costly for employers and result in discrimination against women of childbearing age.⁵ With regard to public provision, several states have enacted paid family leave policies that are in effect (California, New Jersey, New York, and Rhode Island) or soon to be implemented (Massachusetts, Washington, and also Washington D.C.). While these policies vary widely in terms of length of leave, wage replacement rate, and eligibility requirements, they are generally funded through increased

⁵ A related option would be an employer incentive, such as the Tax Cuts and Jobs Act of 2017, which offers a 12.5 to 25 percent tax credit incentive for employers to provide paid family leave to lower-income workers, but the incentive does not offset the cost of providing this benefit and may not see a high take-up rate among employers who did not previously offer paid family leave (Mathur et al. 2017).

payroll taxes either solely on employees or on both employees and employers (National Partnership for Women & Families 2018). For example, California's program, which is funded through employee payroll taxes, costs the average worker \$2 per month in additional payroll taxes (Milkman and Appelbaum 2013).

While public provision has relatively low costs for workers, the costs to employers are potentially greater. Firms in California argued that the indirect costs of the law, such as paying current workers overtime or hiring and training temporary workers to complete the work of the employee on leave, would be detrimental, although post-law surveys found that a majority of employers indicated they were either unaffected or positively affected by the law (Milkman and Appelbaum 2013). Some employers even report being helped by the law, as the policy has replaced the costs of providing employees with paid family leave. Employer surveys in Rhode Island, New Jersey, and New York also indicate widespread support, with roughly two-thirds of employers supporting the law in their state and an additional 10 to 15 percent having neutral views (Bartel et al. 2017). On the federal level, several Congressional representatives from both parties have proposed paid family leave policies (see Mathur et al. 2017 for a discussion of these proposals).

Access to childcare

Employer mandates are probably ill-suited to expand access to childcare. Not only would the cost of such a benefit be burdensome for employers, it also may result in discrimination against workers with children. Evidence indicates that mothers who opt out of employment to care for children are seen as less committed, less capable, and less deserving of employment by firms (Weisshaar 2018) and receive fewer callbacks for interviews (Correll, Benard, and Paik

2007; Weisshaar 2018).⁶ A childcare mandate might exacerbate these gender differences if women are seen as more costly to employ.

Also, as a practical matter, the share of employers offering paid family leave is extremely low (only about one in ten). Thus, expanding public childcare programs may be the most feasible way to increase access. Though some states and cities have enacted universal pre-kindergarten programs, these programs typically provide only one (or two) years of care and currently serve less than a third of four-year-old children. Expanding childcare subsidies for low-income families, while streamlining the application and renewal process and lengthening eligibility periods, could greatly benefit households at the bottom of the income distribution. Other measures such as expanded tax credits or more public funding through sliding scale fees will be needed to reach middle-income families who also face high costs but typically receive little employer or public support.

These public policies have low potential to impose costs on employers, as they are generally funded through taxes on earners; employers even may benefit from a reduction in work absences and interruptions. Moreover, as mentioned earlier, childcare access is believed to be particularly beneficial in reducing gender gaps because it allows women to avoid work interruptions and/or reductions in work hours.

⁶ Family-friendly policies may also lead to employer discrimination: if firms believe women will require a change in hours or decrease productivity after childbirth, they may not hire or invest in women (Blau and Kahn 2013; Thomas 2018). Mallika Thomas (2018) exploits variation from the introduction of the FMLA to examine the effect of increased access to job-protected leave on firm behavior. She finds that since the enactment of FMLA, women under the age of 40 are 5 percent more likely to stay employed but 8 percent less likely to be promoted, as compared to women hired pre-FMLA.

Access to work scheduling flexibility

Employer mandates are likely the most appropriate mechanism by which to promote scheduling flexibility, as scheduling involves workplace practices that cannot be provided through public provision. Policies to promote scheduling flexibility would need to address the different issues faced by hourly workers and salaried workers. For hourly workers, scheduling flexibility means protections against unpredictable schedules. This can be accomplished by equipping the worker with input into both the number of hours per week and the time of day the hours are worked. For salaried workers, scheduling flexibility often means providing the worker with the ability to adjust when (and potentially where) their work is done, as well as the right to request part-time or flexible work. However, scheduling flexibility policies have been found to be associated with both hiring and wage penalties against women (Goldin 2014) and thus may have unintended consequences.

Several cities, including Emeryville, San Francisco, Seattle, New York, as well as the states of New York and Oregon, have enacted scheduling control policies focused on low-wage workers (Williams et al. 2018). While provisions vary, they generally mandate employers to adhere to a minimum amount of advance scheduling notice and to pay workers for any lastminute changes to schedules (Williams et al. 2018). Further, all workers (hourly and salaried) in the cities of Berkeley, San Francisco, and New York, and in the states of New Hampshire and Vermont, have the right to request flexibility in work arrangements (1 Million for Worker Flexibility n.d.). On the federal level congressional representatives from both parties have proposed policies focused both on scheduling flexibility and on scheduling control (1 Million for Worker Flexibility n.d.)

Employers may worry that an inability to change workers' schedules at the last-minute will harm their bottom lines, as many employers—particularly in retail and service—have long relied on "just-in-time" scheduling to match the number of employees to in-store traffic. Experimental evidence indicates that this concern does not bear out: employers who give workers more scheduling control see an increase in sales that is far greater than the additional cost associated with giving workers greater scheduling control, and workers who control their own schedules are significantly more productive on the job (Williams et al. 2018). Scheduling flexibility holds other potential benefits for employers. First, flexibility may boost retention, which could lower recruiting and training costs. On average, replacing a worker costs an employer about 20 percent of the worker's salary (or about 16 percent for workers earning less than \$30,000 per year) (Boushey 2016).

Conclusion

EEO policies, in combination with other factors such as changing gender norms and roles, have contributed to substantial progress toward closing gender gaps in the workplace. However, after the birth of children, women's career trajectories diverge significantly from men's, In this paper, we considered whether family-friendly policies are a potentially promising solution to promoting workplace gender equality among parents. Using data from the American Time Use Survey and the NLSY97, we provide new evidence on employer provision of these policies, finding that the gender differential in access to paid leave through employers is entirely explained by the greater likelihood of women being in part-time jobs. In order to accommodate caregiving responsibilities, women sort into part-time jobs and thereby forego income and various types of paid leave. Offsetting this is the fact that young women are more likely to have

access to paid parental leave There are no significant gender differentials in access to childcare through an employer or access to scheduling flexibility.

Although there are no gender differentials in access to family friendly policies, access to such policies remains rather low in the U.S. Public policies—such as a federal provision for paid family leave, expanded public and/or subsidized childcare, and employer mandates for scheduling control and flexibility—could play an important role in helping all families, particularly those who are low-income, navigate the tension between work and home, and have the potential to continue progress made by EEO toward a more gender equitable workforce.

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	Male	Female	Differential
Panel A: Family-friendly benefits, A	TUS		
Any paid leave	0.713	0.671	-0.043*
	(0.014)	(0.015)	(0.020)
Vacation	0.710	0.659	-0.050*
	(0.014)	(0.015)	(0.021)
Personal	0.628	0.576	-0.051*
	(0.017)	(0.018)	(0.025)
Own illness	0.694	0.651	-0.042*
	(0.015)	(0.015)	(0.021)
Other's illness	0.590	0.534	-0.039
	(0.017)	(0.017)	(0.024)
Elder care	0.353	0.341	-0.011
	(0.016)	(0.015)	(0.022)
Childcare	0.384	0.374	-0.009
	(0.016)	(0.015)	(0.022)
Infant bonding	0.523	0.540	0.017
	(0.017)	(0.016)	(0.023)
Any flexibility	0.552	0.570	0.018
	(0.016)	(0.015)	(0.022)
Hourly flexibility	0.512	0.508	0.029
	(0.015)	(0.016)	(0.021)
Day of week flexibility	0.372	0.401	-0.005
	(0.015)	(0.015)	(0.022)
Location flexibility	0.264	0.250	-0.013
	(0.014)	(0.013)	(0.019)
Daily minutes working from home	18.257	19.866	1.609
	(1.605)	(1.860)	(2.456)

 Table 1. Unadjusted gender differentials in access to family-friendly policies

 Mala
 Famala

 Differential

Panel B: Family-friendly benefits, NLSY97

Paid parental leave	0.325	0.410	0.085***
	(0.006)	(0.007)	(0.009)
Employer-provided childcare	0.081	0.087	0.005
	(0.004)	(0.004)	(0.005)
Scheduling flexibility	0.393	0.437	0.045***
	(0.006)	(0.007)	(0.009)

Source: Authors' calculation based on ATUS Leave module (2011) and NLSY97 (2011, 2013, and 2015). *Note*: ATUS sample is restricted to prime working age, twenty-five to fifty-four years old. NLSY97 sample is twenty-six to thirty-six years old. Percentages and differences are weighted using individual weights. Differentials may be slightly different than the amount obtained by subtracting the female column from the male column due to rounding. In panel A, male n=1,827 for leave, female n=1,934 for leave; male n=1,840 for flexibility, female n=1,945 for flexibility. In panel B, male n=6,781, female n=6,658. *p < .05**p < .01***p < .001

	Ma	ıle	Female			
		Regression		Regression		
	Unadjusted	adjusted	Unadjusted	adjusted		
Panel A: All workers, ATUS						
Part-time	0.143	0.138	0.286***	0.241***		
	(0.009)	(0.009)	(0.011)	(0.010)		
Observations	1,827	1,827	1,934	1,934		
Panel B: Workers with children,	ATUS					
Part-time	0.150	0.137	0.331***	0.269***		
	(0.014)	(0.013)	(0.015)	(0.014)		
Observations	1,078	1,078	1,184	1,184		
Panel C: Workers without childre	en, ATUS					
Part-time	0.137	0.137	0.252***	0.222***		
	(0.012)	(0.012)	(0.015)	(0.014)		
Observations	749	749	750	750		
Panel D: All workers, NLSY97						
Part-time	0.168	0.143	0.313***	0.262***		
	(0.004)	(0.005)	(0.005)	(0.006)		
Observations	6,781	6,781	6,658	6,658		
Panel E: Workers with children, NLSY97						
Part-time	0.118	0.107	0.342***	0.282***		
	(0.005)	(0.006)	(0.007)	(0.008)		
Observations	3,462	3,462	4,571	4,571		
Panel F: Workers without childre	en, NLSY97					
Part-time	0.213	0.173	0.263***	0.233***		
	(0.006)	(0.007)	(0.009)	(0.009)		
Observations	3,319	3,319	2,087	2,087		

Table 2. Gender differences in the share of workers who are part-time

Source: Authors' calculation based on ATUS Leave module (2011) and NLSY97 (2011, 2013, and 2015).

Note: Regression-adjusted means control for individual characteristics and firm characteristics (industry, sector, and occupation). In the ATUS, regression-adjusted means also control for state fixed effects; in the NLSY97, regression-adjusted means also control for region and year fixed effects. Individual controls include race, age, age-squared, relationship status, education, household composition, and citizenship status. Employer controls include industry, occupation, sector, and union coverage. Industry controls use the 2007 Census Industry Classification system

to categorize workers into: agriculture, forestry, fishing; mining, quarrying, and oil; construction; manufacturing; wholesale and retail; information; financial activities; professional and business services; education and health services; leisure and hospitality; other services; and public administration. Occupation controls use the 2010 Census Occupation Classification system to categorize workers into: management, professional, and related; services; sales and office; farming, fishing, and forestry; construction and maintenance; production, transportation, and material moving. Sector controls include government, private for-profit, and private non-profit. All models are weighted using individual weights. Significance stars indicate difference between male and female results. *p < .05**p < .01***p < .001

			run-ime			
	Part-Time Workers					
		Regression		Regression		
	Unadjusted	adjusted	Unadjusted	adjusted		
Panel A: Family-friendly benefi	ts, ATUS					
Any paid leave	0.223	0.346	0.700***	0.708***		
	(0.016)	(0.021)	(0.009)	(0.009)		
Any scheduling flexibility	0.647	0.650	0.514***	0.547***		
	(0.018)	(0.022)	(0.009)	(0.010)		
Minutes working from home	13.534	18.769	20.478*	18.265		
	(2.401)	(3.072)	(1.406)	(1.512)		
Panel B: Family-friendly benefi	ts, NLSY97					
Paid parental leave	0.129	0.158	0.422***	0.414***		
-	(0.007)	(0.008)	(0.005)	(0.005)		
Employer-provided childcare	0.032	0.042	0.096***	0.094***		
	(0.003)	(0.004)	(0.003)	(0.003)		
Regular scheduling flexibility	0.426	0.423	0.412	0.411		
	(0.010)	(0.011)	(0.005)	(0.005)		
Panel C: Hourly wage, NLSY97	7					
Hourly wage	23.54	20.81	26.37	21.48		
	(1.18)	(3.15)	(4.10)	(1.29)		
Panel D: Annual hours worked,	NLSY97					
Annual hours worked	1,127.017	1,293.153	2,094.322***	2,159.485***		
	(13.929)	(17.448)	(7.638)	(7.631)		

Table 3. Family-friendly benefits and hourly wages for part-time and full-time workers

Full-Time Workers

Source: Authors' calculation based on ATUS Leave module (2011) and NLSY97 (2011, 2013, and 2015).

Note: Regression-adjusted means control for individual characteristics and firm characteristics (industry, sector, and occupation). In the ATUS, regression-adjusted means also control for state fixed effects; in the NLSY97, regression-adjusted means also control for region and year fixed effects. See note to table 2 for information on individual and employer controls All models are weighted using individual weights. In panel A, paid leave n=3761, scheduling flexibility n=3,785. For all models in panels B, C, and D, n=14,060. Significance stars indicate difference between part-time and full-time results. *p < .05**p < .01***p < .001

	1	2	3
Panel A: Access to any leav	/e		
Female	-0.056**	-0.050*	-0.006
	(0.020)	(0.021)	(0.021)
Panel B: Access to scheduli	ng flexibility	in lieu of leave	
Female	0.005	-0.014	-0.025
	(0.022)	(0.023)	(0.023)
Panel C: Minutes spent wor	king from hor	ne on a given d	lay
Female	2.030	0.416	0.399
	(3.632)	(3.705)	(4.524)
Employer controls Part-time worker control	NO NO	YES NO	YES YES
	-	-	

Table 4. Primary regression results from ATUS 2011

Source: Authors' calculations using 2011 ATUS leave module.

Notes: Coefficients for individual controls, sector, and part-time are presented in table A1 for panel A and table A4 for panel B. All models include state fixed effects and individual controls. See note to table 2 for information on individual and employer controls All models are weighted using individual weights. For all models in panel A, n=3,761. For all models in panels B and C, n=3,785. * $p < .05^{**}p < .01^{***}p < .001$

	1	2	3	4	5	6	7
	Vacation	Personal	Own Illness	Other's Illness	Elder Care	Childcare	Infant Bonding
Panel A: Specif	ication 1, ir	ndividual co	ontrols only				
Female	-0.066**	-0.073**	-0.060**	-0.058*	-0.024	-0.025	-0.007
	(0.021)	(0.024)	(0.021)	(0.023)	(0.021)	(0.022)	(0.022)
Panel B: Specif	ication 2, ir	ndividual co	ontrols and e	employer co	ontrols		
Female	-0.056*	-0.062*	-0.057**	-0.055*	-0.023	-0.027	-0.019
	(0.021)	(0.024)	(0.022)	(0.024)	(0.023)	(0.023)	(0.023)
Panel C: Specif	ication 3, ir	ndividual co	ontrols, emp	loyer contr	ols, and par	t-time status	control
Female	-0.009	-0.017	-0.010	-0.007	0.002	-0.002	0.019
	(0.021)	(0.025)	(0.022)	(0.023)	(0.023)	(0.024)	(0.024)
Observations	3,673	2,909	3,538	3,117	3,505	3,526	3,521

Table 5. Access to paid leave by allowed paid leave use type from ATUS 2011

Source: Authors' calculations using 2011 ATUS leave module.

Notes: All models include state fixed effects. See note to table 2 for information on individual and employer controls. All models are weighted using individual weights. *p < .05 **p < .01 ***p < .001

1	2	3					
Panel A: Access to paid parental leave							
0.065***	0.045***	0.075***					
(0.009)	(0.010)	(0.010)					
ver-provided or	subsidized chi	ldcare					
0.0001	-0.016**	-0.010					
(0.005)	(0.006)	(0.006)					
scheduling fle	xibility						
0.036***	-0.002	-0.003					
(0.009)	(0.010)	(0.010)					
NO NO	YES NO	YES YES					
	0.065*** (0.009) ver-provided or 0.0001 (0.005) scheduling fle 0.036*** (0.009) NO	urental leave 0.065*** 0.045*** (0.009) (0.010) ver-provided or subsidized chil 0.0001 -0.016** (0.005) (0.006) scheduling flexibility 0.036*** -0.002 (0.009) (0.010) NO YES					

Table 6. Primary regression results from NLSY97

Source: Authors' calculations using 2011, 2013, and 2015 NLSY97.

Notes: All models include region fixed effects, year fixed effects, and individual controls. Coefficients for individual controls, sector, and part-time worker are presented in table A2 for panel A, table A3 for panel B, and table A6 for panel C. See note to table 2 for information on individual and employer controls. All models are weighted using individual weights. For all models, n=14,060. *p < .05**p < .01***p < .001
	1	2	3				
	High school	Some	College or				
	or less	college	more				
Panel A: Access to any lea	ve, ATUS						
Female	-0.066	0.120**	-0.038				
	(0.038)	(0.046)	(0.026)				
Observations	1,194	861	1,917				
Panel B: Access to schedul	ing flexibility in	lieu of leave,	ATUS				
Female	0.068	0.001	-0.094**				
	(0.045)	(0.050)	(0.029)				
Observations	1,194	861	1,917				
Panel C: Access to paid pa	rental leave, NL	SY97					
Female	0.064***	0.080***	0.086**				
	(0.013)	(0.017)	(0.031)				
Observations	8,383	4,406	1,271				
Panel D: Access to employ	ver-subsidized or	provided child	lcare, NLSY97				
Female	-0.006	-0.011	-0.050*				
	(0.007)	(0.010)	(0.020)				
Observations	8,383	4,406	1,271				
Panel E: Access to regular scheduling flexibility, NLSY97							
Female	0.022	-0.027	-0.073*				
	(0.014)	(0.018)	(0.030)				
Observations	8,383	4,406	1,271				

Table 7. Education stratified regression results, ATUS and NLSY97

Source: Authors' calculation based on ATUS Leave module (2011) and NLSY97 (2011, 2013, and 2015).

Notes: All models include geographic fixed effects, individual controls, employer controls, and part-time status controls. See note to table 2 for information on individual and employer controls. All models are weighted using individual weights. *p < .05 **p < .01 ***p < .001

	1	2	3
Female	-0.056**	-0.050*	-0.006
i cinate	(0.020)	(0.021)	(0.021)
Race (ref=white)			
Black	-0.011	0.002	-0.012
	(0.032)	(0.032)	(0.031)
Hispanic	-0.057	-0.035	-0.044
	(0.038)	(0.036)	(0.036)
Other	-0.005	-0.008	-0.001
	(0.052)	(0.049)	(0.045)
Age	0.005	0.005	-0.001
	(0.012)	(0.012)	(0.011)
Age-squared	-0.000	-0.000	0.000
	(0.000)	(0.000)	(0.000)
Relationship (ref=married)			
Cohabiting	-0.017	0.010	-0.008
	(0.046)	(0.045)	(0.044)
Single	-0.020	-0.014	-0.017
	(0.024)	(0.023)	(0.022)

Table A1. Access to any paid leave: Regression results, ATUS 2011

HS or less	-0.170***	-0.052	-0.040
	(0.024)	(0.028)	(0.025)
Some college	-0.089***	-0.011	-0.012
	(0.025)	(0.027)	(0.026)
Household composition (ref=n	o children)		
Youngest child ages 6-17	-0.010	-0.020	-0.005
	(0.025)	(0.024)	(0.023)
Youngest child ages 0-5	0.006	0.005	0.020
	(0.027)	(0.026)	(0.025)
Citizenship (ref=native born ci	itizen)		
Foreign born citizen	-0.051	-0.033	-0.041
	(0.045)	(0.046)	(0.045)
Foreign born noncitizen	-0.246***	-0.209***	-0.203***
	(0.042)	(0.040)	(0.040)
Union coverage		-0.040	-0.027
		(0.068)	(0.057)
Sector (ref=private for profit)			
Government		0.090*	0.084*
		(0.037)	(0.035)
Private nonprofit		-0.002	0.001
		(0.044)	(0.043)
Part time worker			-0.377***
			(0.031)

Constant	0.659**	0.632*	0.692**	
	(0.251)	(0.268)	(0.257)	
State fixed effects	YES	YES	YES	
Industry controls	NO	YES	YES	
Occupation controls	NO	YES	YES	
Observations	3,761	3,761	3,761	
R-squared	0.096	0.177	0.240	

Source: Authors' calculations using 2011 ATUS leave module.

Notes: Industry controls use the 2007 Census Industry Classification system to categorize workers into: agriculture, forestry, fishing; mining, quarrying, and oil; construction; manufacturing; wholesale and retail; information; financial activities; professional and business services; education and health services; leisure and hospitality; other services; and public administration. Occupation controls use the 2010 Census Occupation Classification system to categorize workers into: management, professional, and related; services; sales and office; farming, fishing, and forestry; construction and maintenance; production, transportation, and material moving. All models are weighted using individual weights. *p < .05 **p < .01

	Primary Job			Any Job		
	1	2	3	4	5	6
Female	0.065***	0.045***	0.075***	0.073***	0.051***	0.077***
	(0.009)	(0.010)	(0.010)	(0.009)	(0.010)	(0.010)
Race (ref=white)						
Black	0.039***	0.027*	0.028**	0.042***	0.029*	0.029*
	(0.011)	(0.011)	(0.011)	(0.011)	(0.011)	(0.011)
Hispanic	0.054***	0.027*	0.027*	0.057***	0.030*	0.030*
	(0.012)	(0.012)	(0.011)	(0.012)	(0.012)	(0.012)
Other	0.009	0.026	0.032	0.026	0.046	0.051
	(0.041)	(0.042)	(0.039)	(0.043)	(0.044)	(0.042)
Age	0.099*	0.060	0.032	0.057	0.027	0.006
	(0.050)	(0.049)	(0.048)	(0.052)	(0.051)	(0.050)
Age-squared	-0.002*	-0.001	-0.001	-0.001	-0.000	-0.000
	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)
Relationship (ref=married))					
Cohabiting	-0.061***	-0.040***	-0.031**	-0.070***	-0.051***	-0.045***
	(0.011)	(0.011)	(0.011)	(0.012)	(0.011)	(0.011)
Single	-0.068***	-0.044***	-0.042***	-0.082***	-0.060***	-0.058***
	(0.012)	(0.012)	(0.012)	(0.013)	(0.013)	(0.012)

Table A2. Access to paid parental leave, NLSY97

Education (ref=college or	more)					
HS or less	-0.147***	-0.068***	-0.060***	-0.154***	-0.075***	-0.067***
	(0.010)	(0.011)	(0.011)	(0.011)	(0.011)	(0.011)
Some college	0.031	0.007	0.009	0.027	0.004	0.003
	(0.017)	(0.017)	(0.017)	(0.017)	(0.018)	(0.018)
Household composition (re	ef=no childre	en)				
Youngest child ages 6-						
17	-0.038**	-0.029*	-0.037**	-0.047***	-0.036**	-0.042**
	(0.013)	(0.013)	(0.012)	(0.013)	(0.013)	(0.013)
Youngest child ages 0-						
5	-0.014	-0.003	-0.002	-0.021	-0.009	-0.006
	(0.011)	(0.011)	(0.010)	(0.011)	(0.011)	(0.011)
Citizenship (ref=native bo	rn citizen)					
Foreign born citizen	0.077**	0.043	0.029	0.061*	0.032	0.022
	(0.026)	(0.025)	(0.025)	(0.027)	(0.027)	(0.026)
Foreign born						
noncitizen	-0.068*	-0.060	-0.049	-0.078*	-0.067*	-0.057
	(0.032)	(0.033)	(0.034)	(0.034)	(0.034)	(0.035)
Union coverage		0.085***	0.070***		0.071***	0.061***
		(0.014)	(0.014)		(0.014)	(0.014)
Sector (ref=private for pro	fit)					
Government		-0.003	0.005		-0.003	-0.01
		(0.016)	(0.016)		(0.017)	(0.017)
Private nonprofit		0.059**	0.067**		0.076***	0.073***
		(0.022)	(0.021)		(0.019)	(0.019)
Part time worker			-0.256***			-0.218***
			(0.009)			(0.010)

Constant	0.322***	-0.507	-0.079	0.343***	-0.071	0.709
	(0.006)	(0.754)	(0.736)	(0.006)	(0.816)	(0.811)
Region fixed effects	YES	YES	YES	YES	YES	YES
Year fixed effects	YES	YES	YES	YES	YES	YES
Industry controls	NO	YES	YES	NO	YES	YES
Occupation controls	NO	NO	YES	NO	NO	YES
Observations	14,060	14,060	14,060	14,060	14,060	14,060
R-squared	0.008	0.113	0.153	0.009	0.117	0.144

Source: Authors' calculations using 2011, 2013, and 2015 NLSY97.

Notes: Region fixed effects us Census region categories: Northeast; Midwest; South; and West. Industry controls use the 2007 Census Industry Classification system to categorize workers into: agriculture, forestry, fishing; mining, quarrying, and oil; construction; manufacturing; wholesale and retail; information; financial activities; professional and business services; education and health services; leisure and hospitality; other services; and public administration. Occupation controls use the 2010 Census Occupation Classification system to categorize workers into: management, professional, and related; services; sales and office; farming, fishing, and forestry; construction and maintenance; production, transportation, and material moving. All models are weighted using individual weights. *p < .05 **p < .01 ***p < .001

	Primary Job			Any Job		
	1	2	3	4	5	6
Female	0.0001	-0.016**	-0.010	0.004	-0.014*	-0.009
	(0.005)	(0.006)	(0.006)	(0.006)	(0.006)	(0.006)
Race (ref=white)						
Black	0.034***	0.025***	0.025***	0.036***	0.025***	0.025***
	(0.007)	(0.007)	(0.007)	(0.007)	(0.008)	(0.008)
Hispanic	0.012	0.001	0.001	0.014*	0.003	0.003
	(0.007)	(0.007)	(0.007)	(0.007)	(0.008)	(0.008)
Other	-0.014	-0.015	-0.014	-0.012	-0.013	-0.012
	(0.023)	(0.023)	(0.023)	(0.025)	(0.025)	(0.025)
Age	-0.005	-0.016	-0.022	-0.016	-0.022	-0.026
	(0.030)	(0.030)	(0.030)	(0.032)	(0.032)	(0.032)
Age-squared	0.000	0.000	0.000	0.000	0.000	0.000
	(0.000)	(0.000)	(0.000)	(0.001)	(0.001)	(0.001)
Relationship (ref=married))					
Cohabiting	-0.012	-0.007	-0.005	-0.013	-0.008	-0.007
	(0.006)	(0.006)	(0.006)	(0.007)	(0.007)	(0.007)
Single	-0.019**	-0.014*	-0.013	-0.021**	-0.016*	-0.016*
	(0.007)	(0.007)	(0.007)	(0.007)	(0.007)	(0.007)

Table A3. Access to employer-provided or subsidized childcare, NLSY97

Education (ref=college or	more)						
HS or less	-0.034***	-0.008	-0.006	-0.036***	-0.007	-0.006	
	(0.006)	(0.006)	(0.006)	(0.006)	(0.007)	(0.007)	
Some college	-0.003	-0.007	-0.007	-0.004	-0.009	-0.010	
	(0.010)	(0.011)	(0.011)	(0.011)	(0.011)	(0.011)	
Household composition (r	ef=no childre	n)					
Youngest child ages 6-							
17	-0.010	-0.006	-0.008	-0.007	-0.003	-0.004	
	(0.007)	(0.007)	(0.007)	(0.008)	(0.008)	(0.008)	
Youngest child ages 0-							
5	0.003	0.007	0.007	0.003	0.009	0.009	
	(0.006)	(0.006)	(0.006)	(0.007)	(0.007)	(0.007)	
Citizenship (ref=native bo	orn citizen)						
Foreign born citizen	0.028	0.020	0.017	0.022	0.014	0.012	
	(0.017)	(0.016)	(0.016)	(0.017)	(0.017)	(0.017)	
Foreign born							
noncitizen	-0.034*	-0.035*	-0.033*	-0.039**	-0.042**	-0.040**	
	(0.013)	(0.014)	(0.014)	(0.014)	(0.015)	(0.015)	
Union coverage		0.001	-0.002		-0.003	-0.006	
		(0.008)	(0.008)		(0.008)	(0.008)	
Sector (ref=private for pro	ofit)						
Government		0.015	0.016		-0.018	-0.020	
		(0.010)	(0.010)		(0.010)	(0.010)	
Private nonprofit		0.045***	0.047***		0.028*	0.027*	
		(0.014)	(0.014)		(0.013)	(0.013)	
Part time worker			-0.052***			-0.043***	
			(0.005)			(0.006)	

Constant	0.142	0.249	0.337	0.310	0.268	0.421
	(0.464)	(0.461)	(0.460)	(0.495)	(0.518)	(0.517)
Region fixed effects	YES	YES	YES	YES	YES	YES
Year fixed effects	YES	YES	YES	YES	YES	YES
Industry controls	NO	YES	YES	NO	YES	YES
Occupation controls	NO	NO	YES	NO	NO	YES
Observations	14,060	14,060	14,060	14,060	14,060	14,060
R-squared	0.007	0.029	0.034	0.008	0.035	0.038

Source: Authors' calculations using 2011, 2013, and 2015 NLSY97.

Notes: Region fixed effects use Census region categories: Northeast; Midwest; South; and West. Industry controls use the 2007 Census Industry Classification system to categorize workers into: agriculture, forestry, fishing; mining, quarrying, and oil; construction; manufacturing; wholesale and retail; information; financial activities; professional and business services; education and health services; leisure and hospitality; other services; and public administration. Occupation controls use the 2010 Census Occupation Classification system to categorize workers into: management, professional, and related; services; sales and office; farming, fishing, and forestry; construction and maintenance; production, transportation, and material moving. All models are weighted using individual weights. $*p < .05 \ **p < .01 \ ***p < .001$

	1	2	3
Female	0.005	-0.014	-0.025
	(0.022)	(0.023)	(0.023)
Race (ref=white)			
Black	-0.031	-0.009	-0.006
	(0.035)	(0.034)	(0.034)
Hispanic	-0.037	-0.033	-0.031
	(0.037)	(0.038)	(0.038)
Other	0.074	0.084	0.082
	(0.052)	(0.051)	(0.050)
Age	0.001	-0.003	-0.001
	(0.014)	(0.013)	(0.013)
Age-squared	-0.000	0.000	0.000
	(0.000)	(0.000)	(0.000)
Relationship (ref=married)			
Cohabiting	0.004	-0.013	-0.008
	(0.049)	(0.045)	(0.045)
Single	0.021	0.014	0.014
	(0.027)	(0.026)	(0.026)

Table A4. Access to any scheduling flexibility in lieu of leave, ATUS 2011

HS or less	-0.155***	-0.112***	-0.115***
	(0.026)	(0.030)	(0.030)
Some college	-0.075**	-0.073*	-0.073*
	(0.028)	(0.029)	(0.029)
Household composition (ref=nd	o children)		
Youngest child ages 6-17	0.027	0.018	0.014
	(0.029)	(0.028)	(0.028)
Youngest child ages 0-5	0.044	0.031	0.027
	(0.030)	(0.029)	(0.029)
Citizenship (ref=native born cit	tizen)		
Foreign born citizen	-0.145**	-0.167***	-0.164***
	(0.047)	(0.047)	(0.046)
Foreign born noncitizen	-0.113**	-0.097*	-0.098*
	(0.041)	(0.042)	(0.042)
Union coverage		-0.209**	-0.211**
		(0.076)	(0.073)
Sector (ref=private for profit)			
Government		0.201***	0.199***
		(0.041)	(0.041)
Private nonprofit		0.214***	0.211***
		(0.053)	(0.053)
Part time worker			0.095**
			(0.031)

Constant	0.630*	0.444	0.430	
	(0.274)	(0.294)	(0.294)	
State fixed effects	YES	YES	YES	
Industry controls	NO	YES	YES	
Occupation controls	NO	YES	YES	
Observations	3,785	3,785	3,785	
R-squared	0.063	0.113	0.116	

Source: Authors' calculations using 2011 ATUS leave module.

Notes: Industry controls use the 2007 Census Industry Classification system to categorize workers into: agriculture, forestry, fishing; mining, quarrying, and oil; construction; manufacturing; wholesale and retail; information; financial activities; professional and business services; education and health services; leisure and hospitality; other services; and public administration. Occupation controls use the 2010 Census Occupation Classification system to categorize workers into: management, professional, and related; services; sales and office; farming, fishing, and forestry; construction and maintenance; production, transportation, and material moving. All models are weighted using individual weights. *p < .05 **p < .01

Table A5. Access to scheduling flexibility by allowed use type, ATUS 2011

123Hour of dayDay of weekLocation

Panel A: Specification 1, individual controls only

Female	-0.021	0.020	-0.027
	(0.022)	(0.022)	(0.019)

Panel B: Specification 2, individual controls and employer controls

Female	-0.032	-0.002	-0.015	
	(0.023)	(0.023)	(0.021)	

Panel C: Specification 3, individual controls, employer controls, and parttime status control

Female	-0.038	-0.002	-0.018
	(0.024)	(0.023)	(0.022)
Observations	3,779	3,776	3,775

Source: Authors' calculations using 2011 ATUS leave module.

Notes: All models include state fixed effects. See note to Table 2 for information on individual and employer controls. All models are weighted using individual weights. *p < .05 **p < .01 ***p < .001

	Primary Job			Any Job			
	1	2	3	1	2	3	
Female	0.036***	-0.002	-0.003	0.047***	0.004	0.001	
	(0.009)	(0.010)	(0.010)	(0.010)	(0.011)	(0.011)	
Race (ref=white)							
Black	-0.024*	-0.004	-0.004	-0.025*	-0.001	-0.001	
	(0.011)	(0.012)	(0.012)	(0.011)	(0.012)	(0.012)	
Hispanic	-0.022	-0.031*	-0.031*	-0.025*	-0.030*	-0.030*	
	(0.012)	(0.012)	(0.012)	(0.012)	(0.013)	(0.013)	
Other	-0.015	-0.005	-0.005	0.013	0.012	0.012	
	(0.044)	(0.044)	(0.044)	(0.046)	(0.044)	(0.044)	
Age	-0.020	-0.022	-0.021	-0.016	-0.023	-0.021	
	(0.053)	(0.051)	(0.051)	(0.054)	(0.053)	(0.053)	
Age^2	0.000	0.000	0.000	0.000	0.000	0.000	
	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	
Relationship (ref=married)							
Cohabiting	-0.040***	-0.038**	-0.038***	- 0.040***	- 0.044***	- 0.044***	
	(0.012)	(0.011)	(0.011)	(0.012)	(0.012)	(0.012)	
Single	-0.050***	-0.048***	-0.048***	- 0.049***	- 0.050***	- 0.050***	
	(0.013)	(0.013)	(0.013)	(0.013)	(0.013)	(0.013)	

Table A6. Access to regular scheduling flexibility, NLSY97

HS or less	-0.101***	-0.067***	-0.067***	- 0.113***	- 0.072***	- 0.073***
	(0.011)	(0.011)	(0.011)	(0.011)	(0.012)	(0.012)
Some college	-0.028	0.005	0.005	-0.036*	-0.007	-0.007
	(0.017)	(0.017)	(0.017)	(0.017)	(0.017)	(0.017)
Household composition (ref=	no children)					
Youngest child ages 6-17	-0.028*	-0.012	-0.011	-0.028*	-0.010	-0.010
	(0.014)	(0.014)	(0.014)	(0.014)	(0.014)	(0.014)
				-		
Youngest child ages 0-5	-0.044***	-0.029**	-0.029**	0.050***	-0.031**	-0.031**
	(0.011)	(0.011)	(0.011)	(0.012)	(0.011)	(0.011)
Citizenship (ref=native born citizen)						
Foreign born citizen	0.013	0.004	0.004	0.002	-0.002	-0.001
	(0.026)	(0.025)	(0.025)	(0.027)	(0.026)	(0.026)
Foreign born noncitizen	-0.069	-0.066	-0.067	-0.063	-0.054	-0.055
	(0.035)	(0.036)	(0.036)	(0.037)	(0.038)	(0.037)
T		-0.066***	0.0(5***		- 0.061***	-
Union coverage		-0.000****	-0.065***		0.001	0.060***
		(0.013)	(0.013)		(0.014)	(0.014)
Sector (ref=private for profit)						
Government		0.100***	0.100***		- 0.108***	- 0.108***
Government		0.100	0.100		0.100	0.100
		(0.016)	(0.016)		(0.017)	(0.017)
Private nonprofit		0.168***	0.168***		0.080***	0.080***
		(0.021)	(0.021)		(0.019)	(0.019)

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Part time worker			0.012			0.023
			(0.012)			(0.013)
Constant	0.767	0.860	0.839	0.726	0.850	0.769
	(0.816)	(0.797)	(0.797)	(0.836)	(0.872)	(0.872)
Region fixed effects	YES	YES	YES	YES	YES	YES
Year fixed effects	YES	YES	YES	YES	YES	YES
Industry controls	NO	YES	YES	NO	YES	YES
Occupation controls	NO	YES	YES	NO	YES	YES
Observations	14,060	14,060	14,060	14,060	14,060	14,060
R-squared	0.021	0.069	0.069	0.024	0.083	0.084

Source: Authors' calculations using 2011, 2013, and 2015 NLSY97.

Notes: Region fixed effects use Census region categories: Northeast; Midwest; South; and West. Industry controls use the 2007 Census Industry Classification system to categorize workers into: agriculture, forestry, fishing; mining, quarrying, and oil; construction; manufacturing; wholesale and retail; information; financial activities; professional and business services; education and health services; leisure and hospitality; other services; and public administration. Occupation controls use the 2010 Census Occupation Classification system to categorize workers into: management, professional, and related; services; sales and office; farming, fishing, and forestry; construction and maintenance; production, transportation, and material moving. All models are weighted using individual weights. $*p < .05 \ **p < .01 \ ***p < .001$