

## Telework, Wages, and Time Use in the United States

Sabrina Wulff Pabilonia, Ph.D. (Corresponding Author)  
U.S. Bureau of Labor Statistics

Victoria Vernon, Ph.D.  
SUNY Empire State College

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**Abstract:** Using data on full-time wage and salary workers from the 2017–2018 American Time Use Survey Leave and Job Flexibilities Module, we estimate hourly wage differentials for teleworkers and compare how workers allocate their time over the day when they work from home rather than the office. We find that some teleworkers earn a wage premium, but it varies by gender, parental status, and teleworking intensity. Fathers who telework earn more than fathers in office-based jobs, regardless of teleworking intensity. Women without children who telework occasionally earn more than their office counterparts. In industries and occupations where telework is more prevalent, mothers who work from home most days of the week pay a wage penalty compared to mothers in office-based jobs. Using time diaries, we find differences in work patterns and hours across worker groups that could drive these teleworker wage differentials. Most teleworkers work less on home days; however, those who earn wage premiums are working longer hours on weekdays, regardless of their work location. When teleworking, mothers experience more interruptions in their workdays than other workers, which could have negative effects on their productivity. We also find that teleworkers spend less time on commuting and grooming activities but more time on leisure activities and with family on work-at-home days than on office days, and female teleworkers spend more time sleeping and on household production activities.

**Keywords:** remote work, working from home, telework, wages, time use, commuting, productivity

**JEL codes:** J22, J31, D13

*Contact Information:*

Sabrina Wulff Pabilonia, U.S. Bureau of Labor Statistics, 2 Massachusetts Ave. NE, Rm. 2180, Washington, DC 20212 (e-mail: [Pabilonia.Sabrina@bls.gov](mailto:Pabilonia.Sabrina@bls.gov)); *ORCID*: 0000-0001-9293-2339

Victoria Vernon, SUNY Empire State College, 4 Park Avenue, New York, NY 10016 (e-mail: [Victoria.Vernon@esc.edu](mailto:Victoria.Vernon@esc.edu)); *ORCID*: 0000-0003-1245-8046

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**Ethics approval:** This study uses retrospective deidentified public-use data collected by U.S. Census Bureau. No ethical approval is required.

**Availability of data and material:** This paper uses the public-use (2017-2018) American Time Use Survey (ATUS) leave module microdata files available at <https://www.bls.gov/tus/lvdatafiles.htm> and the main ATUS files available at <https://www.bls.gov/tus/data.htm>.

**Code availability:** <https://doi.org/10.5281/zenodo.4381046>

## 1. Introduction

Over the two decades preceding the COVID-19 pandemic, the number of U.S. workers working from home rose steadily, propelled by advances in communications technology and an expansion of high-speed internet services (Pabilonia and Vernon, 2021). In 2017–2018, according to the American Time Use Survey Leave and Job Flexibilities (ATUS-LV) Module, 25 percent of wage and salary workers did some of their work at home, while 13 percent of workers worked exclusively from home at least once every two weeks (U.S. Bureau of Labor Statistics, 2019). Following the declaration by the World Health Organization (WHO) on March 11, 2020 that the novel coronavirus (COVID-19) outbreak was a pandemic, many workers were pushed into home offices, at least temporarily, in an attempt to slow the spread of the virus.<sup>1</sup> By early May 2020, 35 percent of employed persons reported that they had worked from home at some point in the past four weeks because of the pandemic and, according to the 2020 ATUS, 42 percent of employed persons did some work from home on days worked between May and December 2020 (U.S. Bureau of Labor Statistics, 2021a; 2021b).<sup>2</sup> Many experts believe that this dramatic relocation of work from office to home because of the pandemic will have a lasting impact on the location of work. Dingel and Neiman (2020) estimate that up to 37 percent of all U.S. jobs held at the beginning of 2020 could feasibly be done entirely from home, while Barrero, Bloom, and Davis (2020) predict that 22 percent of all full workdays will be supplied from home after the pandemic ends. This shift in the share of the labor force teleworking is likely

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<sup>1</sup> In the first week after physical distancing measures were implemented, Microsoft reported that their Teams app had 12 million additional users per day (Timberg, Harwell, Reiley, and Bhattacharai 2020).

<sup>2</sup> Numerous real-time surveys also document the dramatic increase in working from home in the U.S. because of the pandemic (see, for example, Adams-Prassl, Boneva, Golin, and Rauch 2020; Bartik, Cullen, Glaeser, Luca, and Stanton 2020; Bick, Blandin, and Mertens 2020; Brynjolfsson, Horton, Ozimek, Rock, Sharma, and Ye 2020). In the most recent monthly CPS (October 2021), 11.6 percent of employed persons report working at home because of the pandemic (U.S. Bureau of Labor Statistics 2021a).

to be permanent, because of better-than-expected experiences working from home (WFH) during the pandemic, investments in physical and human capital to support WFH, and diminished stigma associated with WFH. Thus, the post-COVID-19 era will likely be an era of telework, with many more workers working most of their workdays from home.

Telework (also referred to as telecommuting or remote work) is a formal or informal arrangement that allows workers to work from home or at another location other than a traditional worksite. The flexibility allowed by telework may improve worker and family well-being if the time that would have been spent commuting can be devoted to more useful or enjoyable activities, such as social interactions, household production, or child care. Parents who work from home report that their number one reason for doing so is to coordinate their work schedule with their personal or family needs (Woods, 2020). If WFH makes workers happier and allows them to better balance their work and home responsibilities, it may lead to higher productivity and higher wages. WFH can also lead to higher productivity if workers are better able to concentrate on their job tasks in a home setting, because they are not interrupted as often by coworkers or they have eliminated their taxing commutes and/or reallocated some of their time to sleeping or other relaxing leisure activities. On the other hand, employees who choose to work from home may be different in both observable and unobservable ways from those who work at a traditional workplace, and some may be willing to accept lower wages in exchange for work location flexibility.

In this paper, we use pre-COVID-19 data to answer the following two questions: (1) Do teleworkers earn higher or lower wages than office workers? and (2) Do the time-use patterns of teleworkers and office workers vary in a way that could explain observed differences in wages by teleworker status? Our paper is novel because we investigate the relationships between

telework, wages, and time use by teleworking intensity while prior researchers have focused on the relationship between remote work and wages only for home-based workers (for example, Oettinger 2011; White 2019) or the relationship between remote work and time allocation without regard to the number of days worked at home per week (for example, Giménez-Nadal, Molina and Velilla, 2019; Song and Gao, 2020). We also examine heterogeneity in the relationship between telework and wages by gender and parental status and explore one mechanism that potentially links wages to WFH, namely the allocation of time.

For our analyses, we examine these relationships using a sample of full-time wage and salary workers in white-collar occupations from the 2017–2018 ATUS-LV Module. This module allows us to determine whether a worker can work some or all their workdays exclusively from home on their main job, and how often. In our analyses, we divide regular teleworkers into two types: home-based teleworkers, who work three or more days a week exclusively from home, and occasional teleworkers, who work exclusively at home at least once every two weeks but fewer than three days a week. When we refer to office workers, we mean workers whose location of work is at a traditional worksite for their occupation and industry, which is not necessarily an office setting. By definition, office workers seldom work exclusively from home on their workday, but they may do some work at home in addition to the work they do at their worksite.

To investigate whether home-based and occasional teleworkers earn a wage premium or pay a wage penalty, we first estimate log hourly wage regressions by ordinary least squares (OLS) and then test whether our estimates are robust to bias due to unobservables using an econometric technique that relates selection on observables to selection on unobservables to place bounds on the coefficient estimates on the teleworker status variables. To examine how teleworkers choose to reallocate their time savings when WFH, we compare conditional mean

time use and the timing of daily activities on weekday workdays for teleworkers on home days versus office days, and also compare the time allocation of teleworkers and office workers. We then examine time use on all days to compare how time-use patterns vary by type of worker in order to investigate whether teleworkers differ from office workers in how much time they spend on various activities over the week.

We find that some teleworkers earn a wage premium, but it varies by gender, parental status, and teleworking intensity. Fathers who telework earn more than fathers in office-based jobs, regardless of their teleworking intensity. Women without children who telework occasionally also earn more than their office counterparts. Finally, in industries and occupations where telework is more prevalent, mothers who are home-based teleworkers pay a wage penalty compared to mothers in office-based jobs.

Using time diaries, we find differences in work patterns and hours by gender and parental status that could drive these teleworker wage differentials. Although teleworkers work less on home days, fathers work more on home and office days than men without children. On weekday workdays, women without young children who telework work even more hours than fathers who telework. When teleworking, mothers experience more interruptions in their workdays than the other worker groups, which could have negative effects on their productivity. Mothers also spend more time working in the presence of children than do fathers and less time working overall on home days.

When WFH, teleworkers gain a sizeable time windfall because of a reduction in time spent on commuting and grooming activities. They spend some of this time watching TV and using the computer for leisure. However, there are also gender differences in time allocation by WFH status. Men spend more time eating meals and socializing. Women spend more time

sleeping and on household production activities. When WFH, fathers spend more time on child care and more overall time with their children, and married men spend more time with their partners. Mothers also spend more time with their children, but not with their partners, nor do they increase their primary child care time. We also find evidence that teleworkers shift some of their activities across the days of the week, because on the average day, teleworkers and office workers spend similar amounts of time working for pay, watching TV, and sleeping.

While most white-collar workers are working between 8 a.m. and 5 p.m., teleworkers have greater flexibility in scheduling their hours on their WFH days. Fathers spend more time with their children in the hours before and after school, and women spend more time on household production and care activities during core working hours. Finally, there are some differences in sleep schedules between WFH days and office days, with teleworkers rising later in the morning on their WFH days. Overall, our findings on time use and the timing of activities suggest that teleworking enables families to better balance work and family responsibilities but may have negative effects on mothers' productivity at work.

## **2. Background**

### **2.1 Wage Effects of Telework**

There are various hypothesized ways that telework may affect wages. Teleworkers who commute less may be happier, less tired, and therefore more productive. Commuting to work is one of the least enjoyable daily activities (Kahneman et al., 2004; Kahneman and Krueger, 2006), and thus eliminating it would increase happiness. Golden, Henly, and Lambert (2013) and Kim, Henly, Golden and Lambert (2020) find a positive relationship between flexible schedule control, WFH as part of normal working hours, and worker happiness and job satisfaction. Furthermore, worker happiness is tied to productivity (Oswald, Proto, and Sgroi, 2015). Some of

the time savings from reducing commutes could be reallocated to sleeping, and more time sleeping would have a positive effect on productivity and wages (Gibson and Shrader, 2018; Groen and Pabilonia, 2019). It is also possible that worker productivity is higher while WFH if workers face fewer interruptions or distractions from co-workers (Global Workplace Analytics, 2020b).

A few randomized-controlled trials provide evidence that WFH can be more productive than working in the office. For example, Bloom, Liang, Roberts and Ying (2015) randomly assigned employees at a large Chinese travel company to work from home. They find that the home-based teleworkers are more productive, have fewer unscheduled absences, and lower quit rates than their office counterparts. In a random experiment at a large Italian company, Angelici and Profeta (2020) find that once-a-week teleworkers are more productive and have fewer absences, with stronger effects for women. In another experiment, Dutcher (2012) shows that worker productivity is higher when doing creative tasks, but not routine tasks, from home.

However, productivity also could be negatively impacted if teleworkers are more likely to experience stress or mental health problems because of their inability to separate home and work responsibilities or they experience interruptions from their children while working (Mann and Holdsworth, 2003). Using the ATUS Well-Being Module, Song and Gao (2020) find that teleworking increases fathers' stress levels and lowers mothers' happiness compared to working in a traditional workplace setting, but there are no differences in instantaneous stress or happiness levels for men and women without children. Teleworkers' work hours may also spill over into non-traditional work hours and cause conflicts with other family members. In addition, teleworkers with competing demands on their time may shirk while on the clock. During core work hours, children and chores may call for attention, leading to interruptions in workers'



workflow and potentially differentially affecting mothers' and fathers' productivity. Analyzing the wages of home-based teleworkers on the popular gender-blind online labor platform, Amazon Mechanical Turk, Adams-Prassl (2021) finds a large gender wage gap among parents that results from differences in working patterns. While not evidence of shirking because workers in this case are paid piece rate, mothers complete fewer consecutive tasks in a work session and are more likely to take longer breaks between adjacent tasks on the online platform, with negative effects on their task completion speed and potentially their productivity.

Along with reflecting differences in worker productivity, wages may reflect incentive pay and compensating differentials. From an employer's standpoint, telework arrangements are easier to implement when workers do not require costly supervision or coordination, where teamwork is less important and output is easily measured, and in jobs where workers have more autonomy. If monitoring is costly, managers may grant telecommuting rights to their most trusted and highly productive workers, who have a lower propensity to shirk; however, White's (2019) findings suggest that the costs of monitoring have fallen over the past 40 years. Alternatively, they may pay efficiency wages to elicit greater effort when monitoring is problematic.<sup>3</sup> In areas where office space is more expensive, employers have an additional incentive to encourage WFH with monetary incentives. On the other hand, employers who place a higher value on teamwork may encourage on-site presence with higher wages or promotions, leading to lower wage trajectories for teleworkers even with no differences in individual worker productivity (Rhee, 2008; Bloom et al., 2015; Glass and Noonan, 2016).

Among all flexible workplace practices, work location flexibility is one of the most highly valued by workers, and many workers report being willing to accept lower wages for the

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<sup>3</sup> This may not be as much of an issue currently, as firms are increasingly using surveillance software to ensure that their employees are working. This may also be why telework is more prevalent in computer occupations.

option to work from home (Mas and Pallais, 2017; He, Neumark and Weng, 2021; Maestas, Mullen, Powell, von Wachter, and Wenger, 2018). Workers may value WFH because WFH days could allow couples to better coordinate leisure activities and allow parents to spend more time with their children (Hamermesh, 2002; 2020). In addition, by eliminating the commute, workers save on the monetary costs of travel and prepared foods (Global Workplace Analytics, 2020a).

Policy activists often advocate for the expansion of flexible work location arrangements as a way to keep mothers attached to the labor force. If so, telework has the potential to lead to higher earnings for women as their job tenure increases, thus reducing the gender wage gap and the motherhood wage penalty. However, if women, especially mothers, view WFH as a job amenity while men see it as a demand of the job, men may select themselves into jobs that pay a premium for WFH while women may accept lower pay in exchange for work location flexibility, and therefore WFH could increase the gender wage gap (Maestas et al., 2018; Kleven, Landais, and Sogaard, 2019).

Whether teleworking leads to higher or lower wages is ultimately an empirical question. Prior studies provide mixed evidence on the wage effects of WFH in the U.S., though they vary in how they classify teleworkers, with some including those bringing supplemental work home from the office to catch-up on unfinished projects and others examining only home-based teleworkers. Most use cross-sectional data and do not account for selection effects. Using the 2001 and 2004 Current Population Survey Work Schedules and Work at Home (CPS-WS) Supplements, Weeden (2005) finds a positive relationship between flexible work arrangements and wages, with higher wage premiums in non-manual occupations. Using the 2001 CPS-WS Supplement, Gariety and Shaffer (2007) show wage premiums associated with WFH in some industries, but wage penalties in others. They attribute the negative wage differentials as being

driven by workers' preferences for WFH and the positive differentials as being driven by WFH workers being more productive, either because employers allow their most productive workers to telework or because workers are more productive while WFH. Focusing on home-based workers, Oettinger (2011) documents wage penalties in the 1980 and 1990 Censuses and a small wage premium in the 2000 Census. Between 1980 and 1990, the wage penalties fell fastest in IT-intensive occupations. More recently, using ACS and Decennial Census data and controlling for selection using a Heckman selection model, White (2019) finds that home-based workers went from paying a 26 percent wage penalty in 1980 to earning a 5 percent wage premium in 2014.

## **2.2 Telework and Time Use**

To examine how WFH is associated with time-use patterns, a few studies (Wight and Raley, 2009; Eldridge and Pabilonia, 2010; Genadek and Hill, 2017) match respondents from the 2004 CPS-WS Supplement and the 2004–2005 ATUS. Although teleworkers can be identified in the supplement, the matched sample is much smaller than the new ATUS-LV module. In addition, WFH was not as prevalent in 2004 as it was in 2017–2018.<sup>4</sup> Using the matched sample, Wight and Raley (2009) find that women who ever work from home do less market work than those who do no work from home. They also find that fathers who ever work at home spend less time on primary child care. More recently, Genadek and Hill (2017) examine differences in parents' time with children under the age of 13 by different measures of workplace flexibility and find that mothers, but not fathers, who have work location flexibility spend more total time with their children (almost 50 minutes more) than mothers who do no work from home. Neither

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<sup>4</sup> We estimate that, in 2004, 15 percent of wage and salary workers in the US reported that they did some work at home, but only 3 percent of workers worked exclusively at home at least one day every two weeks (Current Population Survey Data at NBER, 2004). For additional findings from this supplement, see U.S. Bureau of Labor Statistics (2005).

Wight and Raley (2009) nor Genadek and Hill (2017) distinguish between WFH days and days when work is brought home from the office and done in the evening; however, Eldridge and Pabilonia (2010) surmise that most of the work done at home in 2004 was work brought home from the office and done in the evening or over the weekend.

Giménez-Nadal, Molina, and Velilla (2019) and Song and Gao (2020) use ATUS data prior to the release of the ATUS-LV Module to examine the relationship between WFH and workers' subjective well-being. However, these studies could not determine whether workers were WFH occasionally or on most workdays, nor could they identify all teleworkers from the location of work on the respondents' single diary day. Giménez-Nadal, Molina, and Velilla (2019), however, find that working exclusively from home on the diary day results in a shift from market work activities to non-market work and leisure activities during core working hours.

Using the 2017–2018 ATUS-LV Module, Restrepo and Zeballos (2020) find that among all prime-age white-collar workers, those who work at home on their weekday diary day spend less time working, commuting, and on personal care, but more time on leisure, food production, and sleeping. Also using the ATUS-LV Module, Frazis (2020) examines the characteristics of all wage and salary workers who are ever paid to work exclusively at home on their workday, and studies how time is reallocated between activities when workers telework. He finds that teleworkers shift time from commuting and grooming activities to leisure and sleep and that secondary child-care time increases. While we use the same data, the focus of our study and our sample differ. We examine full-time, non-agricultural wage and salary workers in white-collar occupations and classify workers based on their frequency of teleworking. We do not require workers who telework regularly to report being paid for their work at home (although 89 percent of teleworkers state they are paid for work done from home) because all workers are

compensated for their work even if it is delayed compensation in terms of a promotion (Song, 2009).<sup>5</sup> We consider only full-time workers in order to examine the relationships between work location flexibility, time use, and wages for workers who have more similar hours of work. In addition, we examine workers separately by gender, given the sizeable differences in both time allocation and occupations held by men and women (Aguiar and Hurst, 2007; Blau and Kahn, 2017).<sup>6</sup>

### 3. Data and Descriptive Statistics

For our analyses, we use information about wage and salary workers' job flexibilities and work schedules on their main jobs collected as part of the 2017–2018 ATUS-LV Module as well as information collected as part of the main ATUS interview and time diary.<sup>7</sup> The main ATUS sample consists of a sample of people living in households that have completed their final CPS interview occurring 2–5 months earlier. Only one respondent per household is interviewed; however, the ATUS includes a household roster and demographic and labor market information for each respondent and all other household members, including age, education, employment, earnings, and usual weekly hours worked. In addition, one retrospective time diary is collected where the respondent sequentially details how she spent her time over a single 24-hour period starting at 4 a.m. on the day prior to the interview (start and stop times are reported for each activity). Activities are coded into detailed categories and, for most activities, both the location

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<sup>5</sup> For example, an educator may work certain contractual hours in a school building and may consider their Sundays spent grading each week as unpaid even if those hours are part of their usual/customary hours worked. On Sundays, they may have a choice to work in the workplace or work exclusively at home. Thus, our analysis focuses on flexibility in the location of work.

<sup>6</sup> For a more detailed review of the literature on telework and time use in the pre-COVID era, see Pabilonia and Vernon (2021).

<sup>7</sup> Code for replication can be found at: <https://doi.org/10.5281/zenodo.4381046>. The data is available at <https://www.bls.gov/tus/lvdatafiles.htm> (U.S. Bureau of Labor Statistics, 2017–2018). For additional findings from the ATUS-LV Module, see U.S. Bureau of Labor Statistics (2019).

of the activity and who else was present during the activity are also available, except for sleep and personal care activities. Only the respondents' primary activities are recorded, except for secondary child care of children under age 13. We examine major time use categories, including work and work-related activities, travel, personal care, household production, care, and leisure activities and important subcategories, such as commuting, and summary measures of time with family, friends, and coworkers or clients. Appendix Table A1 details how we group activities into categories.

Half of ATUS respondents are surveyed about a weekday and the other half about a weekend day. We use the ATUS-LV Module weights but reweight them to ensure equal-day-of-the-week representation in our male and female samples. The major advantages of the new ATUS-LV Module are that it provides information on WFH feasibility for main jobs that allows us to distinguish among home-based teleworkers, occasional teleworkers, and office workers in a nationally representative dataset and thus allows us to compare time spent on both market work and non-market work activities and the timing of those activities by WFH status. The major drawbacks of the ATUS for our purposes are that time diary data are available for only one person per household on a single day; therefore, we cannot analyze the impact of WFH on spousal time allocation beyond couple time together, nor can we compare WFH days to office days for the same teleworkers.

We initially restrict the sample to full-time, non-agricultural wage and salary workers aged 18–64 who usually work at least 35 hours per week on their main job, because we want to compare workers' time allocation on typical workdays by work location and estimate wage differentials for workers with similar usual hours.<sup>8</sup> We define a “home-based teleworker” as a

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<sup>8</sup> Landivar, Woods, and Livingston (2021) find that mothers working part-time are much less likely to have work-from-home access, or work from home, than those working full-time, and this difference is concentrated among

worker who works exclusively at home three or more days a week, and an “occasional teleworker” as a worker who works exclusively at home at least once every two weeks and at most two days a week.<sup>9</sup> An “office worker” is a worker who either never works exclusively from home or works exclusively at home less than once every two weeks.<sup>10</sup>

Figure 1 illustrates the prevalence of telework (home-based plus occasional) by detailed occupation group for men and women separately. The largest share of teleworkers by far is among computer and mathematical scientists (about 37 percent of women and 35 percent of men telework). We find significant gender differences in the proportion of teleworkers in some occupations, such as legal; community and social service; arts, design, entertainment, and sports; life, physical, and social science; and protective service occupations. Between 20 and 30 percent of men working in legal, business and financial operations, community and social service, and education, training, and library occupations are regular teleworkers. Between 20 and 30 percent of women working in business and financial operations, art, design, entertainment, and sports, and life, physical, and social sciences occupations are regular teleworkers. On the other hand, installation, maintenance, and repair, food preparation and serving related, transportation and material moving, production, and healthcare support occupations have barely any teleworkers. Figure 2 illustrates the incidence of telework by major industry. Financial activities, business and

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those working in managerial and professional occupations. In addition, they find that mothers who work part time earn 63 percent of full-time working mothers’ wages, and only part of the part-time wage gap is explained by differences in the occupational mix.

<sup>9</sup> Note that our definition of home-based teleworker corresponds closely with the home-based worker definition derived from the ACS, which asks respondents “How did this person usually get to work LAST WEEK?” If a respondent answers that they “worked at home,” then they are classified as a home-based worker. In 2019, the ACS changed the phrase “worked at home” to “worked from home” to better reflect how workers refer to this option (U.S. Census Bureau, 2017).

<sup>10</sup>To be classified as a teleworker for our study, respondents answered yes to the following two questions: “Do you ever work at home?” and “Are there days when you work only at home?” We use information from the question “How often do you work only at home?” to classify teleworkers by the intensity of regular teleworking.

professional services, and information industries have the largest shares of teleworkers, with over 20 percent of workers teleworking in each industry.

Because we want to compare wages and time use for workers in similar jobs by teleworker status, we further restrict the sample to those working in white-collar occupations, i.e. those occupations where workers could feasibly telework, for our subsequent analyses.<sup>11</sup> This sample includes those working in the following major occupation groups: management, business, and financial occupations; professional and related occupations; sales and related occupations; and office and administrative support occupations.<sup>12</sup> Altogether, 96 percent of teleworkers are in one of these occupational groupings.

Our analysis sample consists of 321 home-based teleworkers, 595 occasional teleworkers, and 4,681 office workers, or 5.7 percent, 9.8 percent, and 84.5 percent of full-time wage and salary workers in white-collar occupations, respectively.<sup>13</sup> There are no differences in the share of workers who work from home by gender, either overall or by teleworking intensity. Although 5.7 percent are classified as home-based teleworkers, many home-based teleworkers still go into the office occasionally, with 3.2 percent of full-time wage and salary workers in white-collar occupations typically working 5 or more days a week at home and 2.5 percent typically working 3–4 days a week at home.<sup>14</sup>

Teleworkers tend to have more control over their hours of work. For example, around 90 percent of teleworkers report that they also have flexible hours defined as the ability to change

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<sup>11</sup> Seventy percent of full-time wage and salary workers in the ATUS-LV Module work in white-collar occupations. We also performed our analyses using a sample of all full-time, wage and salary workers. Results are similar both quantitatively and qualitatively.

<sup>12</sup> Our sample differs slightly from Restrepo and Zeballos (2020) because we include those working in sales and related occupations given over 10 percent of teleworkers work in these occupations.

<sup>13</sup> If we were instead to consider all full-time, wage and salary workers, we would find that 4 percent were home-based teleworkers and 7 percent were occasional teleworkers.

<sup>14</sup> Workers were also asked how many days they work per week. Workers who telecommute 5+ days a week spend about 0.17 days in the office, while those who telecommute 3–4 days a week spend about 1.83 days in the office.



the time they begin and end their workday, while only 57 percent of office workers report the same.<sup>15</sup> In addition, about 95 percent of teleworkers report working daytime schedules between 6 a.m. and 6 p.m., compared to 91 percent of office workers. Teleworkers are also more likely than office workers to report working more than 5 days per week, which is consistent with teleworkers working when it is more convenient for them (16.6 percent of teleworkers versus 5.7 percent of office workers).

In Table 1, we compare mean demographic and job characteristics across our three worker types by gender. On average, teleworkers earn higher wages. Among males, unconditional mean wages are about 39 percent higher for teleworkers than office workers. Among females, home-based teleworkers earn only 15 percent more than office workers, while occasional teleworkers earn 50 percent more. All teleworkers are more likely to be partnered than office workers. Occasional teleworkers are more educated and more likely to live in a metropolitan area than office workers. They are less likely to be paid hourly, have another adult living in the household, belong to a union, and have a government sector job than office workers. Male teleworkers are older than male office workers. Male occasional teleworkers are less likely to be Hispanic and have an elderly person living in the household than male office workers. Male home-based teleworkers are less likely to be Asian and have a disability than male occasional teleworkers and male office workers. Male home-based teleworkers are more likely than male office workers to have school-age children and more likely to be born in the U.S. than male occasional teleworkers and male office workers. Female teleworkers are more likely to have an employed spouse or partner than female office workers. Female home-based teleworkers are more likely to be paid hourly than female occasional teleworkers, but less likely to have a

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<sup>15</sup> See Mas and Pallais (2020) for a review of alternative workplace arrangements and their prevalence.

graduate degree, a government sector job, and another adult living in the household. Female home-based teleworkers are less likely to have another adult living in the household and more likely to live in a metropolitan area than female office workers.

In Tables 2A–2C, we present mean demographic and job characteristics by parental status and gender for our three worker types to further describe the observable selection into teleworking, because in our multivariate analyses we find that fathers who telework earn a wage premium while men without children do not, women without children who telework occasionally earn a wage premium similar to the one for fathers who telework occasionally, and mothers pay a wage penalty for home-based teleworking in some occupations and industries. In each parental/gender status group, we again find that teleworkers earn more than office workers. And among all three worker types, fathers earn more than men without children, especially among occasional teleworkers. Unconditional mean wages are about 18 percent higher for fathers than men without children among office workers, 9 percent higher for fathers among home-based teleworkers, and 36 percent higher for fathers among occasional teleworkers. Mothers also earn more than women without children among each worker type, but the differences are less than 9 percent, and women earn less than men, except among non-parents who occasionally telework.

For each worker type, we observe some large differences in education, occupation, industry, and the likelihood of being paid hourly by parental status within each gender group, which could explain the observed parental gaps in unconditional wages and potentially the lack of teleworker wage premiums for some demographic groups. Among home-based teleworkers, men without children are much more likely to work in a health/education industry than fathers, while women without children are much less likely to have a graduate degree than mothers but are more likely to be paid hourly. Among occasional teleworkers, men without children are more

likely to be paid hourly and are less likely to have a graduate degree than fathers (though the differences are not statistically significant), and women without children are more likely to be paid by the hour than mothers. Finally, among office workers, men without children are much less likely to have a graduate degree and are more likely to be paid by the hour than fathers. Men without children are also less likely to work in management occupations and are more likely to work in office administration. Among office workers, women without children are less likely to have a graduate degree than mothers. They are also more likely to work in a leisure and hospitality industry and in public administration.

#### 4. Empirical Strategy

To estimate the magnitude and direction of the relationship between teleworking and wages by teleworking intensity, we first estimate log hourly wage regressions by OLS as follows:

$$\log W_i = \beta_0 + \beta_1 \text{Home-based teleworker}_i + \beta_2 \text{Occasional teleworker}_i + \beta_3 X_i + \varepsilon_i \quad (1)$$

where the dependent variable,  $\log W_i$ , is the natural logarithm of the hourly wage on the main job. When the hourly wage is not directly reported, we calculate it as usual weekly earnings divided by usual weekly hours.<sup>16</sup> We multiply top-coded hourly wages and earnings by 1.5, a common practice in the literature (e.g., Autor, Katz, and Kearney, 2008). *Home-based teleworker*<sub>*i*</sub> and *Occasional teleworker*<sub>*i*</sub> are binary variables for the category of teleworker;  $X_i$  includes controls for the demographic and job characteristics of individual *i*;  $\beta_0$  is a constant term;  $\beta_1$  and  $\beta_2$  are the coefficients of interest;  $\beta_3$  is a vector of coefficients; and  $\varepsilon_i$  represents the error term. Vector  $X_i$  includes a quartic polynomial in age and binary variables for highest education degree (some college, college, graduate degree), race and ethnicity (non-Hispanic black, non-Hispanic Asian,

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<sup>16</sup> Sixty-two percent of our sample do not report an hourly wage. When respondents report that usual weekly hours vary (2.2 percent), we set hours at 40.

Hispanic), lives with a spouse or partner, spouse or partner is employed, children (age 0–5, age 6–17), lives with another adult age 18–69, lives with an elderly person age 70+, foreign born, has a disability, Census region (Midwest, Northeast, West), metropolitan residence, paid hourly, union member, government sector job, survey year, 10 industry groups and 12 occupation groups.<sup>17</sup> We estimate separate regressions by gender and by gender/parental status.

All existing studies, including this one, acknowledge the difficulty of disentangling a causal relationship between wages and work location arrangements. Our OLS estimates may be biased because of unobserved worker and/or firm heterogeneity that is correlated with both wages and teleworker status. For example, individuals with better negotiation skills or advanced computer training may be both more likely to work from home and receive higher wages. Thus, the coefficients on the teleworker variables would combine the effects of WFH with the impact of these skills on wages and thus will overestimate the true impact of remote work on wages. As another example, Briscoe, Wardell, and Sawyer (2011) find a positive association between workplace size and the probability of WFH among high-skilled IT workers. Because larger firms pay higher wages than smaller firms (Bloom, Guvenen, Smith, Song, and von Wachter 2018), our OLS estimates are again likely to be biased upward.

To account for this unobserved heterogeneity, we estimate bounds on the coefficients on the teleworker variables using an econometric technique first introduced by Altonji, Elder, and Taber (2005) but recently popularized by Oster (2019). The method relates selection on observables to selection on unobservables using changes in estimated coefficients when observables are included in the model along with an assumption about the relative effect on

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<sup>17</sup> Children include own household and non-household children listed on the ATUS household roster. This can include household stepchildren.

coefficient stability of including observables versus unobservables. Specifically, Oster betas, denoted by  $\beta^*$ , are calculated as:

$$\beta^* = \tilde{\beta} - \delta[\hat{\beta} - \tilde{\beta}] \left( \frac{R_{\max} - \tilde{R}}{\tilde{R} - \hat{R}} \right) \quad (2)$$

where  $\tilde{\beta}$  and  $\tilde{R}$  are the coefficient on the telework indicator and the R-squared using the full set of controls, respectively, and  $\hat{\beta}$  and  $\hat{R}$  are the coefficient on the telework indicator and the R-squared from a regression with no controls (but including the other telework indicator), respectively. We assume that the selection bias from the observables and the selection bias from the unobservables are proportional ( $\delta = 1$ ) and have the same sign and that  $R_{\max} = 1.3 * \tilde{R}$ .<sup>18</sup> If the range of estimates bounded by the OLS estimate and the Oster beta includes zero, then the OLS estimates are not robust to correcting for omitted variable bias.

A limitation of this bounding technique is that we may not be able to learn about the relationship between unobservables, such as worker trust, and teleworker status from our observables because observable characteristics such as worker tenure are not among our controls, though we attempt to correct for this omitted variable in our main analysis by including age and managerial occupation status, assuming that more trusted employees become managers.<sup>19</sup> In addition, we estimate separate models restricting the sample to workers in the three major industry groups (financial activities; business and professional services; and information) in

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<sup>18</sup> These were estimated using the STATA command `psacalc.ado` (Oster, 2013). Oster (2019) suggests that  $R_{\max} = 1.3 * \tilde{R}$  is an adequate assumption based on a comparison of plausibly biased observational estimates with evidence on causal effects from randomized control trials. She argues that an  $R_{\max} = 1$  is too high, especially if measurement error is likely.

<sup>19</sup> In an exploratory analysis, we matched four months of diaries from the ATUS-LV Module (April–July 2018) to the January 2018 CPS Job Tenure Supplement. Job tenure was higher for men who were teleworkers but lower for women who were teleworkers, compared to their office counterparts (9.0 years versus 7.2 years for men and 7.3 years versus 8.5 years for women). However, the job tenure differences by teleworker status were not statistically significantly different, likely due to the smaller sample sizes. In addition, the male teleworkers in this smaller matched sample had noticeably higher average wages than those in the main analysis sample, so it is not surprising that they also had higher tenure. Thus, it is not apparent from this analysis that employers are granting telework based on job seniority, at least they are not doing so for women.

which over 26 percent of workers are teleworkers and then alternatively the top six occupation groups (computer and mathematical science; legal; business and financial operations; management; arts, design, entertainment, and sports; and sales and related occupations) in which over 17 percent of workers are teleworkers, because employers selectively allowing their workers to telework based upon higher trustworthiness or productivity is less likely to be an issue in occupations or industries where telework is a more common practice. Still, we interpret our estimates as conditional correlations rather than causal effects, given the multitude of selection issues surrounding the choice to work from home by the worker and the decision to allow an employee to telework by the employer and the strong assumption that the unobservables are correlated with the observables.

To examine differences in time-use patterns on home days versus office days for teleworkers and between teleworkers by work location and office workers on typical weekday workdays, we predict mean minutes spent in daily activities for respondents who worked at least four hours on their diary day.<sup>20</sup> Similar to Nätti, Tammelin, Anttila, and Ojala (2011), we control for various background characteristics that are correlated with differences in time allocation. Thus, we estimate the following linear models by OLS:<sup>21</sup>

$$Y_i = \alpha_1 \textit{Work-at-home day for teleworker}_i + \alpha_2 \textit{Work-at-office day for teleworker}_i + \alpha_3 \textit{Work-at-office day for office worker}_i + \alpha_4 Z_i + \mu_i \quad (3)$$

where the dependent variable,  $Y_i$ , represents the total daily minutes spent in an activity (work, leisure, household production, child care, etc.) or with family, friends, co-workers or clients, and

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<sup>20</sup> We explored including those who worked at least 60 minutes on their diary day; however, the higher work time restriction leads to more similar mean working times across worker types/locations without a significant drop in observation counts.

<sup>21</sup> While not all workers report doing each activity on their randomly selected diary day, they likely all do these activities regularly. In this case, estimation by OLS is appropriate.

alone; the *Work-at-home day for teleworker<sub>i</sub>* indicator equals one if the teleworker (because of the limited sample of WFH days we pool home-based and occasional teleworkers) worked at home for at least four hours and worked in the office for zero minutes, and zero otherwise (they may have also worked at another location besides their home such as a coffee shop, although time worked at other places is minimal). The *Work-at-office day for teleworker<sub>i</sub>* indicator equals one if the teleworker worked in the office for at least four hours and zero otherwise, and the *Work-at-office day for office worker<sub>i</sub>* indicator equals one if the office worker worked in the office for at least four hours and zero otherwise (they may have also taken work home); the vector  $Z_i$  includes the controls for the characteristics included in  $X_i$  with the addition of log hourly wage and month of interview indicators;  $\alpha_1$ ,  $\alpha_2$ , and  $\alpha_3$  are coefficients to be estimated;  $\alpha_4$  is a vector of coefficients to be estimated; and  $\mu_i$  represents the error term.<sup>22</sup> These models omit the constant term. In the results section, we present predictions from these models for mean time spent in activities on weekday workdays in 2017–2018 for our three groups of workers.

Finally, to examine differences in time allocation across worker types for the average day, we estimate the following linear models by OLS:

$$Y_i = \gamma_1 \text{Home-based teleworker}_i + \gamma_2 \text{Occasional teleworker}_i + \gamma_3 \text{Office worker}_i + \gamma_4 A_i + \eta_i \quad (4)$$

where  $Y_i$ , *Home-based teleworker<sub>i</sub>*, *Occasional teleworker<sub>i</sub>*, and *Office worker<sub>i</sub>* are defined as above; the vector  $A_i$  includes the controls in  $Z_i$  with additional controls for Saturday and Sunday/holiday time diaries;  $\gamma_1$ ,  $\gamma_2$ , and  $\gamma_3$  are coefficients to be estimated;  $\gamma_4$  is a vector of coefficients to be estimated; and  $\eta_i$  represents the error term.

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<sup>22</sup> We also estimate conditional means where we only control for demographic characteristics (not shown); and the results are similar, suggesting that job characteristics other than work location arrangements do not affect time allocation.

## 5. Results

### *5.1 Do Teleworkers Earn Higher or Lower Wages than Office Workers?*

Table 3 presents OLS coefficient estimates and Oster betas. In most instances, the OLS coefficients represent the upper bound. For the average worker (column 1), we find that male home-based teleworkers earn more than male office workers (an 8–15 percent wage premium), while female home-based teleworkers do not earn a wage premium.<sup>23</sup> However, for the average female worker (Panel B, column 1), we find that female occasional teleworkers earn a wage premium of 7–16 percent. While the coefficient on occasional teleworker for the average male worker is statistically significant and shows a 9 percent wage premium, the estimate is not robust to correcting for omitted variable bias, as the Oster beta is negative. In a robustness check, we pool male and female workers and include interactions between our teleworker variables and gender in equation 1. The coefficient on the gender interaction term with home-based teleworker is negative, of similar magnitude as the coefficient on home-based teleworker, and statistically significant. The coefficient on the gender interaction term with occasional teleworker, however, is not statistically significant and close to zero (see Appendix Table A.2).

Looking at the results by parental status (Table 3, columns 2 and 3), we find that fathers who are home-based teleworkers earn 11–15 percent more than fathers who are office workers, and fathers who are occasional teleworkers earn 11–19 percent more than fathers who are office workers. Because the OLS estimates are similar for fathers, we estimate a specification with a teleworker indicator and its interaction with home-based teleworker and cannot reject the hypothesis that the teleworker wage differential is the same regardless of their teleworking

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<sup>23</sup>  $(e^{\beta} - 1) \times 100$  is the percentage change in the wage associated with a unit change in the indicator variable.



intensity (results not shown). Men without children who are home-based teleworkers earn 9–16 percent more than men without children who are office workers; however, the coefficient is not statistically significant at conventional levels, which is likely because of the smaller sample size. Men without children who are occasional teleworkers do not earn a wage premium, while women without children who are occasional teleworkers earn 14–23 percent more than women without children who are office workers. Women without children who are home-based teleworkers do not earn a wage premium. Mothers who telework either occasionally or most of their days also do not earn a wage premium.

In columns 4–9 of Table 3, we present estimates where we restrict the sample to the industries, and then occupations, where teleworking is more prevalent and thus selection issues are less likely to be causative. In the top industries specifications (columns 4–6 of Table 3), we observe that for fathers the OLS coefficient estimates and Oster betas are larger than in the baseline results. In the financial activities, business and professional services, and information industries, fathers who are home-based teleworkers earn 22–23 percent more than fathers who are office workers, and fathers who are occasional teleworkers earn 23–28 percent more than fathers who are office workers. Again, we see that the OLS estimates do not vary a lot by teleworking intensity for fathers. For women without children, the OLS estimate and Oster beta are similar to those in the baseline results for those who are occasional teleworkers: they earn 15–23 percent more than their office counterparts. In the top industries, men without children who are home-based teleworkers do not earn a wage premium; however, this result should be interpreted with caution, because the number of home-based teleworkers is small. In the occupation specifications (columns 7–9), we again find that fathers earn more if they telework and women without children earn more if they occasionally telework, although with the smaller

sample size, the former is not statistically significant for home-based teleworkers. We also observe a new result in the top industries and top occupations specifications—mothers who are home-based teleworkers pay a wage penalty of 14–22 percent. This result is consistent with either a negative productivity story resulting from greater interruptions from children during the workday or mothers’ willingness to accept lower wages for the option to work from home to better balance work and home responsibilities.

## ***5.2 Time-Use Patterns***

In Tables 4A and 4B, we present conditional mean time spent on each activity, and then separately total time spent with family, friends, coworkers or clients, and alone, on weekday workdays in minutes per day.<sup>24</sup> Note that time with children is the sum of all time spent on activities during which at least one own child under age 18 was present. We also show time working with children present, secondary child care, and the number of work episodes.

Male teleworkers on WFH days gain about an hour from not having to commute (Table 4A).<sup>25</sup> They gain an additional 14 minutes by reducing their time spent on grooming activities. However, on average, teleworkers on home days work about 27 minutes less on their main jobs than on-site workers. While on office days teleworkers and office workers have similar hours of work, we see that teleworkers on their office days do relatively more of their work from home (34 minutes versus 10 minutes). Looking at daily work hours across worker type and location by parental status, we find that teleworking men without children work about 40 minutes less on home days than on-site workers, while teleworking fathers work 35 minutes less on home days than office days and 18 minutes less than fathers who are office workers (only the differences for

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<sup>24</sup> We also examine all workdays when workers work at least four hours, and results are similar (Appendix Table A.3.A and A.3.B); however, we prefer to focus on weekdays, because teleworkers work primarily on weekdays and we may pick up some work brought home from the office by including weekend days in the analyses.

<sup>25</sup> We calculate commuting time using the trip tour methodology described in Kimbrough (2019).

men without children are statistically significant). Comparing hours across parental status, we find that fathers work longer hours in each worker group, especially fathers who are teleworkers. On WFH days, fathers work 33 minutes longer than men without children. On office days, fathers who are teleworkers work 25 minutes longer than men without children who are teleworkers, while fathers who are office workers work only 11 minutes longer than men without children who are office workers. On office days, male teleworkers and office workers spend similar amounts of time with coworkers and clients, suggesting that the level of teamwork and face-to-face interaction required for workers who telework and those who do not is similar.<sup>26</sup> On WFH days, men report more work episodes, showing that they had more interruptions in their workday, with no differences by parental status. Our findings that only fathers who are teleworkers earn wage premiums and that they work longer hours than men without children who are teleworkers are consistent with prior research that finds firms reward workers who labor long hours (for example, Goldin, 2014).

In terms of work-life balance, we find that male teleworkers spend 13–20 minutes longer eating their meals on WFH days than do male on-site workers. When WFH, male teleworkers spend more time caring for family members and pets than male on-site workers (14–16 minutes more). Fathers who telework spend 25 minutes more on primary child-care activities, about 100 minutes more in the presence of their children, and over 2.5 hours more caring for their children as a secondary activity than on-site workers. Fathers who work from home sometimes have children in their presence while working (24 minutes more per day on average than office workers). This is not surprising, because children’s school hours are usually less than the hours worked each day by parents with full-time jobs. Male teleworkers also spend 30 minutes more

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<sup>26</sup> In estimations not shown, we find no differences in time with coworkers by parental status.

watching TV and using computers for leisure and 22 minutes more on social activities on WFH days than on office days. They also spend more time with their partner than do on-site workers (54–67 minutes more). Finally, male teleworkers spend more time alone on WFH days (6 hours and 11 minutes more).

Female teleworkers and office workers on their office days spend 53 minutes commuting to work (Table 4B). Going into the office also requires an additional 22 minutes of grooming time for female teleworkers. Thus, female teleworkers on home days also experience a significant time windfall by eliminating their commutes and other preparations for work. Female teleworkers also work less at their main jobs on home days than on office days (26 minutes less), but the difference is not statistically significant. However, female teleworkers on office days work 29 minutes more than office workers, because they do additional work from home (34 minutes more). Looking at daily work hours across worker type and location by parental status, we find that women without children who are teleworkers work 27 minutes less on home days than office days but work 47 minutes more on their office days than do office workers. Mothers who are teleworkers work 52 minutes less on home days than office days and 40 minutes less than mothers who are office workers.<sup>27</sup> Comparing hours across parental status, we find that women without children work substantially longer hours than mothers in each worker group, especially those who are teleworkers. On WFH days, women without children work 81 minutes longer than mothers. Among teleworkers on office days, women without children work 56 minutes longer than mothers. Among those in office-based jobs, women without children work 21 minutes longer than mothers. In addition, among teleworkers, the hours of women without children are even higher than the hours of fathers. Like male teleworkers, female teleworkers and

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<sup>27</sup> Although these differences are not statistically significant on weekday workdays, they are large and are statistically significant in the all workdays sample (see Appendix Table A.3.B)

office workers spend similar amounts of time with coworkers and clients on office days. Consistent with prior research by Adams-Prassl (2021), mothers have slightly more work episodes than fathers on WFH days, 3.5 versus 3.2 episodes, while women without children only have 2.6 work episodes on WFH days. It is possible that the greater number of interruptions in their work could lead to mothers being less productive at work, which could explain the teleworker wage penalty for mothers in some occupations and industries. The wage premium for women without children who are occasional teleworkers is consistent with their working longer hours. Working longer hours on weekday workdays with few interruptions for non-work activities may be especially valued by firms if coordination of work activities with coworkers during core business hours is important (Cubas, Juhn, and Silos, 2021).

In terms of work-life balance, we find that female teleworkers spend substantially more time engaging in home production activities on WFH days than on office days (32 minutes more). They also enjoy 33 minutes more time watching TV and using a computer for leisure and sleep 27 minutes longer. In contrast to male teleworkers, female teleworkers do not spend more time with their partners on WFH days, which may be because men's wives tend to work fewer hours than women's husbands. However, on WFH days, they spend more time with their partners than female office workers (47 minutes more). Female teleworkers also spend more time alone on WFH days than office days (5 hours and 49 minutes more).

Mothers who are teleworkers spend more total time around their children on WFH days than office days (over 2 hours more), more time on secondary child care (3.2 hours more), and more time working with their children in their presence (45 minutes more). Their primary child care time, however, does not vary by the location of their work. In fact, when WFH, fathers spend more time on primary child care than mothers (70 versus 60 minutes). Mothers, however,

have children in their presence during work episodes conducted from home to a greater extent than fathers (31 minutes more). It is possible that this additional time spent working with children present could also lead to mothers being less productive at work, which could explain the parental differences in wage premiums for female teleworkers. Thus, full-time employed mothers may select themselves into jobs that do not require long hours, thus forgoing compensation and promotion but gaining better work-life balance.

Next we compare time allocation on the average day for our three worker types to examine whether teleworkers prefer certain activities over others compared to office workers or whether they shift certain activities from office days to home days or from workdays to non-workdays to create more balance in their lives. We find a few noteworthy differences from our weekday workday results and between our two teleworker types. (See Appendix Tables A.4.A and A.4.B for the full set of results). On the average day, male teleworkers and male office workers work the same amount of time, suggesting neither overworking nor shirking by teleworkers (but some shifting of work time), which is contrary to prior researchers' findings based on non-diary survey data that suggests teleworkers work longer hours (Noonan and Glass, 2012). Only male home-based teleworkers spend less time commuting than male office workers (20 minutes less). This suggests that the commute may be slightly longer for male occasional teleworkers than male office workers, which de Vos, Meijers, and van Ham (2018) and de Vos, van Ham, and Meijers (2019) found to be true for some workers in the Netherlands.<sup>28</sup> Fathers who are home-based teleworkers spend 19–21 minutes more on primary child care and about 1.8

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<sup>28</sup> Rhee (2008) argues that when telecommuting is adopted, workers may be more likely to choose to commute to a distant workplace than to a nearby one. To further examine commuting time differences, we pool males and females due to the small sample size and similar commute times and then estimate commuting time on workdays with additional controls for office workdays and home workdays by worker type. Occasional teleworkers spend 8 minutes more time commuting to the office than office workers, while home-based teleworkers spend 12 minutes more time commuting to the office than office workers.

hours more on secondary child care than fathers who are occasional teleworkers or office-based workers. We also do not find any differences in TV and computer time on the average day across worker types, suggesting that men prefer to spend the same amount of time watching TV, playing video games, and engaging on social media but teleworkers do more of these activities on their WFH weekdays while office workers do more of these activities on non-workdays.

For women, on average, total work time on the average day does not vary by teleworker status. However, among women without children, home-based teleworkers work more than occasional teleworkers and office workers (51–57 minutes more), and among mothers, home-based teleworkers work less than occasional teleworkers and office workers (50–61 minutes less). However, comparing across parental status among home-based teleworkers, women without children work over two hours longer than mothers. Comparing across parental status among occasional teleworkers, women without children work only 10 minutes longer than mothers. We find no differences by worker type in the time women spend on household production, sleeping, and watching TV on the average day, which indicates that they are shifting these activities across the days of the week. Female home-based teleworkers, however, spend more time on sports and active leisure (10–11 minutes more).

Time-use differences between teleworkers and office workers could also be biased because of selection into telework and omitted workplace variables. To verify that our results are robust to omitted variable bias, we estimate linear regressions by OLS (varying the omitted worker group and including a constant term) and calculate Oster betas. For all our statistically significant results, the Oster bounds exclude zero, suggesting that our time-use results are robust to bias from unobservables (Appendix Tables A.5 and A.6).

### ***5.3. Timing of Activities: Teleworking 8 to 5?***

Workers may also vary the timing of their activities over the day between WFH days and office days. In Figures 3–6, we show the share of workers among teleworkers on WFH days, teleworkers on office days, and office workers on office days who are participating in select activities (work and work-related activities, household production and care, time with own children, and sleep) at each minute of the day on weekday workdays. In Figure 3, we observe that most workers in all three groups are working during traditional core working hours (8 a.m. to 5 p.m.), with a large dip in the share of all workers working at lunchtime. However, male teleworkers are slightly less likely to be working in the after-school hours (3 to 5 p.m.) on WFH days than on office days, although they are just as likely to be working on office days as office workers during those hours (Panel A). This could explain why teleworkers’ work time on weekday workdays is less than the work time of on-site workers. Male teleworkers on WFH days are more likely to be doing household production and care activities and spending time with children during these after-school hours (Figures 4 and 5). For example, consider the 4 p.m. diary time. At 4 p.m., 74 percent of male teleworkers on office days are working, while only 58 percent of male teleworkers on WFH days are working (Figure 3 Panel A). Twenty-two percent of male teleworkers on WFH days are doing household production and care activities (Figure 4 Panel A), while only 5 percent of male teleworkers on office days are doing these activities. Among fathers who are teleworkers, 28 percent are spending time with children on their WFH days (Figure 5 Panel A), while on 7 percent are spending time with children on their office days.

Even though their work time is not statistically significantly different across work locations, female teleworkers are less likely to be working during core working hours on WFH days than on office days (Figure 3 Panel B), suggesting that WFH gives many teleworkers greater flexibility to balance household and family responsibilities over the day. We also observe



that female teleworkers on WFH days spend more time around the lunch hour and during afternoon hours on household production and care activities and more time with children throughout the traditional workday than on-site workers (Figures 4 and 5, Panel B).

Looking at sleep (Figure 6), we find that a greater share of teleworkers is sleeping later in the morning on WFH days than on office days. This suggests that WFH allows workers' waking hours to shift to later in the day, i.e., they wake later and go to sleep later, which may be a sign that standard work schedules do not sync with circadian rhythms or that night owls select telework. On average, male teleworkers on WFH days wake up at 6:36 a.m. but on office days they wake up at 6:17 a.m., while female teleworkers on WFH days wake up at 6:48 a.m. but on office days they wake up at 6:05 a.m. (19 and 43 minutes earlier on office days, respectively) (Table 5). The latter difference in female wake times also corresponds with the earlier finding that female teleworkers get more sleep on WFH days. On non-workdays, all worker types wake up at similar times, suggesting that night owls do not select into telework. Thus, there may be positive productivity effects resulting from increased sleep and quality of sleep on WFH days because of differences in the timing of sleep.

## **6. Discussion and Conclusion**

We use pre-pandemic data from the ATUS-LV Module to gain insights into the relationship between telework and wages and explore a potential mechanism—time allocation—by which they may be linked. Understanding how being able to work entire workdays from home affects wages and how teleworkers allocate their time is important for post-pandemic policy design of family-friendly workplaces where telework will be ever more prevalent and children

will spend most of their parents' workdays in schools.<sup>29</sup> Because the relationships between teleworking, wages, and time use vary by gender, our study is also relevant to gender equality policy making.

We show that mean wages are higher for teleworkers than office workers; however, once we account for observable demographic and job characteristics and correct for bias in the coefficients on the teleworker variables from the unobservables using Oster's bounding technique, we find that most women do not earn a wage premium. Only women without children under age 18 who occasionally work from home still earn a wage premium, while mothers who work most days of the week at home pay a wage penalty in some occupations and industries. Thus, our results suggest that increasing the number of telework days for women will not reduce the gender wage gap or motherhood wage gap, although it may still improve mothers' well-being and possibly even allow some mothers to take part in the labor force who otherwise would not. On the other hand, increasing the frequency of telework may even increase the motherhood wage gap, because our results suggest higher wage premiums for fathers who are home-based teleworkers. We also find that fathers earn wage premiums when they occasionally work from home. Our results showing differences between teleworking mothers and fathers are consistent with prior research showing that mothers are more willing to pay for location flexibility and they may have lower productivity when working remotely because of interruptions throughout their workdays. Mothers who telework take more breaks throughout their workdays and are potentially interrupted more by their children who are more likely to be in their presence while they work from home. We also do not find that men without children earn a wage premium for

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<sup>29</sup> See Yamamura and Tsutsui (2021) for an examination of the impact of closing schools in Japan on working from home during the pandemic. See Pabilonia and Vernon (2022) for an examination of time allocation by parents in dual-earner couples with children by the couple's work location arrangements during the pandemic when many U.S. children were in virtual schooling.

teleworking once we control for observables, such as education and job characteristics, which vary by parental status. From the time diaries, we find evidence that the above-mentioned groups of teleworkers who earn wage premiums are working longer hours on weekday workdays than the groups of teleworkers who do not earn wage premiums, which is consistent with previous researchers' finding that those working long hours earn a wage premium and that those who work in occupations that require the coordination of activities during core business hours also earn a wage premium.

Differences in time allocation among teleworkers by work location and between teleworkers and office workers also suggest that WFH improves workers' work-life balance because they spend less time commuting and grooming. Workers may be more alert on their jobs when they can skip their morning commutes and other preparations for going into the office, resulting in higher productivity on their WFH days. Female workers get more sleep on WFH days, which may also boost their productivity. Our results suggesting no statistically significant difference in work time on the average day by worker type lead us to conclude that workers are not shirking on the job or being overworked as the boundaries between work and home life blur. Teleworkers also spend more time watching TV and using the computer for leisure on their WFH days than office days, though not on the average day. This suggests that teleworkers adjust the timing of some activities over the days of the week, which also could enhance their well-being.

Female teleworkers, but not male teleworkers, use some of their time windfall from the elimination of their long commutes to do more household production activities on their WFH days. However, it is possible that an expansion of telework could decrease the gender care gap, because fathers spend more time on primary child care when they work from home while

mothers do not, and fathers who are home-based teleworkers spend relatively more time on secondary child care on the average day than occasional teleworkers and office workers.

Parents spend more total time with their children and at different times of the day when they work from home instead of at the office. Thus, telework potentially has positive implications for child development—because children receive more parental time overall, more parental time in the hours after school, when they may need it most, and more primary child care time from their fathers (Fiorini and Keane, 2014; Hsin and Felfe, 2014; Caetano, Kinsler, Teng, 2019)—and positive implications for parents’ well-being, because parents enjoy spending time with their children more than doing other activities (Connelly and Kimmel, 2015; Musik, Meier, and Flood, 2016).

Our study has several limitations. Our sample of teleworkers is quite small. Therefore, we cannot investigate these relationships for more detailed occupation groups nor can we investigate time allocation differences by work location by type of telecommuter. We also have only one observation per household. Therefore, we cannot observe couples trading off tasks based on the work location of both partners. We may learn more about the tradeoffs couples make and in which occupations and under what circumstances telework is more productivity enhancing using data during the pandemic when teleworking is more prevalent.

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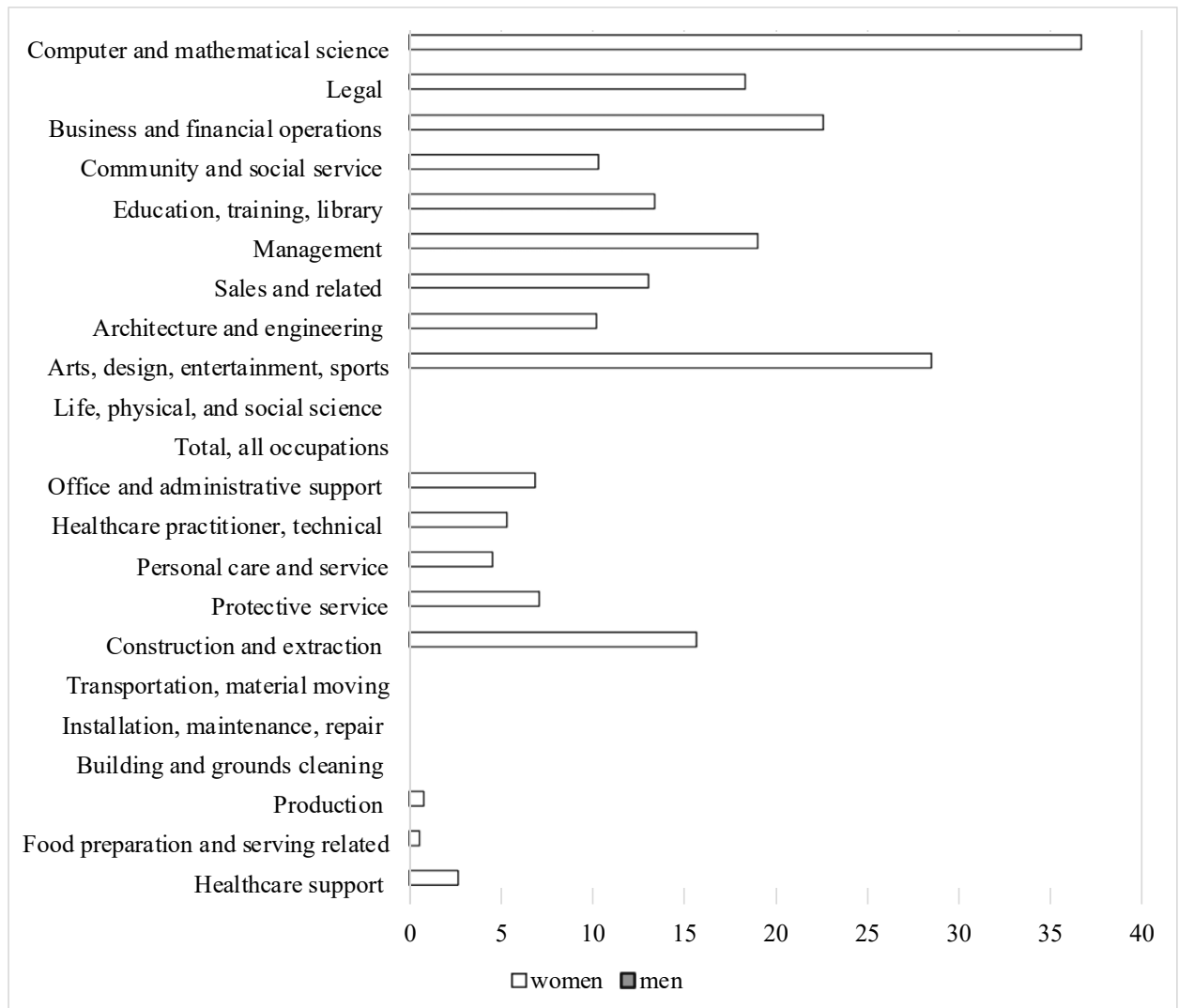
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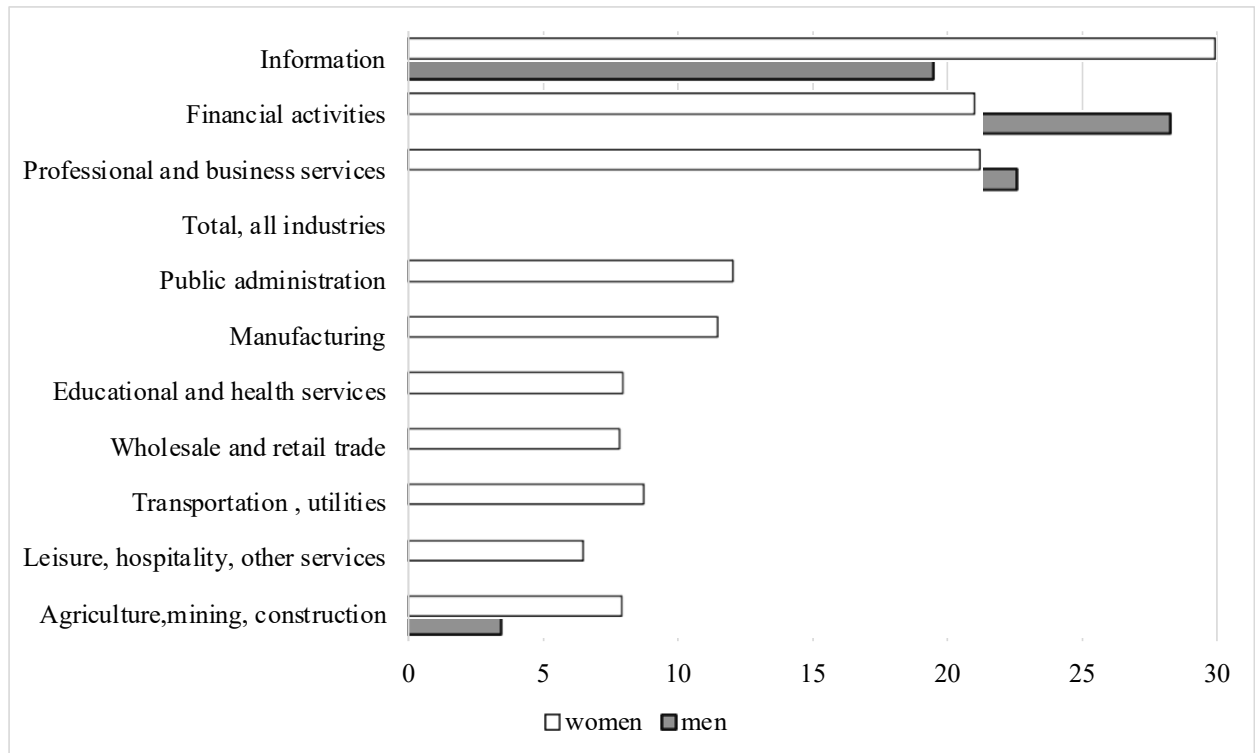


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**Fig. 1** Percent of teleworkers among full-time wage and salary workers by occupation group, 2017–2018

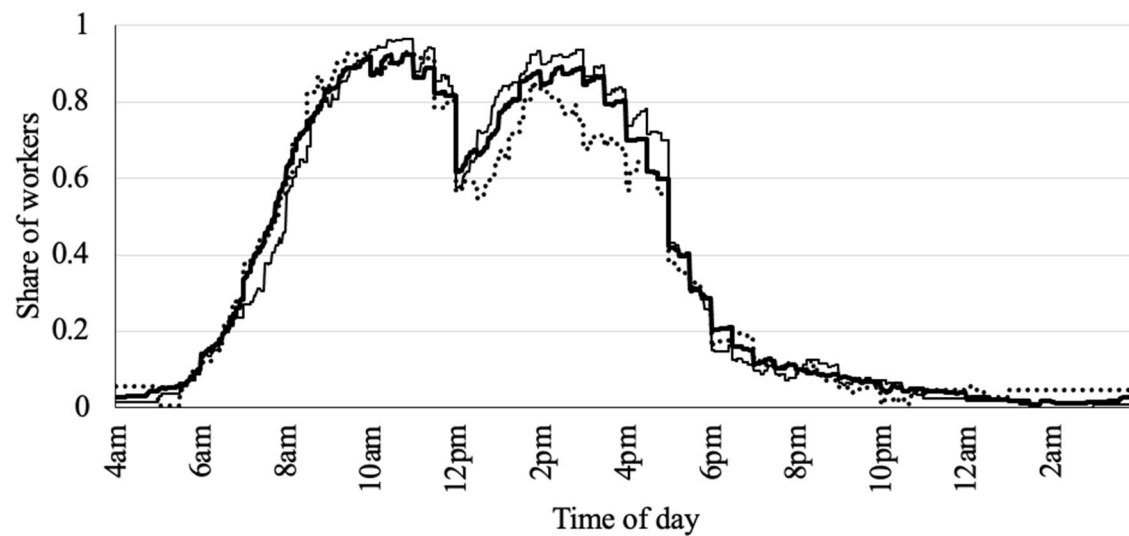
Source: Author's calculations using ATUS-LV module (2017–2018)



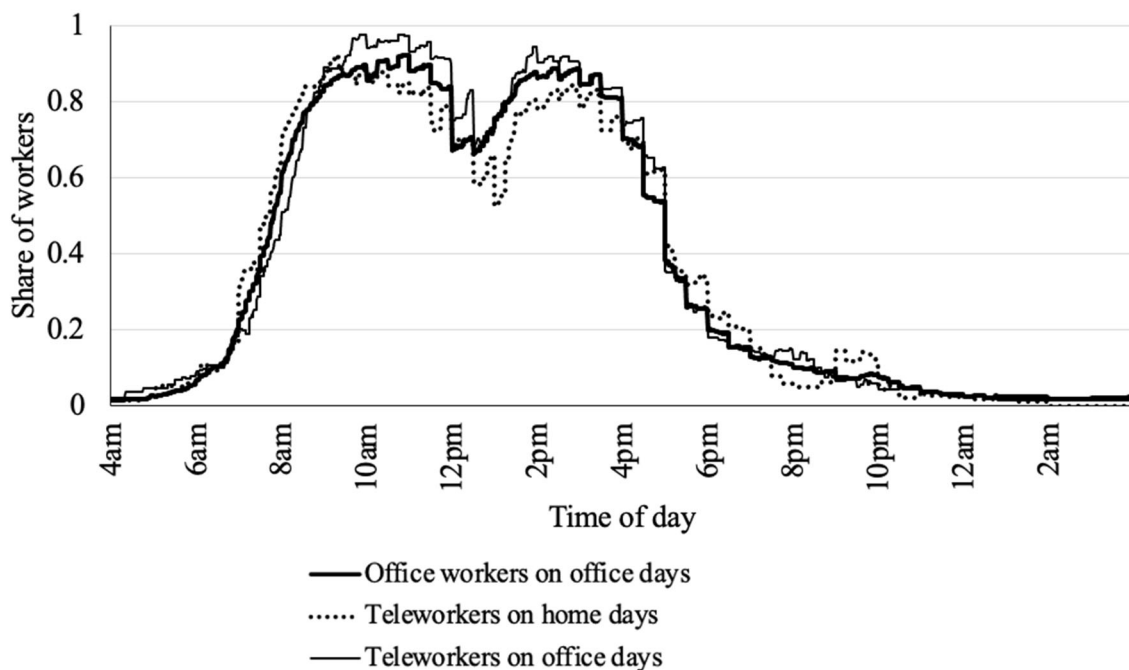
**Fig. 2** Percent of teleworkers among full-time wage and salary workers by industry, 2017–2018

Source: Author's calculations using ATUS-LV module (2017–2018)

Panel A. Work, men



Panel B. Work, women

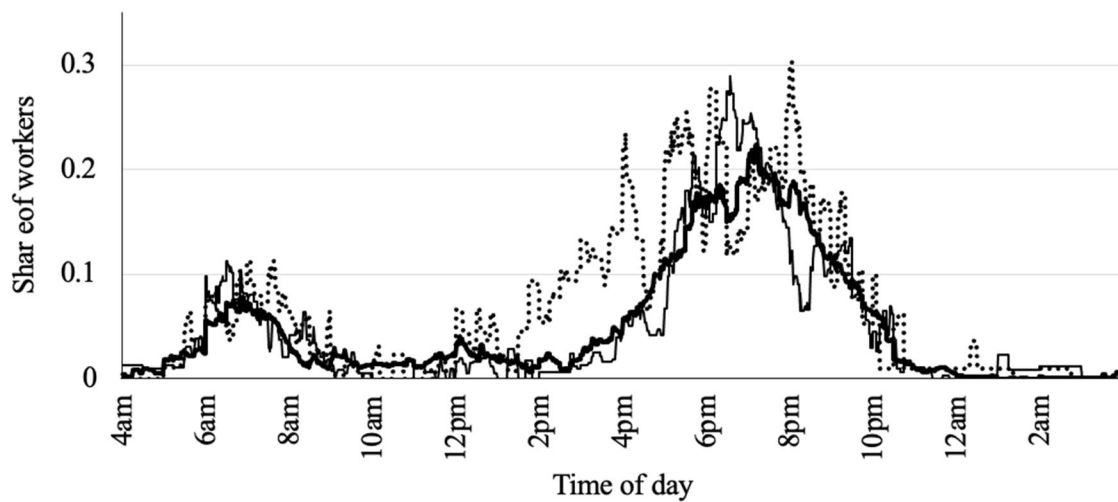


**Fig. 3** Time use by type of worker and work location. Monday-Friday workdays. Work and work-related activities.

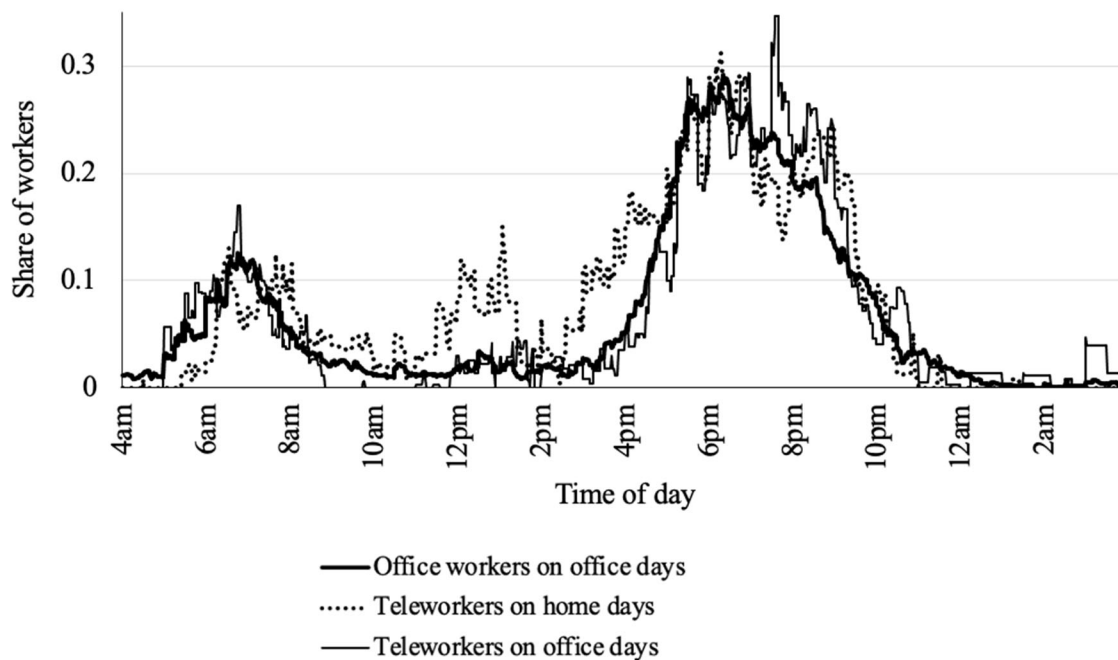
Note: Workdays are days on which the respondent reports at least four hours of work. Sample sizes are: men (N = 794, 81, 128), women (N = 1,045, 75, 106) for the three groups of workers respectively.

Source: Author's calculations using ATUS-LV module (2017–2018)

Panel A. Household production and care, men



Panel B. Household production and care, women

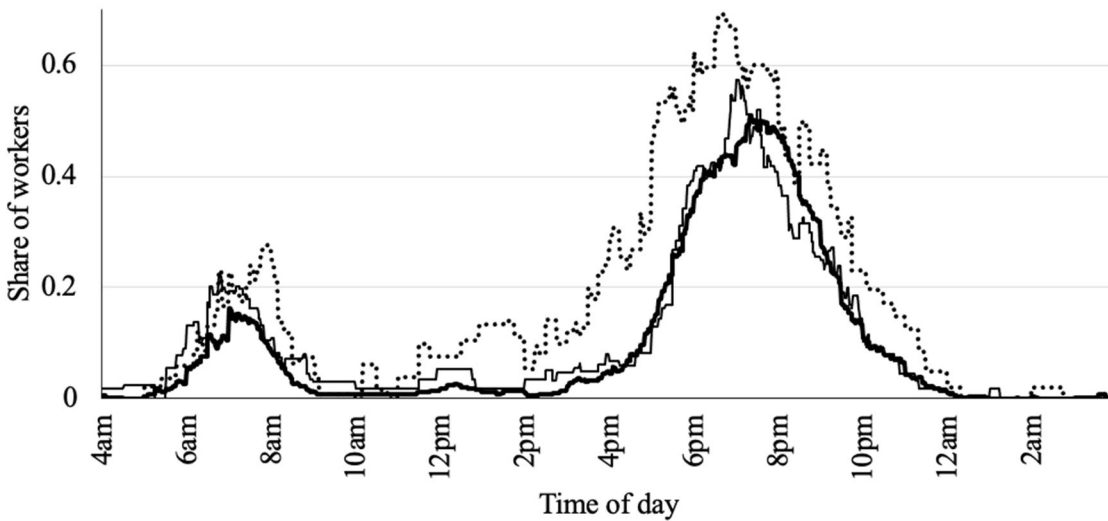


**Fig. 4** Time use by type of worker and work location. Monday-Friday workdays. Household production and care.

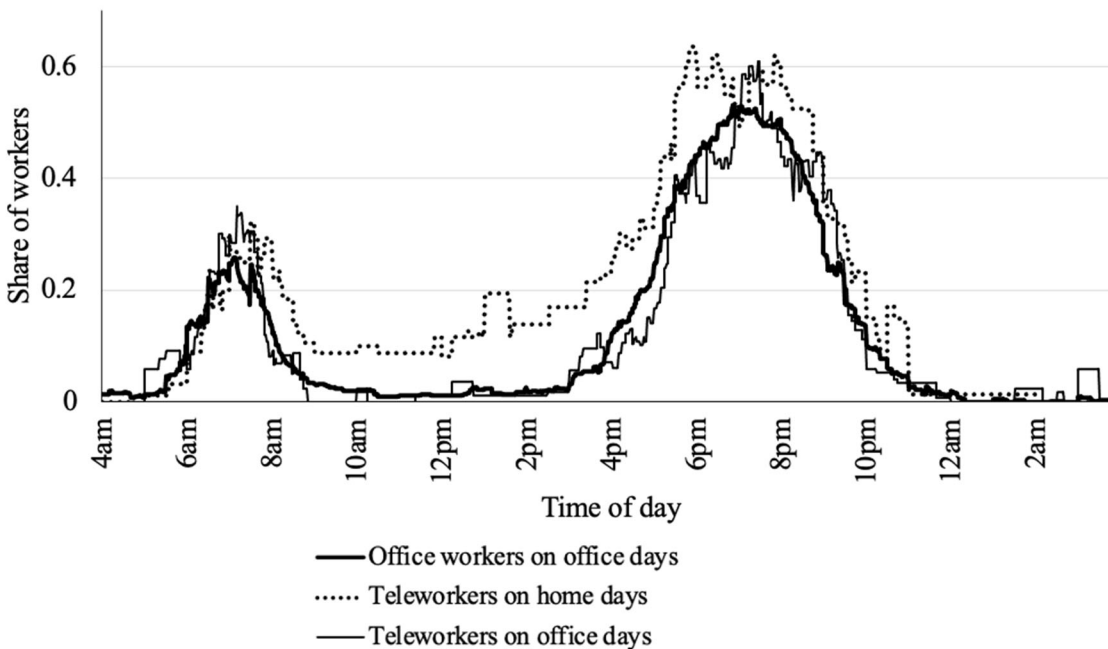
Note: Workdays are days on which the respondent reports at least four hours of work. Sample sizes are: men (N = 794, 81, 128), women (N = 1,045, 75, 106) for the three groups of workers respectively.

Source: Author's calculations using ATUS-LV module (2017–2018)

Panel A. Time with children, fathers



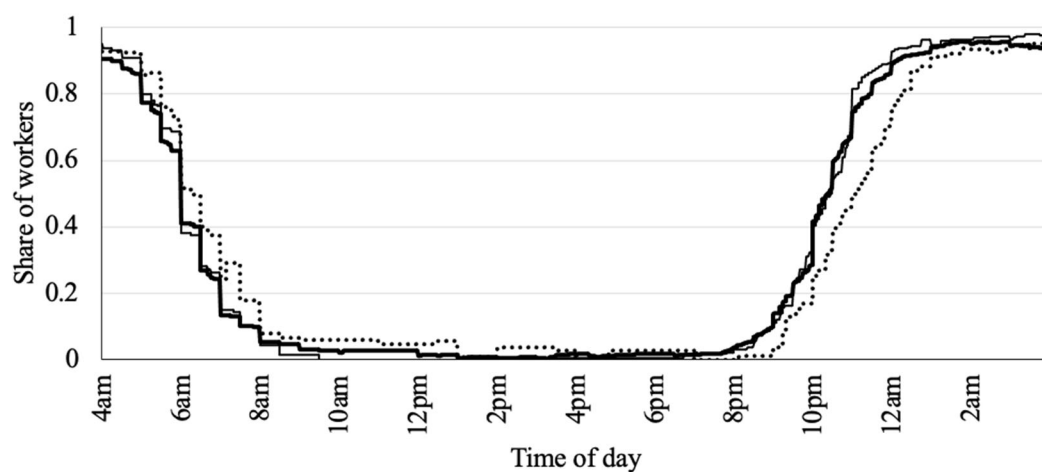
Panel B. Time with children, mothers



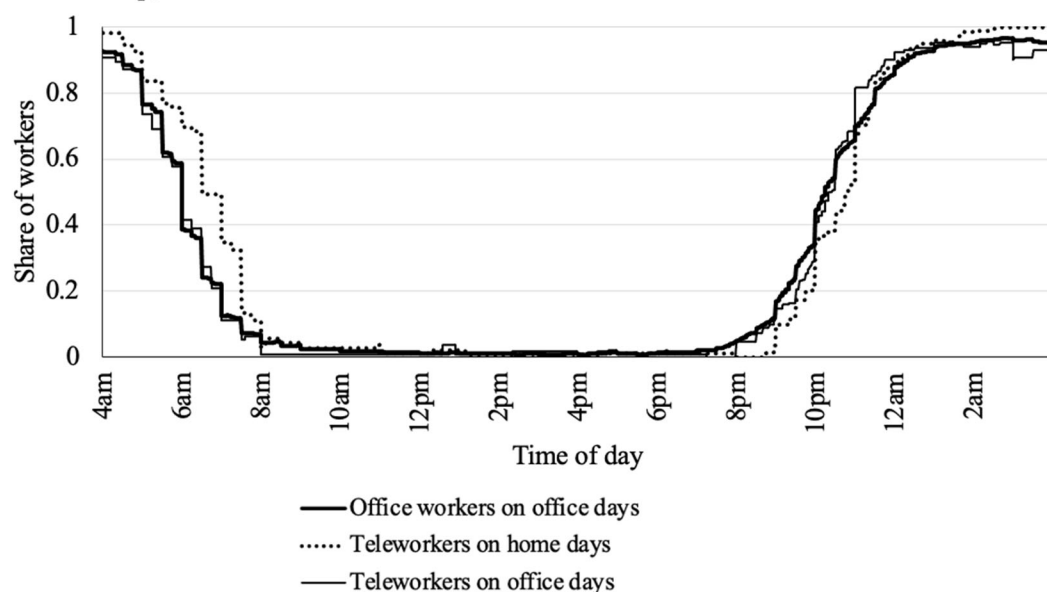
**Fig. 5** Time use by type of worker and work location. Monday-Friday workdays. Time with children.

Note: Workdays are days on which the respondent reports at least four hours of work. Sample sizes for time with own children graphs are: Fathers (N = 409, 45, 67), Mothers (N = 476, 31, 54) for the three groups of workers respectively. Time with children includes time spent working. Source: Author's calculations using ATUS-LV module (2017–2018).

Panel A. Sleep, men



Panel B. Sleep, women



**Fig. 6** Time use by type of worker and work location. Monday-Friday workdays. Sleep.

Note: Workdays are days on which the respondent reports at least four hours of work. Sample sizes are: men (N = 794, 81, 128), women (N = 1,045, 75, 106) for the three groups of workers respectively.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table 1. Sample means by work-at-home status and gender

	Men			Women		
	Home-based teleworkers	Occasional teleworkers	Office workers	Home-based teleworkers	Occasional teleworkers	Office workers
N	152	307	2,028	171	288	2,663
Share of white-collar workers	0.065	0.116	0.819	0.051	0.084	0.865
Wage, 2018 dollars	47.984* (25.066)	49.187* (27.759)	34.751 (22.948)	<b>30.280**</b> (15.716)	<b>39.628*</b> (21.243)	26.372 (17.613)
Age	44.422* (11.250)	42.614* (10.82)	39.790 (11.975)	41.983 (11.134)	42.456 (11.165)	41.443 (12.490)
Black, non-Hispanic	0.126	0.105	0.089	0.145	0.111	0.121
Asian, Non-Hispanic	<b>0.031*</b>	<b>0.101</b>	0.081	0.034	0.081	0.061
Hispanic	0.063	0.045*	0.124	0.087*	0.072	0.121
Some college	<b>0.179</b>	<b>0.089*</b>	0.204	<b>0.170*</b>	<b>0.134*</b>	0.268
College degree	0.425	0.484*	0.380	0.345	0.421*	0.314
Graduate degree	0.331	0.363*	0.229	<b>0.309</b>	<b>0.396*</b>	0.213
Lives with spouse/partner	0.792*	0.731*	0.654	0.652*	0.666	0.609
Spouse/partner employed	0.573	0.513	0.503	0.585*	0.600*	0.529
Own children age ≤5	0.149	0.201	0.171	0.126	0.164	0.143
Own children age 6–17	0.284*	0.242	0.206	0.198	0.258	0.206
Other adult age 18–69	0.162	0.164*	0.258	0.232*	0.188*	0.311
Elderly person age 70+	0.030	0.002*	0.025	0.012	0.007	0.039
Has a disability	<b>0.000</b>	<b>0.032</b>	0.024	0.031	0.021	0.028
Foreign born	<b>0.089*</b>	<b>0.160</b>	0.152	0.086	0.100	0.120
Metropolitan residence	<b>0.939</b>	<b>0.976*</b>	0.895	0.958*	0.960*	0.870
Midwest	0.234	0.176	0.239	0.199	0.172	0.240
Northeast	0.225	0.244**	0.178	0.182	0.196	0.174
West	0.275	0.202	0.234	0.257	0.267*	0.191
Year 2018	0.400	0.557	0.504	0.570	0.561	0.519
Weekend/holiday diary day	<b>0.230*</b>	<b>0.321</b>	0.300	0.295	0.350	0.294
Paid hourly	0.162*	0.138*	0.377	<b>0.283*</b>	<b>0.174*</b>	0.524
Union member	0.050*	0.037*	0.091	0.071*	0.058*	0.133
Government job	0.103*	0.103*	0.170	<b>0.087*</b>	<b>0.209*</b>	0.269
<i>Occupation:</i>						
Management	0.200	0.260	0.223	0.187*	0.240*	0.146
Business and financial operations	0.075	0.113*	0.084	0.136*	0.157*	0.080
Computer and mathematical science	0.238*	0.234*	0.098	0.129*	0.070*	0.026
Architecture and engineering	0.071	0.070	0.113	0.003	0.014	0.014
Life, physical, and social science	0.030	0.013	0.037	0.008	0.037**	0.013
Community and social service	0.044	0.021	0.018	0.031	0.025	0.038
Legal	0.026	0.034	0.019	0.012	0.027	0.015
Education, training, library	0.091	0.053	0.058	0.129*	0.150	0.144
Arts, design, entertainment, sports	0.016	0.021	0.032	0.073	0.047*	0.023
Healthcare practitioner, technical	0.012*	0.014*	0.055	0.054*	0.057*	0.156
Sales and related	0.146	0.113	0.133	0.084*	0.065	0.076
Office and administrative support	0.050*	0.047*	0.130	0.154*	0.110*	0.270
<i>Industry:</i>						
Agriculture, mining, construction	0.018*	0.038	0.055	0.006	0.012	0.015
Manufacturing	0.109	0.151	0.144	0.098	0.074*	0.053
Wholesale and retail trade	0.076	0.070*	0.176	0.050	0.079*	0.112
Transportation and utilities	0.016	0.029*	0.055	0.017	0.025	0.025
Information	0.055*	0.036	0.030	0.073	0.028	0.017
Finance activities	0.212*	0.196*	0.093	0.178*	0.179*	0.104
Business and professional services	0.282*	0.313*	0.165	0.249*	0.176*	0.106
Health & education	0.161*	0.100*	0.166	0.257*	0.287*	0.444
Leisure, hospitality, other services	0.050	0.028	0.051	0.037	0.064	0.059
Public administration	0.020*	0.040	0.066	0.036	0.077	0.064

Note: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Standard deviations are in parentheses for continuous variables. Sample: full-time wage and salary workers age 18–64 in white-collar occupations. \* indicate differences are statistically significant with respect to office workers at the 5% level,



based on two-tailed t-tests. In **bold**: differences between home-based and occasional teleworkers are statistically significant at the 5% level based on two-tailed t-tests.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table 2A. Sample means for home-based teleworkers by gender and parental status

	Men		Women	
	No children	Parent	No children	Parent
N	63	88	82	89
Wage, 2018 dollars	<b>46.254</b> (24.929)	<b>50.246</b> (24.700)	28.024* (14.711)	34.946* (16.793)
Age	<b>45.930</b> (13.424)	<b>42.449</b> (7.290)	<b>42.543</b> (12.672)	<b>40.827*</b> (6.790)
Black, non-Hispanic	<b>0.145</b>	<b>0.100</b>	0.158	0.117
Asian, Non-Hispanic	0.021	0.046	0.028	0.046
Hispanic	0.062	0.066	0.113	0.035
Some college	0.192	0.161	0.158	0.192
College degree	0.378	0.487	0.336	0.362
Graduate degree	0.382	0.264	<b>0.259</b>	<b>0.414</b>
Lives with spouse/partner	<b>0.667</b>	<b>0.957</b>	<b>0.542</b>	<b>0.877</b>
Spouse/partner employed	<b>0.480</b>	<b>0.695</b>	<b>0.478</b>	<b>0.805*</b>
Own children age ≤ 5	-	<b>0.343</b>	-	<b>0.388</b>
Own children age 6–17	-	<b>0.657</b>	-	<b>0.612</b>
Other adult age 18–69	0.182	0.137	<b>0.292</b>	<b>0.109</b>
Elderly person age 70+	0.053	0.000	0.019	0.000
Has a disability	0.000	0.000	<b>0.047</b>	<b>0.000</b>
Foreign born	0.112	0.060	<b>0.056</b>	<b>0.148</b>
Metropolitan residence	0.920	0.964	0.971	0.933
Midwest	0.202	0.277	0.181	0.236
Northeast	0.263	0.175	0.136	0.278
West	0.320	0.217	0.294	0.18
Year 2018	0.445	0.343	0.546	0.619
Weekend/holiday diary day	0.224	0.239	0.241	0.408*
Paid hourly	0.177	0.143	<b>0.327*</b>	<b>0.194</b>
Union member	0.077	0.016	0.061	0.091
Government job	0.150	0.043	0.082	0.097
<i>Occupation:</i>				
Management	0.231	0.160	0.159	0.242
Business and financial operations	0.098	0.046	0.138*	0.133
Computer and mathematical science	0.182	0.312	0.136	0.114*
Architecture and engineering	0.061	0.083	0.002	0.005*
Life, physical, and social science	<b>0.053</b>	<b>0.000</b>	0.011	0.000
Community and social service	0.064	0.018	0.035	0.019
Legal	0.030	0.020	0.000	0.036
Education, training, library	0.119	0.055	0.151	0.083
Arts, design, entertainment, sports	0.021	0.009	0.088	0.041
Healthcare practitioner, technical	0.000	0.029	0.051*	0.060*
Sales and related	0.103	0.202	0.063	0.126
Office and administrative support	0.037	0.066	0.161*	0.137
<i>Industry:</i>				
Agriculture, mining, construction	0.000	0.043	0.004	0.009
Manufacturing	0.065	0.167	0.098	0.099
Wholesale and retail trade	0.082	0.068	0.042	0.066
Transportation and utilities	<b>0.029</b>	<b>0.000</b>	0.025	0.000
Information	0.025	0.094	0.096	0.025
Finance activities	0.232	0.186	0.141	0.253
Business and professional services	0.228	0.354	0.235	0.279
Health & education	<b>0.236</b>	<b>0.063</b>	0.296	0.175*
Leisure, hospitality, other services	0.074	0.018	0.043	0.025
Public administration	0.030	0.006	0.020	0.069

Note: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Standard deviations are in parentheses for continuous variables. Sample: full-time wage and salary workers age 18–64 in white-collar occupations. \* indicate differences are statistically significant with respect to male workers at 5% level based on two-tailed t-tests. In **bold**: differences by parental status are statistically significant at the 5% level based on two-tailed t-tests. Source: Author's calculations using ATUS-LV module (2017–2018)

Table 2B. Sample means for occasional teleworkers by gender and parental status

	Men		Women	
	No children	Parent	No children	Parent
N	136	171	139	149
Wage, 2018 dollars	<b>42.381</b> (23.196)	<b>57.701</b> (30.623)	<b>39.211</b> (21.162)	<b>40.192*</b> (21.420)
Age	<b>42.904</b> (12.699)	<b>42.251</b> (7.922)	<b>43.877</b> (13.40)	<b>40.523</b> (6.659)
Black, non-Hispanic	<b>0.145</b>	<b>0.057</b>	<b>0.105</b>	<b>0.121</b>
Asian, Non-Hispanic	0.090	0.114	0.093	0.065
Hispanic	0.048	0.041	0.07	0.075
Some college	0.117	0.054	0.123	0.148
College degree	0.482	0.486	0.405	0.442
Graduate degree	0.311	0.429	0.392	0.400
Lives with spouse/partner	<b>0.545</b>	<b>0.964</b>	<b>0.522</b>	<b>0.863*</b>
Spouse/partner employed	<b>0.400</b>	<b>0.657</b>	<b>0.453</b>	<b>0.801</b>
Own children age ≤ 5	-	<b>0.454</b>	-	<b>0.389</b>
Own children age 6–17	-	<b>0.546</b>	-	<b>0.61</b>
Other adult age 18–69	<b>0.234</b>	<b>0.079</b>	<b>0.272</b>	<b>0.074</b>
Elderly person age 70+	0.000	0.005	0.000	0.017
Has a disability	0.031	0.034	<b>0.036*</b>	<b>0.000</b>
Foreign born	<b>0.127</b>	<b>0.202</b>	0.120*	0.072*
Metropolitan residence	0.979	0.972	0.959	0.961
Midwest	0.141	0.222	0.105	0.262
Northeast	0.232	0.258	0.208	0.18
West	<b>0.275</b>	<b>0.112</b>	0.294	0.231*
Year 2018	0.544	0.574	0.542	0.586
Weekend/holiday diary day	0.316	0.327	0.362*	0.334
Paid hourly	0.178	0.089	<b>0.201*</b>	<b>0.136</b>
Union member	0.027	0.048	0.044	0.076
Government job	0.082	0.129	0.223	0.189
<i>Occupation:</i>				
Management	0.243	0.282	0.24	0.238
Business and financial operations	0.114	0.111	0.181	0.124
Computer and mathematical science	0.237	0.230	0.055*	0.090*
Architecture and engineering	0.082	0.071	0.002*	0.03
Life, physical, and social science	0.020	0.005	0.030	0.045*
Community and social service	0.019	0.023	0.033	0.015
Legal	0.026	0.043	0.009	0.05
Education, training, library	0.040	0.069	0.181	0.107
Arts, design, entertainment, sports	0.021	0.022	0.053	0.037
Healthcare practitioner, technical	0.008	0.021	0.056*	0.056
Sales and related	0.120	0.104	0.061	0.070
Office and administrative support	0.071	0.018	0.093	0.131*
<i>Industry:</i>				
Agriculture, mining, construction	0.043	0.031	0.010	0.014
Manufacturing	0.179	0.117	0.061	0.091
Wholesale and retail trade	0.066	0.075	0.090	0.065
Transportation and utilities	0.021	0.038	0.022	0.029
Information	0.029	0.044	0.029	0.027
Finance activities	0.177	0.219	0.180	0.177
Business and professional services	0.324	0.301	0.169*	0.186
Health & education	0.094	0.107	0.311*	0.254*
Leisure, hospitality, other services	0.035	0.019	0.053	0.078
Public administration	0.032	0.050	0.075	0.079

Note: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Standard deviations are in parentheses for continuous variables. Sample: full-time wage and salary workers age 18–64 in white-collar occupations. \* indicate differences are statistically significant with respect to male workers at 5% level based on two-tailed t-tests. In **bold**: differences by parental status are statistically significant at the 5% level based on two-tailed t-tests. Source: Author's calculations using ATUS-LV module (2017–2018)

Table 2C. Sample means for office workers by gender and parental status

	Men		Women	
	No children	Parent	No children	Parent
N	1,018	1,010	1,440	1,223
Wage, 2018 dollars	<b>32.516</b> (22.940)	<b>38.411</b> (22.522)	<b>25.609*</b> (17.011)	<b>27.796*</b> (18.604)
Age	<b>39.126</b> (13.785)	<b>40.886</b> (8.070)	<b>42.811*</b> (14.173)	<b>38.713*</b> (7.821)
Black, non-Hispanic	<b>0.103</b>	<b>0.066</b>	<b>0.119*</b>	<b>0.123*</b>
Asian, Non-Hispanic	<b>0.072</b>	<b>0.096</b>	0.06	0.061*
Hispanic	<b>0.137</b>	<b>0.104</b>	0.113*	0.134
Some college	<b>0.227</b>	<b>0.166</b>	0.256	0.290*
College degree	0.390	0.363	0.317*	0.308*
Graduate degree	<b>0.184</b>	<b>0.303</b>	<b>0.197</b>	<b>0.240*</b>
Lives with spouse/partner	<b>0.476</b>	<b>0.947</b>	<b>0.529</b>	<b>0.758*</b>
Spouse/partner employed	<b>0.397</b>	<b>0.679</b>	<b>0.437</b>	<b>0.700</b>
Own children age ≤ 5	-	<b>0.454</b>	-	<b>0.408*</b>
Own children age 6–17	-	<b>0.546</b>	-	<b>0.591*</b>
Other adult age 18–69	<b>0.329</b>	<b>0.141</b>	<b>0.361*</b>	<b>0.214*</b>
Elderly person age 70+	0.032	0.015	0.043	0.029
Has a disability	<b>0.025</b>	<b>0.022</b>	<b>0.032</b>	<b>0.018</b>
Foreign born	<b>0.136</b>	<b>0.179</b>	0.107*	0.144*
Metropolitan residence	0.896	0.894	0.872	0.865*
Midwest	0.224	0.265	0.233	0.251
Northeast	<b>0.208</b>	<b>0.129</b>	0.178	0.165
West	0.223	0.253	0.183*	0.205*
Year 2018	0.504	0.504	0.535	0.489
Weekend/holiday diary day	0.300	0.299	0.287	0.305
Paid hourly	<b>0.428</b>	<b>0.293</b>	0.532*	0.508*
Union member	0.084	0.101	0.134*	0.130*
Government job	0.168	0.173	0.271*	0.265*
<i>Occupation:</i>				
Management	<b>0.193</b>	<b>0.273</b>	0.144*	0.147*
Business and financial operations	0.089	0.077	0.076	0.085
Computer and mathematical science	0.093	0.106	0.028*	0.021*
Architecture and engineering	0.113	0.112	0.010*	0.021*
Life, physical, and social science	0.041	0.029	0.012*	0.013*
Community and social service	0.017	0.018	0.028*	0.055*
Legal	0.016	0.025	0.014	0.014
Education, training, library	<b>0.052</b>	<b>0.069</b>	<b>0.143*</b>	<b>0.145*</b>
Arts, design, entertainment, sports	<b>0.042</b>	<b>0.015</b>	0.025*	0.017
Healthcare practitioner, technical	0.055	0.054	0.152*	0.162*
Sales and related	0.142	0.119	0.072*	0.082*
Office and administrative support	<b>0.147</b>	<b>0.101</b>	<b>0.290*</b>	<b>0.232*</b>
<i>Industry:</i>				
Agriculture, mining, construction	0.054	0.057	0.014*	0.017
Manufacturing	0.142	0.146	0.055*	0.051
Wholesale and retail trade	0.191	0.149	0.117*	0.104
Transportation and utilities	0.055	0.055	0.024*	0.028
Information	0.026	0.036	0.019*	0.012
Finance activities	0.103	0.075	0.101	0.110
Business and professional services	0.155	0.181	0.106*	0.108
Health & education	0.156	0.183	<b>0.424*</b>	<b>0.482</b>
Leisure, hospitality, other services	0.057	0.042	<b>0.065</b>	<b>0.047</b>
Public administration	0.060	0.075	<b>0.076*</b>	<b>0.041</b>

Note: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Standard deviations are in parentheses for continuous variables. Sample: full-time wage and salary workers age 18–64 in white-collar occupations. \* indicate differences are statistically significant with respect to male workers at 5% level based on two-tailed t-tests. In **bold**: differences by parental status are statistically significant at the 5% level based on two-tailed t-tests. Source: Author's calculations using ATUS-LV module (2017–2018)

Table 3. Log hourly wage regressions

	Baseline Results			Top Industries			Top Occupations		
	All	Parents	No children	All	Parents	No children	All	Parents	No children
	1	2	3	4	5	6	7	8	9
<i>Panel A. Men</i>									
Home-based teleworker	0.138** (0.064)	0.137* (0.078)	0.152 (0.098)	0.099 (0.064)	0.197** (0.090)	0.027 (0.098)	0.088 (0.072)	0.127 (0.098)	0.071 (0.101)
Occasional teleworker	0.086** (0.038)	0.173*** (0.052)	0.036 (0.051)	0.119** (0.051)	0.245*** (0.062)	0.033 (0.077)	0.095** (0.042)	0.173*** (0.056)	0.055 (0.061)
Observations	2,487	1,270	1,217	850	447	403	1,565	843	722
R-squared	0.486	0.447	0.515	0.463	0.472	0.513	0.476	0.436	0.517
Oster beta (home-based)	0.078	0.107	0.083	0.099	0.209	0.021	0.035	0.101	0.019
Oster beta (occasional)	-0.001	0.102	-0.050	0.065	0.207	-0.005	0.004	0.102	-0.031
N (home-based)	151	88	63	89	57	32	115	68	47
N (occasional)	307	171	136	159	89	70	237	136	101
<i>Panel B. Women</i>									
Home-based teleworker	-0.021 (0.051)	-0.042 (0.060)	-0.012 (0.069)	-0.015 (0.073)	-0.150 (0.091)	0.032 (0.096)	-0.034 (0.062)	-0.128* (0.075)	0.027 (0.085)
Occasional teleworker	0.152*** (0.043)	0.074 (0.053)	0.205*** (0.062)	0.141** (0.066)	0.016 (0.071)	0.204* (0.105)	0.180*** (0.056)	0.065 (0.068)	0.272*** (0.078)
Observations	3,122	1,461	1,661	833	404	429	1,281	605	676
R-squared	0.451	0.492	0.445	0.425	0.544	0.393	0.509	0.530	0.534
Oster beta (home-based)	-0.070	-0.127	-0.043	-0.035	-0.200	0.028	-0.093	-0.198	-0.010
Oster beta (occasional)	0.064	-0.024	0.133	0.060	-0.062	0.138	0.083	-0.023	0.189
N (home-based)	170	88	82	92	47	45	113	59	54
N (occasional)	288	149	139	120	66	54	188	101	87

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Robust standard errors are in parentheses. Industries: information, financial, and prof & business services. Occupations: management, business & financial, computer and math science, legal, arts, design, entertainment, and sales. Control variables include a quartic polynomial in age and indicators for year, Census region, lives with spouse or partner, spouse/partner employed, highest level of educational attainment (some college, college degree, graduate degree), race/ethnicity (black, Asian, Hispanic), own children age 0–5, own children age 6–17, other adult age 18–69, elderly person age 70+, disability status, foreign born status, metropolitan residence, paid hourly, union member, government sector job, 10 industry groups, and 12 occupation groups. Oster betas assume  $\delta = 1$  and  $R_{\max} = 1.3 \cdot \tilde{R}$ . \*\*\* indicates statistical significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table 4A. Conditional mean time use for men, Monday-Friday workdays (minutes/day)

MEN Time Use Activities	Teleworkers on home days	Teleworkers on office days	Office workers on office days	Differences between groups
	1	2	3	4
N	82	128	798	
Work & work-related activities	522	535	538	
Work at main job	505	532	533	3>1*
Work from workplace	-	494	520	3>2***
Work from home	495	34	10	all ***
Work from other place	12	4	2	
Fathers	520	555	538	
Men without children	487	530	527	3>1*
Travel	-	90	88	
Commuting	-	64	60	
Non-work-related	34	28	29	
Personal care	560	577	563	
Sleep	453	468	457	
Grooming	22	36	40	3>1*** 2>1*** 3>2*
Meals	80	67	60	1>3*** 1>2*
Household production	50	47	45	
Food preparation	21	16	16	
Housework	13	20	15	
Buying goods and services	7	5	8	3>2**
Household management	8	7	6	
Care	41	27	25	1>3** 1>2*
Primary child care (fathers)	70	45	45	1>3** 1>2**
Leisure	225	164	181	1>3*** 1>2*** 3>2*
Social activities	45	23	33	1>2*
Sports and active leisure	10	13	15	
Relaxing	22	15	19	
TV and computer for leisure	139	109	106	1>3** 1>2**
With children <18 (fathers)	236	136	140	1>3*** 1>2***
With spouse/partner (couples)	233	179	166	1>3*** 1>2**
With friends	3	13	18	3>1***
With coworkers/clients	7	426	432	3>1*** 2>1***
Alone	694	323	324	1>3*** 1>2***
Child present during work (fathers)	26	5	2	1>3** 1>2*
Secondary child care (kids age<13)	295	108	129	1>3*** 1>2***
Number of work episodes	3.15	2.66	2.51	1>3*** 1>2**
Fathers	3.16	2.71	2.48	1>3**
Men without children	3.19	2.52	2.54	1>3** 1>2*

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Workdays are days on which the respondent reports at least 4 hours of work. The table contains conditional mean values computed from OLS regressions with the following set of controls: year, month, log hourly wage, Census region, lives with spouse or partner, spouse/partner employed, quartic polynomial in age, highest level of educational attainment (some college, college degree, graduate degree), race/ethnicity (black, Asian, Hispanic), own children age 0–5, own children age 6–17, other adult age 18–69, elderly person age 70+, disability status, foreign born status, metropolitan residence, paid hourly, union member, government job, industry, occupation. Column 4 shows whether the group differences are statistically significant.

\*\*\* indicates statistical significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table 4B. Conditional mean time use for women, Monday-Friday workdays (minutes/day)

WOMEN Time Use Activities	Teleworkers on home days	Teleworkers on office days	Office workers on office days	Differences between groups
	1	2	3	4
N	75	106	1,045	
Work & work-related activities	523	553	524	2>3**
Work at main job	521	547	518	2>3**
Work from workplace	-	503	508	
Work from home	507	41	7	all***
Work from other place	6	2	3	
Mothers	464	516	504	
Women without children	545	572	525	2>3**
Travel	-	85	79	
Commuting	-	53	53	
Non-work-related	31	33	26	
Personal care	566	564	579	
Sleep	476	449	460	1>2*
Grooming	35	57	59	3>1*** 2>1***
Meals	53	56	57	
Household production	89	57	63	1>3** 1>2**
Food preparation	33	27	27	
Housework	28	16	17	1>3** 1>2*
Buying goods and services	16	9	12	
Household management	12	5	7	1>2*
Care	30	34	33	
Primary child care (mothers)	60	62	67	
Leisure	200	146	162	1>3*** 1>2***
Social activities	32	23	31	
Sports and active leisure	16	11	9	
Relaxing	28	15	19	
TV and computer for leisure	116	83	93	1>3** 1>2**
With children <18 (mothers)	297	170	174	1>3*** 1>2***
With spouse/partner (couples)	198	170	151	1>3*
With friends	21	14	18	
With coworkers/clients	11	449	439	3>1*** 2>1***
Alone	639	290	284	1>3*** 1>2***
Child present during work (mothers)	57	12	6	1>3** 1>2*
Secondary child care (kids age<13)	389	200	157	1>3*** 1>2***
Number of work episodes	2.75	2.68	2.5	
Mothers	3.46	2.59	2.51	1>3*** 1>2***
Women without children	2.57	2.70	2.50	

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Workdays are days on which the respondent reports at least 4 hours of work. See the notes for Table 4A for control variables. Column 4 shows whether the group differences are statistically significant.

\*\*\* indicates significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table 5. Mean wake-up time of teleworkers and office workers (clock time)

	1	2	3	4
	Teleworkers on home days	Teleworkers on office days	Office workers on office days	Differences between groups
<i>Panel A. Weekday workdays</i>				
Men	6:36	6:17	6:10	1>3***1>2***
Women	6:48	6:05	6:08	1>3***1>2***
N men	78	124	752	
N women	74	99	994	
	Home-based teleworkers	Occasional teleworkers	Office workers	
<i>Panel B. Non-workdays</i>				
Men	7:49	7:30	7:49	
Women	7:45	7:50	7:47	
N men	61	143	947	
N women	86	151	1,326	

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Workdays are weekdays with at least 4 hours of work. Non-workdays are all other days. The sample is restricted to workers who report waking between 4 a.m.–12 p.m. on their diary day. Columns 1–3 show mean wake-up times by worker group. Column 4 shows whether the group differences are statistically significant. \*\*\* indicates significance at the 0.01 level.

Source: Author's calculations using ATUS-LV module (2017–2018)



## APPENDIX

Table A.1. Variables from the American Time Use Survey

<b>Time-Use Category</b>	<b>ATUS Activity Tier Codes and Variables</b>
Work and work-related activities	T1 = 5
Work at main job	T1 = 5 & T2 = 1 & (T3 = 1   T3 = 99)
Work from workplace	T1 = 5 & T2 = 1 & T3 ≠ 2 & TEWHERE = 2
Work from home	T1 = 5 & T2 = 1 & T3 ≠ 2 & TEWHERE = 1
Work from other place	T1 = 5 & T2 = 1 & T3 ≠ 2 & TEWHERE ≠ 1 or 2
Travel	T1 = 18
Commuting	T1 = 18 & T2 = 5. Adjusted using trip tour methodology.
Non-commuting	T1 = 18 (excluding T2 = 5). Adjusted using trip tour methodology.
Personal care	T1 = 1, T1 = 8 & (T2 = 4   T2 = 5), T1 = 11
Grooming	T1 = 1 & T2 = 2
Sleep	T1 = 1 & T2 = 1
Other personal care	T1 = 1 & T2 = 3, 4, 5, or 99, T1 = 8 & T2 = 4, 5
Meals	T1 = 11
Household production	T1 = 2 & T2 ≠ 6, T1 = 7, T1 = 8 (T2 ≠ 4, 5, 7), T1 = 9 & T2 ≠ 3, T1 = 10
Buying goods and services	T1 = 7, T1 = 8 & T2 ≠ 4, 5, 7, T1 = 9 & T2 ≠ 3, T1 = 10
Housework (cleaning, laundry)	T1 = 2 & T2 = 1
Food preparation and clean-up	T1 = 2 & T2 = 2
Home and vehicle maintenance	T1 = 2 & (T2 > 2 & T2 ≤ 99 & T2 ≠ 6, 9)
Household Management	T1 = 2 & T2 = 9
Care	T1 = 2 & T2 = 6, T1 = 3, T1 = 4, T1 = 8 & T2 = 7, T1 = 9 & T2 = 3
Primary child care for household and non-household children	T1 = 3 & T2 ≤ 3, T1 = 4 & T2 ≤ 3
Adult care	T1 = 3 & (T2 = 4, 5), T1 = 4 & (T2 = 4, 5)
Pet care & veterinary services	T1 = 2 & T2 = 6, T1 = 8 & T2 = 7, T1 = 9 & T2 = 3

Table A.1. Variables from the American Time Use Survey (Continued)

<b>Time-Use Category</b>	<b>ATUS Activity Tier Codes and Variables</b>
Leisure	T1 = 6, T1 = 12, T1 = 14, T1 = 13 & T2>=2, T1 = 15, T1 = 16 & (T2 = 1 & T3<=2), T1 = 50
Social and organizational activities (socializing, Sports and active leisure	T1 = 6, T1 = 12 & T2 ≠ 3, T1 = 14, T1 = 13 & T2>=2, T1 = 15 T1 = 13 & T2 = 1
Relaxing (listening to music, reading, conversations, relaxing, doing nothing)	T1 = 12 & T2 = 3 & T3 ≠ 3, 4, 7, 8
Watching TV and using computer for leisure	T1 = 12 & T2 = 3 & T3 = 3
Time with family and friends	
Time with own children under age 18	All activities where TUWHO = 22 or TUWHO = 40
Time with spouse/partner (excluding work time)	TRTSPOUSE, TRTUNMPART
Time with coworkers/clients (including at work)	TRTCCC_WK
Time with friends	TRTFRIEND
Time Alone (including at work)	TRTALONE_WK
Secondary child care	TRTCC

Note: T1 refers to the first-tier activity code. T2 refers to the second-tier activity code. T3 refers to the third-tier activity code.

TEWHERE refers to the location of the activity. TUWHO refers to who was in the room or who accompanied you on an activity. Trip tour methodology on average increases work-related-travel time by 3 minutes for men and 8 minutes for women compared to reported commute time (Kimbrough, 2019). In turn, non-work-related travel is reduced by the same amount. This methodology classifies as commute time any trip chains that contain no stop of more than 30 minutes and either begin at home and end at work or begin at work and end at home. The travel time (but not the stop time) on such tours is summed to calculate each worker's commute, or work-related time.

Table A.2. Log hourly wage regressions, pooled sample

	All	Parents	No children
	1	2	3
Home-based teleworker	0.164*** (0.062)	0.141* (0.078)	0.192** (0.091)
Occasional teleworker	0.108*** (0.039)	0.185*** (0.053)	0.057 (0.053)
Home-based teleworker x female	-0.197** (0.079)	-0.191* (0.098)	-0.221** (0.112)
Occasional teleworker x female	0.018 (0.057)	-0.119* (0.072)	0.111 (0.081)
Female	-0.111*** (0.019)	-0.107*** (0.027)	-0.109*** (0.027)
N	5,609	2,731	2,878
R <sup>2</sup>	0.481	0.495	0.471

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Robust standard errors are in parentheses. See the notes for Table 3 for control variables.

\*\*\*indicates significance at the 0.01 level, \*\*at the 0.05 level, and \*at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table A.3.A. Conditional mean time use for men, Monday-Sunday typical workday (minutes/day)

MEN Time Use Activities	Teleworkers on home days	Teleworkers on office days	Office workers on office days	Differences between groups
	1	2	3	4
N	92	134	928	
Work & work-related activities	512	534	538	
Work at main job	495	531	533	3>1** 2>1**
Work from workplace	-	493	521	
Work from home	485	34	10	all ***
Work from other place	12	3	3	
Fathers	520	551	539	
Men without children	471	527	528	2>1** 3>1***
Travel	-	90	87	
Commuting	-	63	58	
Non-work-related	36	29	28	
Personal care	568	577	565	
Sleep	458	470	460	
Grooming	21	36	41	3>1*** 2>1*** 3>2*
Meals	84	66	60	1>3*** 1>2**
Household production	54	47	46	
Food preparation	21	16	16	
Housework	15	19	16	
Buying goods and services	8	5	9	3>2**
Household management	9	6	6	
Care	39	26	24	1>3** 1>2*
Primary child care (fathers)	69	45	45	1>3** 1>2**
Leisure	234	166	180	1>3*** 1>2***
Social activities	51	23	32	1>3* 1>2**
Sports and active leisure	10	12	14	
Relaxing	21	15	18	
TV and computer for leisure	134	111	107	1>3** 1>2*
With children <18 (fathers)	238	141	142	1>3*** 1>2***
With spouse/partner (couples)	239	185	169	1>3*** 1>2**
With friends	6	12	18	3>1*
With coworkers/clients	8	426	433	3>1*** 2>1***
Alone	686	324	321	1>3*** 1>2***
Child present during work (fathers)	26	6	3	1>3** 1>2*
Secondary child care (kids age<13)	296	112	132	1>3*** 1>2***
Number of work episodes	3.10	2.66	2.52	1>3*** 1>2*
Fathers	3.11	2.69	2.48	1>3**
Men without children	3.13	2.55	2.55	1>3* 1>2*

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Columns 1–3 contains conditional mean values computed from OLS regressions. See the notes for Table 4A for control variables. Regressions also include controls for Saturday and Sunday/holiday time diaries. Column 4 shows whether the group differences are statistically significant.

\*\*\* indicates significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table A.3.B. Conditional mean time use for women, Monday-Sunday typical workday (minutes/day)

WOMEN Time Use Activities	Teleworkers on home days	Teleworkers on office days	Office workers on office days	Differences between groups
	1	2	3	4
N	89	109	1,176	
Work & work-related activities	514	553	523	2>1* 2>3**
Work at main job	513	547	517	2>3**
Work from workplace	-	503	508	
Work from home	499	41	7	all ***
Work from other place	6	2	3	
Mothers	455	516	507	2>1** 3>1*
Women without children	532	571	523	2>3**
Travel	-	86	79	
Commuting	-	53	52	
Non-work-related	33	35	27	
Personal care	567	561	578	
Sleep	476	447	459	1>3* 1>2*
Grooming	34	57	58	3>1*** 2>1***
Meals	54	55	56	
Household production	92	58	63	1>3** 1>2***
Food preparation	33	26	27	
Housework	28	18	18	1>3*
Buying goods and services	14	9	12	
Household management	15	5	7	1>3* 1>2**
Care	31	33	32	
Primary child care (mothers)	59	60	66	
Leisure	202	149	164	1>3*** 1>2***
Social activities	32	25	32	
Sports and active leisure	16	11	9	
Relaxing	28	16	19	
TV and computer for leisure	118	84	94	1>3** 1>2**
With children <18 (mothers)	315	173	172	1>3*** 1>2***
With spouse/partner (couples)	213	174	152	1>3**
With friends	21	14	18	
With coworkers/clients	12	447	437	3>1*** 2>1***
Alone	631	294	286	1>3*** 1>2***
Child present during work (mothers)	69	14	6	1>3*** 1>2**
Secondary child care (kids age<13)	396	205	158	1>3*** 1>2*** 2>3*
Work episodes	2.74	2.68	2.49	1>3*
Mothers	3.30	2.61	2.52	1>3*** 1>2**
Women without children	2.57	2.7	2.48	

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Columns 1–3 contains conditional mean values computed from OLS regressions. See the notes for Table 4A for control variables. Regressions also include controls for Saturday and Sunday/holiday time diaries. Column 4 shows whether the group differences are statistically significant.

\*\*\* indicates significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table A.4.A. Time use conditional means for men, Monday-Sunday typical day of the week (minutes/day)

MEN Time Use Activities	Home-based teleworkers	Occasional teleworkers	Office workers	Differences between worker types
	1	2	3	4
N	152	307	2,028	
Work & work-related activities	346	374	357	
Work at main job	338	368	352	
Work from workplace	81	273	320	all***
Work from home	233	79	17	all***
Work from other place	24	16	15	
Fathers	354	378	360	
Men without children	326	363	347	
Travel	75	91	87	
Commuting	17	37	37	3>1*** 2>1***
Non-work-related	53	54	48	
Personal care	611	615	607	
Sleep	508	498	496	
Grooming	25	36	37	3>1*** 2>1***
Meals	77	75	68	
Household production	87	79	86	
Food preparation	22	20	23	
Housework	37	31	36	
Buying goods and services	18	18	18	
Household management	10	10	9	
Care	42	34	34	
Primary child care (fathers)	78	57	59	1>3* 1>2*
Leisure	278	247	268	1>2* 3>2**
Social activities	63	53	61	
Sports and active leisure	17	18	21	
Relaxing	35	24	27	
TV and computer for leisure	148	143	150	
With children <18 (fathers)	281	268	250	
With spouse/partner (couples)	304	296	274	1>3*
With friends	27	26	44	3>1** 3>2***
With coworkers/clients	55	258	275	3>1*** 2>1***
Alone	518	323	315	1>3*** 1>2***
Child present during work (fathers)	11	10	3	1>3*
Secondary child care (kids age<13)	362	258	252	1>3*** 1>2***

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Columns 1–3 contains conditional mean values computed from OLS regressions. See the notes for Table 4A for control variables. Regressions also include controls for Saturday and Sunday/holiday time diaries. Column 4 shows whether the group differences are statistically significant.

\*\*\* indicates significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table A.4.B. Time use conditional means for women, Monday-Sunday typical day of the week (minutes/day)

WOMEN Time Use Activities	Home-based teleworkers	Occasional teleworkers	Office workers	Differences between worker types
	1	2	3	4
N	171	288	2,663	
Work & work-related activities	351	333	337	
Work at main job	347	328	332	
Work from workplace	53	230	302	all***
Work from home	266	80	13	all***
Work from other place	28	18	16	
Mothers	263	324	313	2>1** 3>1**
Women without children	391	334	340	1>3** 1>2**
Travel	64	78	79	
Commuting	12	26	33	3>1*** 3>2** 2>1*
Non-work-related	52	52	45	
Personal care	607	613	625	
Sleep	504	497	504	
Grooming	37	48	53	3>1*** 2>1*** 3>2**
Meals	59	64	63	
Household production	115	121	116	
Food preparation	31	39	36	2>1*
Housework	44	45	46	
Buying goods and services	24	26	25	
Household management	16	10	10	1>3*
Care	42	47	45	
Primary child care (mothers)	81	87	83	
Leisure	261	248	237	1>3*
Social activities	74	58	60	
Sports and active leisure	22	11	12	1>3** 1>2*
Relaxing	30	27	25	
TV and computer for leisure	121	136	126	
With children <18 (mothers)	346	307	287	1>3**
With spouse/partner (couples)	290	280	243	1>3** 2>3**
With friends	45	36	30	
With coworkers/clients	46	222	273	3>1*** 2>1*** 3>2**
Alone	510	313	297	1>3*** 1>2***
Child present during work (mothers)	18	21	5	1>3* 2>3**
Secondary child care (kids age<13)	399	357	285	1>3*** 2>3***

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Columns 1–3 contains conditional mean values computed from OLS regressions. See the notes for Table 4A for control variables. Regressions also include controls for Saturday and Sunday/holiday time diaries. Column 4 shows whether the group differences are statistically significant.

\*\*\* indicates significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author's calculations using ATUS-LV module (2017–2018)

Table A.5. Coefficients on ‘Work at home day for teleworker’, Monday-Friday workdays

Time Use Activities	MEN (N=1,008)				WOMEN (N=1,226)			
	Rel. to office workers on office day		Rel. to teleworkers on office days		Rel. to office workers on office day		Rel. to teleworkers on office days	
	Coefficient (S.E)	Oster beta	Coefficient (S.E)	Oster beta	Coefficient (S.E)	Oster beta	Coefficient (S.E)	Oster beta
	1	2	3	4	5	6	7	8
Work & work-related activities	-16.5 (15.9)	-24	-13.1 (18.5)	-23	-0.5 (16.1)	2	-30.2 (21.2)	-41
Work at main job	-27.7* (15)	-31	-26.8 (17.7)	-31	3.1 (16.3)	6	-25.7 (21.3)	-42
Travel	-46.1*** (6.2)	-45	-48.0*** (8.4)	-47	-48.3*** (5.2)	-49	-54.5*** (7.2)	-62
Commuting	-51.4*** (3.7)	-48	-56.6*** (5.9)	-56	-52.4*** (2.9)	-53	-52.6*** (5.3)	-54
Nonwork-related	5.8 (5.5)	4	6.7 (6.6)	5	5.4 (4.8)	6	-1.9 (6.3)	-6
Personal care	-2.8 (12.8)	0	-16.2 (15.5)	-23	-12.8 (10.5)	-18	2.1 (16)	4
Sleep	-4.4 (12.5)	-3	-14.9 (14.7)	-21	15.8 (10)	12	26.3* (15.8)	26
Grooming	-18.4*** (2.8)	-19	-13.7*** (3.6)	-10	-24.1*** (5.1)	-23	-22.4*** (5.8)	-19
Meals	20.4*** (6.4)	21	13.0* (7.6)	9	-4.7 (4.7)	-6	-3.7 (6)	-5
Household production	5.5 (6.7)	6	2.8 (8.1)	2	26.7** (11.8)	28	32.0** (12.6)	37
Food preparation	5.3 (3.3)	6	5.2 (3.9)	5	5.5 (6)	6	5.4 (6.6)	6
Housework	-1.0 (4.2)	-1	-6.3 (5.5)	-8	11.7** (5.4)	13	11.9* (6.4)	14
Buying goods and services	-1.8 (2.1)	-2	1.6 (2.4)	3	3.7 (4.4)	4	7.3 (4.7)	9
Household management	3.0 (3.4)	2	2.3 (3.8)	2	5.8 (4)	6	7.4* (4.1)	8
Care	16.3** (6.6)	14	14.1* (7.2)	9	-2.4 (5.7)	-1	-3.5 (7.3)	-1
Primary child care (parents)	24.6** (9.7)	22	24.6** (11.8)	21	-7.7 (13.5)	-9	-2.0 (14.7)	-1
Leisure	43.7*** (15)	46	60.3*** (16.3)	73	37.3*** (12.6)	36	54.0*** (15.2)	67
Social activities	12.5 (10.7)	10	22.2* (11.4)	26	0.8 (9.1)	1	8.0 (9.8)	12
Physical activity	-4.4 (4.3)	-6	-2.8 (5.3)	-4	6.8 (5.7)	6	5.1 (6.4)	3
Relaxing	3.6 (6)	4	7.4 (6.6)	10	9.3 (8.7)	8	12.9 (9)	12
TV and computer for leisure	32.3** (12.9)	37	29.9** (14.4)	36	23.3** (11.5)	24	33.5** (14)	40
With own children <18 (parents)	96.3*** (18.1)	94	99.9*** (24.2)	99	123.6*** (40.1)	126	127.0*** (43.6)	133
With spouse/partner (couples)	66.7*** (22)	65	53.8** (24.6)	40	47.6* (27.1)	47	28.1 (32.3)	15
With friends	-14.8** (5.8)	-16	-9.7 (6.2)	-9	3.2 (10.8)	2	7.2 (11.2)	8
With coworkers/clients	-424.8*** (17.8)	-421	-419.0*** (24.9)	-396	-428.4*** (15.2)	-423	-438.7*** (28.8)	-482
Alone	369.8*** (23.8)	382	370.4*** (30.7)	394	355.3*** (42.6)	351	349.2*** (47.5)	331
Child present at work (parents)	24.2** (9.8)	24	21.6* (11.1)	18	50.5** (21.4)	51	44.5* (24.3)	40
Secondary child care (kids<13)	166.2*** (37.6)	137	187.4*** (41.2)	179	231.8*** (40.5)	241	188.5*** (46.4)	191

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Robust standard errors are in parentheses. Columns 1–3 contains conditional mean values computed from OLS regressions. See the notes for Table 4A for control variables.

Regressions also include a constant term. Oster betas assuming  $\delta = 1$  and  $R_{\max} = 1.3 \cdot \tilde{R}$ . \*\*\* indicates significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author’s calculations using ATUS-LV module (2017–2018)



Table A.6. Coefficients on ‘Home-based teleworker’, all days

Time Use Activities	MEN (N=2,487)				WOMEN (N=3,122)			
	Rel. to office worker		Rel. to occasional teleworkers		Rel. to office worker		Rel. to occasional teleworkers	
	Coefficient (S.E)	Oster beta	Coefficient (S.E)	Oster beta	Coefficient (S.E)	Oster beta	Coefficient (S.E)	Oster beta
	1	2	3	4	5	6	7	8
Work & work-related activities	-11.2 (20.8)	-23	-27.6 (22.4)	-52	13.5 (17.3)	12	18.3 (21.2)	19
Work at main job	-13.9 (20.4)	-25	-29.7 (21.9)	-54	16.6 (17.1)	15	19.6 (21.2)	18
Travel	-12.4 (10.4)	-13	-16.2 (11.5)	-19	-15.1 (9.3)	-16	-14.4 (10.2)	-15
Commuting	-20.2*** (5.4)	-21	-19.6*** (5.8)	-20	-20.3*** (7.8)	-21	-13.7* (8.1)	-10
Nonwork-related	4.3 (9.3)	4	-1.1 (10.3)	-3	6.6 (5.4)	6	0 (6.6)	-3
Personal care	3.9 (13.1)	10	-4.0 (14.3)	2	-18.0 (11.4)	-20	-6.0 (13.4)	-2
Sleep	11.4 (10.8)	18	9.4 (12.2)	18	0.5 (8.8)	0	5.6 (11.4)	7
Grooming	-12.0*** (2.7)	-12	-10.6*** (3)	-10	-15.8*** (3.5)	-16	-10.6*** (4)	0
Meals	8.2 (5.3)	8	1.9 (6.3)	-1	-3.9 (4.3)	-4	-4.7 (5.1)	-6
Household production	1.3 (8.1)	3	8.0 (8.8)	14	-0.7 (9.7)	2	-5.8 (12.1)	-5
Food preparation	-1.0 (3.1)	-1	1.6 (3.3)	3	-4.2 (3.7)	-3	-7.9* (4.7)	-8
Housework	0.7 (6.8)	1	5.5 (7.1)	8	-2.5 (6.6)	-1	-1.4 (8)	1
Buying goods and services	0.3 (2.7)	2	0.1 (3.5)	2	-0.3 (4.1)	0	-2.1 (4.9)	-3
Household management	1.4 (3.2)	1	0.8 (3.6)	0	6.2* (3.6)	6	5.7 (4)	5
Care	8.4 (5.7)	9	8.1 (6.4)	8	-3.2 (5.5)	-2	-4.6 (6.4)	-3
Primary child care (parents)	18.6* (9.9)	20	20.3* (11.8)	24	-2.4 (10.5)	-1	-6.5 (11.7)	-7
Leisure	10 (15.2)	14	31.6* (16.5)	47	23.5* (13.9)	24	12.5 (16.1)	8
Social activities	2.4 (9.1)	4	9.0 (10.8)	13	13.9 (10.4)	15	15.7 (12)	19
Physical activity	-2.1 (4.2)	-3	-0.1 (5.2)	-1	10.0** (5)	10	10.2* (5.3)	10
Relaxing	10 (9.9)	8	11.4 (10.3)	9	5.5 (6.5)	5	3.8 (7.6)	1
TV and computer for leisure	-1.4 (11.7)	4	5.0 (13.1)	16	-4.8 (13.3)	-4	-15.4 (15.2)	-21
With own children <18 (parents)	30.8 (22)	37	12.7 (27.7)	13	59.4** (29.3)	56	39.2 (34.3)	25
With spouse/partner (couples)	30.1* (16.6)	40	7.5 (20.8)	12	47.3** (23.6)	47	10.4 (27.5)	-9
With friends	-17.2** (8)	-15	1.1 (8.8)	13	14.5 (10.9)	15	9.5 (12.1)	7
With coworkers/clients	-220.1*** (23)	-226	-203.0*** (25.6)	-203	-226.9*** (19.8)	-227	-176.3*** (26.3)	-144
Alone	202.9*** (26.2)	200	195.6*** (29.8)	186	213.5*** (26.2)	209	197.3*** (32.2)	176
Child present at work (parents)	8.1* (4.9)	8	1 (7.1)	-2	13.0* (6.8)	14	-2.9 (9.4)	-8
Secondary child care (parents)	110.0*** (35.4)	103	104.2*** (39.2)	91	114.3*** (37.4)	119	41.8 (42.3)	23

Notes: ATUS leave module weights reweighted separately for equal-day-of-the-week representation for our male and female samples are used. Robust standard errors are in parentheses. Columns 1–3 contains conditional mean values computed from OLS regressions. See the notes for Table 4A for control variables. Regressions also include a constant term and controls for Saturday and Sunday/holiday diaries. Oster betas assuming  $\delta = 1$  and  $R_{\max} = 1.3 \cdot \tilde{R}$ . \*\*\* indicates significance at the 0.01 level, \*\* at the 0.05 level, and \* at the 0.10 level.

Source: Author’s calculations using ATUS-LV module (2017–2018)