

Extending Unemployment Insurance to Contingent Workers

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Abstract

This paper studies the implications of extending unemployment insurance to contingent workers. I document greater dispersion and larger changes in hours worked by contingent workers than by traditional employees, as well as an 11 percent gap in hourly wages. Guided by these results, I develop a model in which individuals choose between contingent work and traditional employment. Contingent work offers hours flexibility to individuals but traditional employment earns a higher wage in equilibrium and has the security of unemployment insurance. Firms hire traditional employees before observing their TFP and must pay administrative costs to hire or fire traditional employees. They can hire (less productive) contingent workers flexibly without these constraints. I find that extending unemployment insurance to contingent workers would generate welfare losses of 2.8 percent in consumption-equivalent units, decreases GDP by 3.0 percent, and increases the aggregate unemployment insurance bill from 2.7 percent of GDP to 7.0 percent.

JEL Codes: J22, J23, J65

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1 Introduction

Contingent workers, which include independent contractors, freelancers, consultants, gig workers, temporary agency workers, and on-call workers, typically conduct work for clients on an as-needed basis without long-term contracts. This benefits workers who prefer flexible work schedules and allows firms to quickly adjust production in response to changes in demand. However, the short-term nature of these arrangements also leaves contingent workers exposed to shocks that impact wages and opportunities for them to work. This exposure is exacerbated by the lack of policy protections (such as minimum wages) and employer-sponsored benefits (like health insurance and pension plans) that are commonly enjoyed by *traditional employees* (non-contingent workers). For example, contingent workers in the U.S. had been excluded from unemployment insurance (UI) programs until the Pandemic Unemployment Assistance (PUA) program of 2020, and even this is only a temporary measure during the Covid pandemic¹. PUA data shows that substantial demand for UI exists from contingent workers².

This paper studies the implications of extending UI benefits to contingent workers, while considering the incentives that drive individuals' choices between traditional employment and contingent work and firms' hiring decisions. In the first part of this paper, I use the National Longitudinal Survey of Youth 1979 to document job characteristics of contingent workers and traditional employees. I find there is greater dispersion in hours worked by contingent workers than by traditional employees: while both types work 44 hours per week

¹In comparison, the Social Security Act first established unemployment insurance programs for traditional employees in the U.S. in 1935.

²In the week ending on March 13, 2021, 7,349,663 contingent workers filed for continued UI benefits under the PUA program ([U.S. Department of Labor, 2021](#)). For comparison, in the same week 4,200,171 traditional employees filed for continued claims under regular state UI programs, and there were 5,515,355 continued claims under the Pandemic Emergency Unemployment Compensation program (a temporary federal program that extends the UI benefit period for traditional employees).

on average, the standard deviation of contingent workers' weekly hours is 27 hours, while it is 15 hours for traditional employees. In addition, I find that contingent workers' annual income is lower by 33 percent, their hourly wages are lower by 11 percent, and their job spells are 11 weeks shorter on average relative to those of traditional employees, even after controlling for demographics and individuals' unobserved heterogeneity.

In the second part of the paper, I develop a structural model that highlights the economic trade offs faced by individuals when deciding to engage in contingent work and by firms when deciding to hire traditional employees and contingent workers. Each period, unemployed individuals choose to remain unemployed or to search for contingent work or traditional employment. They find their chosen job type with some probability and otherwise stay unemployed until the following period. There are two main tradeoffs between job types. First, individuals who find contingent work can flexibly choose the number of hours to supply, whereas traditional employees receive a higher wage but must supply a fixed number of hours. Second, traditional employees receive UI benefits (which is financed by a proportional income tax) when they become unemployed. Contingent workers do not receive UI benefits after losing their job and must live on their savings until they find another job. Heterogeneous preferences over leisure and idiosyncratic productivity shocks give rise to a mix of contingent workers and traditional employees in the equilibrium of the economy.

Firms in the model also face a trade off between hiring traditional employees and contingent workers. They must hire traditional employees before observing their idiosyncratic TFP shock, but they can hire contingent workers afterwards. This creates an asymmetric distortion as firms that receive a high TFP shock can flexibly adjust total employment upwards while firms that receive a low shock cannot immediately lower their total employment. However, traditional employees are more productive since they are hired first and receive

training before production starts. In addition, firms must pay administrative costs to hire or fire traditional employees. This further distorts firms' decisions because firms that expect to receive a low TFP shock have to pay firing costs in order to adjust their employment level downwards. In contrast, firms that expect to receive a high TFP shock can choose between paying adjustment costs to hire (more productive) traditional employees or hiring contingent workers freely without adjustment costs.

In the third part of my paper, I use the model to analyze how permanently extending UI to contingent workers would affect labor markets and welfare in the economy. In the baseline policy, UI replaces 40 percent of wage income for traditional employees after they lose their job. The alternative policy under consideration provides the same 40 percent replacement rate for contingent workers who become unemployed. I find that this policy extension generates welfare losses of 2.8 percent in consumption-equivalent units. In addition, aggregate GDP decreases by 3.0 percent and the aggregate UI bill increases from 2.7 percent of GDP to 7.0 percent.

The source of these welfare losses is twofold. First, contingent workers who lose their job have less incentive to immediately search for a new job because the UI benefits helps them smooth consumption. Thus, total employment decreases. Second, extending UI benefits increases the expected value of searching for contingent work, and so the contingent share of the labor force increases from 14.3 to 25.2 percent. This means that a greater share of the total labor hours are supplied by contingent workers, who are less productive for firms. These two factors together cause GDP to fall. Consequently, while extending UI benefits helps more individuals smooth consumption during unemployment spells, the level decrease in consumption dominates and results in welfare losses.

Related Literature

This paper contributes to the literature that studies the use of contingent work and traditional employment by individuals and firms. Previous papers studying individuals' decisions include [Garin, Jackson, Koustas, and McPherson \(2020\)](#), [Koustas \(2018\)](#), [Mas and Pallais \(2017\)](#), and [Lim \(2017\)](#). [Garin, Jackson, Koustas, and McPherson \(2020\)](#) and [Koustas \(2018\)](#) find that individuals start freelance work in the online platform economy in order to smooth consumption after receiving low-income shocks in their primary job. [Mas and Pallais \(2017\)](#) conduct an experiment to estimate individuals' value of alternative work arrangements. They find that only a tail of workers are willing to pay for scheduling flexibility, although their analysis was limited to job applicants at a national call center and thus is not necessarily representative of the U.S. workforce. [Lim \(2017\)](#) estimates that young mothers value schedule flexibility in self-employment at \$7,400 annually, which is about 25 percent of the average wage and salary earnings among this group. My paper develops a model that incorporates a trade off between schedule flexibility and wage differentials into the choice to be a contingent worker or a traditional employee.

On the firm side, [Dube and Kaplan \(2010\)](#) find that firms outsourced janitorial and security guard work in order to reduce compensation to workers. Similarly, [Goldschmidt and Schmieder \(2017\)](#) study German firms that outsource logistics, cleaning, security, and food services. They conclude that firms outsource these labor services to avoid paying establishment-level wage premia to workers outside their core workforce. By surveying private sector establishments, [Houseman \(2001\)](#) finds that, in addition to reducing labor costs, firms hire on-call, contract, and temporary agency workers in order to adjust for workload fluctuations and staffing absences and to screen workers for regular positions. My paper builds on these empirical results as firms in my model consider relative wages, productivity

risk, and labor adjustment costs when choosing how many traditional employees and contingent workers to hire. I contribute to this strand of the literature by using this model to quantify the impact of each of these factors in firms' hiring decisions. To my knowledge, this is the first paper that considers how individuals' choices to engage in contingent work interact with firms' hiring decisions in general equilibrium.

Finally, this paper is related to the literature that studies UI policies in the presence of incomplete markets ([Hansen and Imrohoroglu, 1992](#); [Shimer and Werning, 2008](#); [Koehne and Kuhn, 2015](#); [Kroft and Notowidigdo, 2016](#); [Braxton, Herkenhoff, and Phillips, 2020](#); [Birinci and See, 2021](#)). Unlike these previous papers, my paper considers both contingent workers and traditional employees and studies how differences between their UI policy treatment distort their labor markets. [Abraham, Houseman, and O'Leary \(2020\)](#) discuss the historical reasons for the UI policy differences and propose several policy reforms. My paper develops a structural model to quantitatively analyze UI policies when considering how the economic trade offs between contingent work and traditional employment shape labor markets. Studying both individual and firm motivations in a single theoretical framework allows me to estimate how UI policy reforms would affect equilibrium outcomes such as the gap between wages of traditional employees and contingent workers, the share of the workforce that chooses contingent work, the distribution of hours worked, the ratio of contingent workers to traditional employees hired by firms, and total output. Taken together, these changes have important implications for the overall welfare effects of UI policy reforms.

Layout

The rest of the paper is organized as follows. Section [2](#) provides details of the data used in this paper and reports the empirical results about the differences between contingent workers

and traditional employees. Section 3 describes the model and equilibrium, while Section 4 discusses the mechanisms of the model. Section 5 describes the calibration and Section 6 presents the main results. Lastly, Section 7 concludes.

2 Empirical Findings

The primary data source used in this paper is the National Longitudinal Survey of Youth 1979 (NLSY79). The NLSY79 is a national panel survey of the cohort of individuals born in the years 1957-1964 and it is administered by the U.S. Bureau of Labor Statistics. The survey was conducted annually between 1979 through 1994, and has been conducted biannually in even-numbered years since then. I mainly use the data starting in 1994 as that was the first year that includes questions about whether an individual worked a contingent job (defined below). I also use the survey’s data on individual and household demographics, work characteristics, income, and assets.

Starting in 1994 (and except in 2000), the NLSY79 included questions about whether the respondent was an independent contractor, consultant, freelancer, temporary agency worker, on-call worker, or contract worker at each job they had during the survey period. I define each of a respondent’s jobs as *contingent work* if they respond “yes” to any of these questions, and *traditional work* if they responded “no” to any of these questions and are not self-employed. Next, I calculated the total income earned by the respondent from each job they worked since the last survey interview, and defined a respondent’s *primary job* as the one that earned them the most income. I then defined a respondent as a *contingent worker* if their primary job was contingent work, and a *traditional employee* if their primary job was traditional work.

Table 1: Summary Statistics

	Employee Only (1)	Contingent Only (2)	Primary Emp, Also Cont (3)	Primary Cont, Also Emp (4)
Observations per Year	5194	263	80	20
Female	0.48	0.46	0.47	0.55
White	0.84	0.77	0.83	0.78
Black	0.13	0.19	0.15	0.18
Non-White, Non-Black	0.03	0.04	0.02	0.04
No Degree	0.09	0.14	0.08	0.07
High School Degree	0.52	0.51	0.50	0.43
Associate or Junior College	0.09	0.08	0.12	0.14
Bachelor's Degree	0.18	0.17	0.18	0.19
Graduate Degree	0.10	0.10	0.13	0.16
Never Married	0.14	0.18	0.14	0.21
Married	0.63	0.53	0.63	0.48
Other Marital Status	0.23	0.28	0.23	0.31
No Children	0.40	0.47	0.39	0.45
1 Child	0.22	0.18	0.21	0.22
2 or More Children	0.38	0.34	0.41	0.34
No Health Insurance	0.12	0.29	0.15	0.37
Current/Former Employer	0.64	0.26	0.51	0.29
Spouse's Employer	0.12	0.16	0.16	0.12
Bought Directly	0.02	0.07	0.06	0.06
Government Program	0.03	0.08	0.04	0.02
Other Source	0.06	0.13	0.08	0.15

Contingent Only are workers who were independent contractors, consultants, freelancers, temporary agency workers, on-call workers, or contract workers at each of their jobs during the survey period. *Employee Only* includes workers who did not hold any contingent work job and are not self-employed. *Primary Emp, Also Cont* are workers whose primary job (the job that earned them the most income in the survey period) was traditional employment but also held at least one contingent work job. *Primary Cont, Also Emp* are workers whose primary job was contingent work but also held at least one traditional employment job.

Table 1 shows summary statistics of individuals in the NLSY79, categorized by job type. Columns (1) and (2) include individuals who only held traditional employment or contingent jobs, respectively. Column (3) includes individuals whose primary job was traditional employment and also worked at least one contingent job in the survey period, while Column (4) includes individuals who were primarily contingent workers but also had at least one traditional employment job in the survey period. Based on the definition of job types,

approximately 5 percent of the individuals in the sample are contingent workers. This share is lower than the 10 percent that has previous studies have reported (Katz and Krueger, 2019; Lim, Miller, Risch, and Wilking, 2019), although this is not surprising as Abraham, Haltiwanger, Sandusky, and Spletzer (2018) find that household survey data usually gives lower estimates of the contingent share than estimates from administrative data due to under-reporting.

Relative to traditional employees, a slightly smaller share of contingent workers are white, married, have at least a high school degree, or have children. This is consistent with previous findings (Gale, Holmes, and John, 2018). In addition, they are less likely to have health insurance, and contingent workers who do have health insurance are less likely to receive it from their employer. The next subsection examines empirical differences between the work characteristics of traditional employment and contingent work that are economically relevant for individuals' decisions over job type.

2.1 Income, Wages, and Employment Spells

In this subsection, I examine the differences between income and work characteristics of traditional employees and contingent workers. Table 2 shows the regression results from the following equation:

$$Y_{it} = \alpha + \beta_1 ContOnly_{it} + \beta_2 EmpPrim\&ContSec_{it} + \beta_3 ContPrim\&EmpSec_{it} + \beta_4 X_{it} + \gamma_i + \theta_t + \epsilon_{it} \quad (1)$$

where i denotes households and t denotes the interview year. The dependent variable Y_{it} is the log of the individual's real annual income ($\ln(AnnInc_{it})$), the log of the real hourly wage in the primary job ($\ln(Wage_{it})$), and the number of weeks worked in the

primary job ($WksInJob_{it}$) in the respective regressions. $ContOnly_{it}$ is an indicator variable for whether the individual only worked contingent jobs since the last survey interview, $EmpPrim\&ContSec_{it}$ is an indicator for whether the individual is primarily a traditional employee but also had contingent work, and $ContPrim\&EmpSec_{it}$ is an indicator for whether the individual is primarily a contingent worker but also had a traditional employment job. Thus, the primary coefficients of interest are β_1 , β_2 , and β_3 as they show the difference in the dependent variables for these groups of workers relative to individuals who only worked traditional employment jobs. The remaining terms are a vector of demographic characteristics³ X_{it} , an individual fixed effect γ_i , a time fixed effect θ_t , and the residual ϵ_{it} .

The first two columns of Table 2 show that individuals who work any contingent job earn less than individuals who only have traditional employment jobs. In particular, individuals earn 32.6 percent less per year if they only work in contingent jobs, even after accounting for individual fixed effects. The difference is slightly lower for individuals who are primarily contingent workers but have traditional employment (18.1 percent) and those who are primarily traditional employees with some contingent work (8.4 percent). The magnitude of these differences is about twice as large when individual fixed effects are excluded, as in Column (1). This suggests there is some selection into who chooses to work contingent jobs. Nevertheless, even after controlling for the unobserved heterogeneity, the differences in annual income for contingent workers remains both statistically and economically significant.

Next, I analyze how much of the difference in annual income for contingent workers comes from differences in hourly wages, job spell length, and hours worked. Columns (3) and

³The demographic characteristics included in X_{it} include the individual's sex, race, education, marital status, and a quadratic in age. In the regressions with individual fixed effects, the sex and race variables are removed from X_{it} as they do not change over time.

Table 2: Coefficients from Regressions on Work Characteristics

	ln(<i>AnnInc</i>)		ln(<i>Wage</i>)		<i>WksInJob</i>	
	(1)	(2)	(3)	(4)	(5)	(6)
<i>ContOnly</i>	−0.632*** (0.032)	−0.326*** (0.027)	−0.232*** (0.021)	−0.108*** (0.019)	−20.40*** (1.702)	−10.98*** (2.069)
<i>Emp&Cont</i>	−0.168*** (0.044)	−0.084* (0.038)	−0.026 (0.026)	−0.032 (0.022)	4.87 (3.987)	6.44** (3.122)
<i>Cont&Emp</i>	−0.286*** (0.084)	−0.181* (0.083)	−0.057 (0.070)	−0.003 (0.053)	−1.42 (−0.247)	−1.53 (10.062)
Demographics	Yes	Yes	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes	Yes	Yes
Individual FE	No	Yes	No	Yes	No	Yes
R^2	0.247	0.690	0.187	0.615	0.070	0.256
N	55252	54626	59732	59241	61011	60530

Statistically significant at the: *10% level **5% level ***1% level

Huber-White standard errors are in parentheses. Person-level survey weights were used in all regressions.

ContOnly is a dummy variable for workers who only held contingent jobs in the survey period, *Emp&Cont* indicates workers whose primary job was traditional employment but also held at least one contingent job, and *Cont&Emp* indicates workers whose primary job was contingent work but also held at least one traditional employment job. Thus, coefficients denote differences from only traditional employees.

(4) show regression results when the dependent variable is the log of the worker's hourly wage from their primary job, and the dependent variable in columns (5) and (6) is the spell length of the individual's primary job (in weeks). The first row of Column (4) shows that individuals who work only contingent jobs have about a 10.8 percent lower hourly wage and their primary job spell was about 11 weeks shorter, relative to individuals who only have traditional employment. This is a difference of 11.8 percent as the spell length for only traditional employees is 93 weeks⁴. These differences in the hourly wage and job spell length are both statistically and economically significant, and they help explain the lower annual income for individuals only working contingent jobs. This contrasts with the results for individuals with both traditional employment and contingent work. Their hourly wages and job spell lengths are not statistically different from those of individuals with only traditional

⁴The true difference in the spell length is likely even larger as the jobs of some contingent workers (such as temporary agency workers, independent contractors, and freelancers) are coded as a single employment spell even though they had assignments or contracts at different firms.

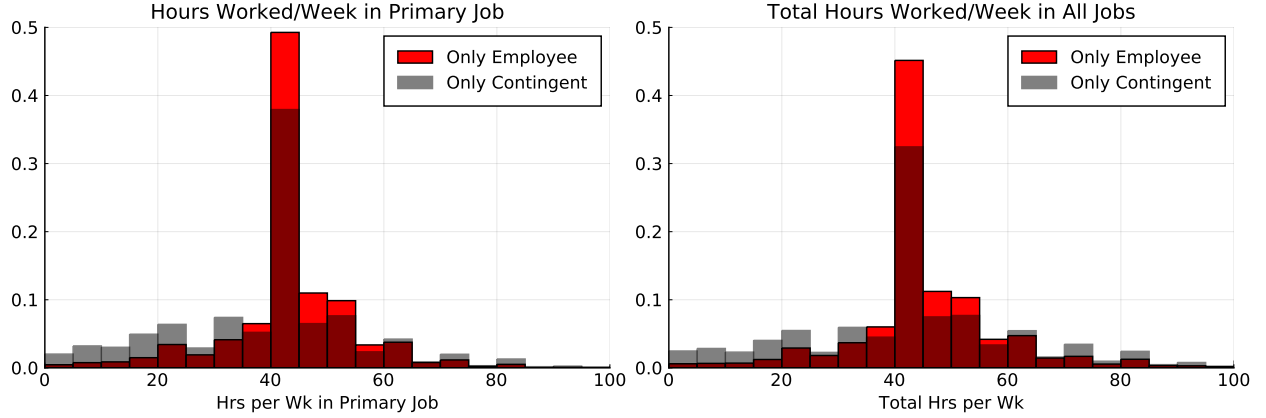
employment.

2.2 Hours Flexibility

Now, I examine the weekly hours of only traditional employees and only contingent workers. On average, both types work approximately 40 hours per week, which is a standard, full-time work week in the U.S. However, the standard deviation for contingent workers is 17 hours, while the standard deviation for traditional employees is 11 hours. The left panel of Figure 1 shows a visual representation of these hours distributions. Looking at the red bars, approximately 50 percent of individuals who are only traditional employees work 40-45 hours per week in their primary job, and an additional 20 percent work 45-55 hours per week. In comparison, the distribution for individuals who are only contingent workers (the light gray bars) is much less concentrated around the standard, full-time workweek. While the greatest share of these workers also falls into the 40-45 hours per week bin, this share is only 35 percent. Furthermore, 50 percent of only contingent workers are fairly evenly distributed over the intervals of less than 40 hours or more than 55 hours per week. Thus, it is more common for contingent workers to work either part-time or over-time, compared to traditional employees who generally have standard, full-time work weeks.

I next consider the possibility that individuals supplement their hours by working multiple jobs. The right panel of Figure 1 displays the distributions of total hours worked per week, which includes hours worked in the primary job and in secondary jobs whose with job spells overlapping with the primary job spell. Contingent workers and traditional employees both work 44 total hours per week on average. The standard deviation of contingent workers is 27 hours, which is nearly double the standard deviation of 15 hours for traditional employees. In addition, the figure shows that total hours is even more uniformly distributed for

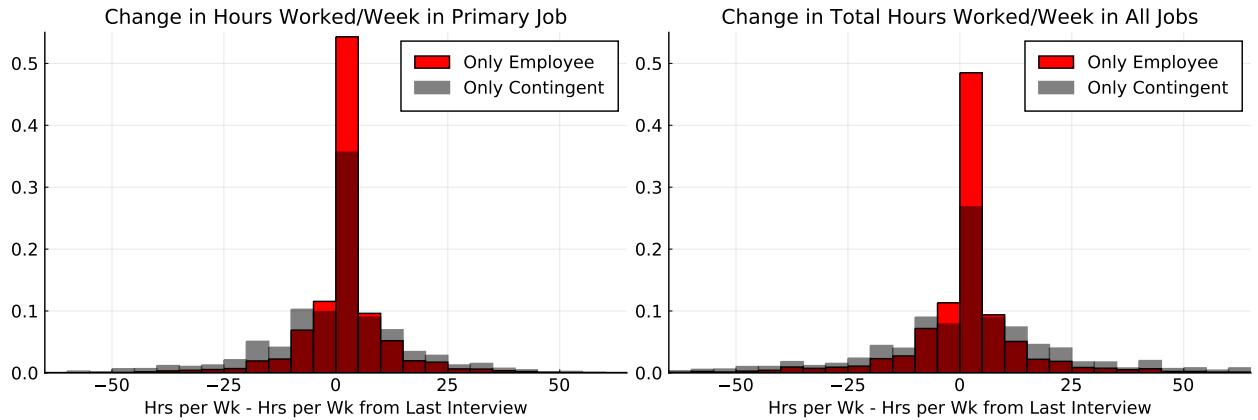
Figure 1: Distribution of Hours Worked, by Job Type



Only contingent are workers who were independent contractors, consultants, freelancers, temporary agency workers, on-call workers, or contract workers at each of their jobs during the survey period. *Only employee* includes workers who did not hold any contingent work job and are not self-employed.

contingent workers than when considering hours only in the primary job, while the distribution for traditional employees remains mostly unchanged. Thus, the differences in the hours distributions between the two types of workers become even starker when considering both primary and secondary jobs.

Figure 2: Distribution of Change in Hours Worked, by Job Type



Only contingent are workers who were independent contractors, consultants, freelancers, temporary agency workers, on-call workers, or contract workers at each of their jobs during the survey period. *Only employee* includes workers who did not hold any contingent work job and are not self-employed.

The evidence up to this point indicates that contingent workers are more likely to work either part-time or over-time. Now look at whether individuals have more flexibility to

change their hours from one period to the next. Figure 2 shows the distribution of the change in hours worked per week, where the change is between the current and previous survey periods (a two-year lag). The left panel includes hours worked in the primary job while the right panel considers total hours worked in all jobs. Most individuals with only traditional employment work the same number of weekly hours from one survey period to the next since their distribution in both panels is concentrated around 0. The distribution for only contingent workers shows that a greater share of these workers have larger changes in hours (in magnitude). This shows that there is greater volatility in the hours worked by contingent workers from one period to the next.

Figure 3: Distribution of Change in Hours Worked for Stayers vs Movers, by Job Type

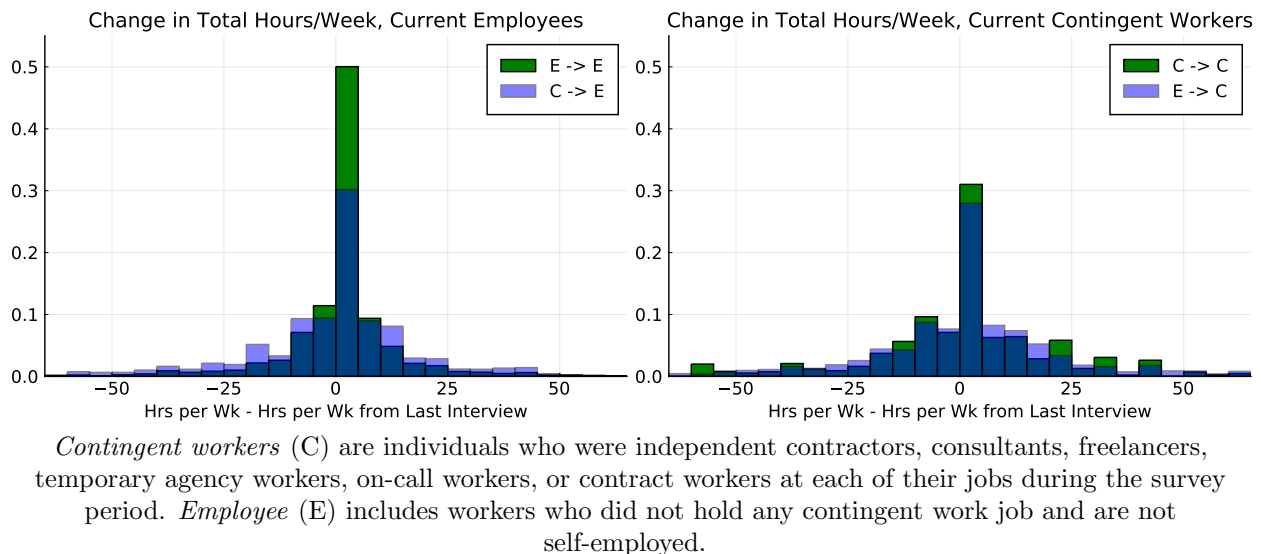


Figure 3 further divides the distributions of change in total hours based on whether individuals had the same job type (dark green bars) or opposite job type (light blue bars) in the previous survey period. The distributions for individuals who were contingent workers in both periods or traditional employees in both periods are slightly more concentrated around 0 but remain largely unchanged. What is interesting to note is that the distribution for individuals who were contingent workers in both periods is very similar to the distributions for

individuals who switched job type (in either direction). These results suggest that it may be easier for contingent workers to adjust their hours than it is for traditional employees.

The data show that contingent workers have lower income, earn lower hourly wages and have a higher dispersion in both weeks and weekly hours worked. The model in the following section captures these features as results of optimal individual behavior in an environment where workers differ in their valuation of leisure and firms, who are subject to productivity shocks, face a trade-off between hiring flexible contingent labor hours and rigid, more productive traditional employees.

3 Model

This section presents a model of the labor market for traditional employees and contingent workers. Unemployed individuals choose whether to remain unemployed or to search for contingent work or traditional employment. Contingent workers get to make an intensive labor supply choice each period and are compensated at the contingent work wage. Traditional employees must work a fixed, given number of hours and receive the traditional employment wage, which is higher in equilibrium than contingent workers' wages. Individuals remain in the same job type until they receive an exogenous separation shock. UI benefits depend on the type of job an individual last held; in the baseline policy, only traditional employees who lose their job receive UI benefits. Thus, when unemployed individuals choose their job type, they must weight hours flexibility against higher wages and the security of UI.

Firms receive persistent idiosyncratic TFP shocks and they choose their factor inputs to maximize the present discounted value of profits. They hire traditional employees before their idiosyncratic TFP is realized, while they can choose contingent worker hours after

they receive their productivity shock. Since traditional employees are hired first, firms have time to train them before production starts. In contrast, contingent workers must spend some of their time on the job to train and thus have lower productivity. In equilibrium, this productivity difference supports higher wages for traditional employees. This gives rise to the trade-off individuals face between hours flexibility and higher wages. In addition, firms must pay hiring or firing costs in order to adjust their level of traditional employment from their choice in the previous period. Thus, firms face a trade-off between productivity, ability to adjust their workforce after observing their TFP shock, and administrative costs to adjust their number of traditional employees. Wages for traditional employees and contingent workers are determined in separate Walrasian markets.

3.1 Environment

Time is discrete and continues forever. The economy is populated by overlapping generations of individuals that live for S periods. Each period t , a continuum of individuals of measure φ is born. Each individual i draws a type θ_i from a distribution $G_1(\theta)$, which is fixed over their lifetime and governs their disutility from work. At each age s of their working life, individuals get an idiosyncratic productivity shock $z_{i,t} \sim G_2(z \mid z_{i,t-1}, s)$. These productivity shocks are perfectly observable by firms, and so individuals who work n hours are compensated for their $z_{i,t}n$ effective hours. Unemployed individuals also receive i.i.d. taste shocks $\epsilon = (\epsilon^E, \epsilon^C, \epsilon^U) \sim G_3(\epsilon)$ over traditional employment E , contingent work C , and unemployment U , and they choose whether to remain unemployed or to search for a traditional employment job or contingent work⁵. If they choose to search, they find their chosen job type with probability

⁵Assuming that individuals receive taste shocks over job type is a computational convenience to smooth the continuation value in their Stage 2 optimization problems, as in [McFadden \(1973\)](#) and [Iskhakov, Jørgensen, Rust, and Schjerning \(2017\)](#). So long as the variance of the distribution of taste shocks is small enough, this assumption will not drive the main results.

p^j (for $j \in \{E, C\}$), and they remain unemployed otherwise. Contingent workers choose how much labor to supply $n_{i,t}^C \in [0, H]$ for that period and earn wage w_t^C . Traditional employees must supply a fixed amount of labor \tilde{n}^E at wage w_t^E . Individuals remain with the same type of job until they receive an exogenous separation shock δ^j that depends on their job type $j \in \{E, C\}$. Unemployed individuals receive UI benefits that depend on their most recent type of job and expire with probability δ^U each period⁶. The government finances these benefits with a proportional income tax τ_t to balance their budget each period. Individuals can save by buying shares $a_{i,t+1}$ of firm ownership at price q_t . They seek to maximize their expected sum of periodical utilities $u(c, n; \theta)$, discounted by the factor $\beta \in (0, 1)$.

The economy also has a continuum of firms of measure φ_F . Each firm receives an idiosyncratic TFP shock $z_t^F \sim G_4(z_t^F \mid z_{t-1}^F)$, which depends on their shock z_{t-1}^F from the previous period. Firms produce output according to the production function $f(\ell_t^E, \ell_t^C)$, where ℓ_t^E and ℓ_t^C are the number of traditional employee and contingent worker hours hired⁷, respectively. Traditional employees are more productive. However, firms must choose ℓ_t^E knowing only the distribution $G_4(\cdot \mid z_{t-1}^F)$, whereas they can choose ℓ_t^C after observing their z_t^F for the period. In addition, firms pay hiring or firing costs $\phi(\ell_{t-1}^E, \ell_t^E)$ to adjust traditional employment from their level in the previous period ℓ_{t-1}^E . These administrative costs are a waste. Firm owners seek to maximize the expected sum of profits, discounted by the rate β .

⁶The assumption that UI benefits expire stochastically simplifies the solution of the model as it eliminates the need to include the benefit duration as a state variable. This assumption is standard in the literature, as in [Koehne and Kuhn \(2015\)](#).

⁷Firms are not restricted to hiring an integer number of workers or all of the hours supplied by a given individual. In addition, firms have perfect information about an individual's idiosyncratic productivity and wages are paid per effective hour. Thus, ℓ_t^E and ℓ_t^C actually denote the effective hours that a firm hires. While a traditional employee must supply a fixed number of hours \tilde{n}^E , these hours can be divided among many firms. This assumption is a simplification to avoid keeping track of firm-worker matches that does not affect the essential trade-offs that the model is meant to highlight.

3.2 Timing

As previously mentioned, time is discrete. Each period is divided into two stages. At the beginning of Stage 1, a new cohort of individuals is born. Individuals are born unemployed with initial stock ownership a_1 and draw their type $\theta_i \sim G_1(\theta)$. All individuals receive their idiosyncratic productivity shock; newborns draw from the unconditional distribution $G_2(z)$. Each unemployed individual then receives their taste shock over job type and chooses whether to remain unemployed or to search for the contingent worker market or the traditional employee market. They find their chosen market with probability p^j for $j \in \{E, C\}$. Firms choose how many traditional employee hours to hire, given their previous period's traditional employment choice and TFP shock. At the end of Stage 1, the wage w_t^E clears the market for traditional employees and firms pay their employment adjustment costs.

In Stage 2, individuals in the contingent worker market choose how much labor to supply. Firms observe their idiosyncratic productivity for the period. They then choose how many contingent worker hours to hire and the wage w_t^C clears this market. Firms produce output and pay wages and dividends, the unemployed receive UI benefits, and individuals consume based on their consumption and savings decisions. At the end of the period, jobs separate with probability δ^j for $j \in \{E, C\}$, unemployment benefits expire with probability δ^U , individuals retire with probability d_s , and individuals age S die.

3.3 Individuals' Problem

This subsection presents individuals' decision problems; I will suppress subscripts i , s , and t when they are not needed for clarity. First, consider an individual who begins Stage 1 unemployed. They know their age s , type θ , stock shares a from the previous period, and their UI replacement rate b^k , which depends on their previous job type and whether their

benefits have expired ($k \in \{E, C, EXP\}$). They receive their taste shocks $\epsilon = (\epsilon^E, \epsilon^C, \epsilon^U)$ over job types, and their productivity shock z is also realized. The log of this shock is the sum of a deterministic component in age and a stochastic component that follows a Markov chain. Given these states, the individual chooses their job type by solving:

$$V_s(a, z, b^k, \epsilon; \theta) = \max \left\{ p^E V_s^E(a, z; \theta) + (1 - p^E) V_s^U(a, z, b^k; \theta) + \epsilon^E, \right. \\ \left. p^C V_s^C(a, z; \theta) + (1 - p^C) V_s^U(a, z, b^k; \theta) + \epsilon^C, V_s^U(a, z, b^k; \theta) + \epsilon^U \right\} \quad (2)$$

where V^E , V^C , and V^U are the Stage 2 value functions of traditional employees, contingent workers, and unemployed individuals, respectively.

In Stage 2, contingent workers take their state $(a, z; \theta)$ as given and choose consumption c , stock shares a' , and how much labor to supply n^C to solve the following problem:

$$V_s^C(a, z; \theta) = \max_{c, a', n^C} u(c, n^C; \theta) + (1 - d_s) \beta \left((1 - \delta^C) \mathbb{E}_{z'} [V_{s+1}^C(a', z'; \theta) \mid z] \right. \\ \left. + \delta^C \mathbb{E}_{z', \epsilon'} [V_{s+1}(a', z', b^C, \epsilon'; \theta) \mid z] \right) + d_s \beta V_{s+1}^R(a'; \theta) \quad (3)$$

$$\text{s.t.} \quad c + qa' = (1 - \tau)w^C z n^C + ((1 - \tau)\pi + q)a$$

$$c, a' \geq 0$$

$$H \geq n^C \geq 0$$

where b^C is the fraction of wages that UI replaces for contingent workers if lose their job, τ is the tax rate, d_s is the probability of retiring at age s , $V_s^R(a'; \theta)$ is the value of retiring with shares a' , and π is firm profits, which are paid as dividends. Traditional employees solve a similar problem, although they take their labor supply \tilde{n}^E as given and only choose

consumption c and stock shares a' . Their Stage 2 problem is:

$$\begin{aligned}
V_s^E(a, z; \theta) = \max_{c, a'} & u(c, \tilde{n}^E; \theta) + (1 - d_s)\beta \left((1 - \delta^E) \mathbb{E}_{z'} [V_{s+1}^E(a', z'; \theta) \mid z] \right. \\
& \left. + \delta^E \mathbb{E}_{z', \epsilon'} [V_{s+1}(a', z', b^E, \epsilon'; \theta) \mid z] \right) + d_s \beta V_{s+1}^R(a'; \theta) \\
\text{s.t.} \quad & c + qa' = (1 - \tau)w^E z \tilde{n}^E + ((1 - \tau)\pi + q)a \\
& c, a' \geq 0 \\
& H \geq \tilde{n}^E \geq 0 \quad \text{given}
\end{aligned} \tag{4}$$

where b^E is the fraction of wages that UI replaces for traditional employees if they lose their job. Individuals who are unemployed in Stage 2 choose consumption c and stock shares a' to solve:

$$\begin{aligned}
V_s^U(a, z, b^k; \theta) = \max_{c, a'} & u(c, 0; \theta) + (1 - d_s)\beta \left[(1 - \delta^U) \mathbb{E}_{z', \epsilon'} [V_{s+1}(a', z', b^k, \epsilon'; \theta) \mid z] \right. \\
& \left. + \delta^U \mathbb{E}_{z', \epsilon'} [V_{s+1}(a', z', b^{EXP}, \epsilon'; \theta) \mid z] \right] + d_s \beta V_{s+1}^R(a'; \theta) \\
\text{s.t.} \quad & c + qa' = (1 - \tau)b^k w^k z \tilde{n}^E + ((1 - \tau)\pi + q)a \\
& c, a' \geq 0
\end{aligned} \tag{5}$$

where $b^{EXP} = 0$ denotes that UI benefits have expired. Lastly, individuals who have retired do not work and choose consumption and stock shares a' to solve the following deterministic problem:

$$\begin{aligned}
V_s^R(a; \theta) = \max_{c, a'} & u(c, 0; \theta) + \beta V_{s+1}^R(a'; \theta) \\
\text{s.t.} \quad & c + qa' = ((1 - \tau)\pi + q)a \\
& c, a' \geq 0
\end{aligned} \tag{6}$$

3.3.1 Preferences

The period utility function is

$$u(c, n; \theta) = \frac{c^{1-\gamma} - 1}{1 - \gamma} - \theta \frac{n^{1+\nu}}{1 + \nu} \quad (7)$$

as in [Heathcote, Storesletten, and Violante \(2014\)](#). γ represents the inverse of the intertemporal elasticity of substitution for consumption, ν controls the elasticity of labor supply. These parameters are common among individuals. The parameter θ governs the weight on an individual's disutility from work. It varies among the population but remains fixed throughout a given individual's life. As an individual's ideal labor supply depends on their θ type, this parameter is important for determining the cross-sectional distribution over job types and hours.

3.4 Firms' Problem

Now, I present the firms' problem. Each firm enters the period knowing their previous period's shock z_{-1}^F and traditional employment level ℓ_{-1}^E . In Stage 1, they choose how many traditional employees ℓ^E to hire for this period in order to maximize expected profits, before observing their TFP shock z^F for the period. Since this productivity shock follows a Markov chain, the firm's Stage 1 value function $V^{F1}(z_{-1}^F, \ell_{-1}^E)$ can be written as:

$$V^{F1}(z_{-1}^F, \ell_{-1}^E) = \max_{\ell^E \geq 0} \int V^{F2}(z^F, \ell^E) dG_4(z^F | z_{-1}^F) - \phi(\ell_{-1}^E, \ell^E) \quad (8)$$

where $V^{F2}(z^F, \ell^E)$ is the firm's value function in the second stage. At the beginning of Stage 2, the firm observes its TFP shock z^F . They take their choice for traditional employees as given and choose contingent worker hours to solve the following problem:

$$V^{F2}(z^F, \ell^E) = \max_{\ell^C \geq 0} z^F f(\ell^E, \ell^C) - w^E \ell^E - w^C \ell^C + \beta V^{F1}(z^F, \ell^E) \quad (9)$$

3.4.1 Production Function

I assume that the firm production function takes the following form:

$$f(\ell^E, \ell^C) = (\ell^E + \lambda \ell^C)^\alpha \quad (10)$$

where $\lambda, \alpha \in (0, 1)$. Traditional employees and contingent workers are perfect substitutes in production, up to the factor λ . This parameter represents the share of working time that contingent workers must spend for on-the-job training; traditional employees receive this training when they are hired in Stage 1, before production begins. The parameter α governs the curvature of the production function, and the assumption that $\alpha < 1$ means there are decreasing returns to scale.

3.4.2 Labor Adjustment Costs

In the model, firms must pay administrative hiring or firing costs to adjust their traditional employment ℓ^E from its level in the previous period ℓ_{-1}^E . Following the prior literature ([Hall, 2004](#); [Ejarque and Portugal, 2007](#)) this adjustment cost takes the form:

$$\phi(\ell_{-1}^E, \ell^E) = \frac{\phi}{2} \frac{(\ell_{-1}^E - \ell^E)^2}{\ell_{-1}^E} \quad (11)$$

This form implies that adjustment costs are convex in the net change in traditional employment and have constant returns to scale. As a result, firms would prefer to spread out adjustments to traditional employment over several periods rather than making large changes in a single period. The ℓ_{-1}^E in the denominator implies that for a given net change, firms with high levels of traditional employment in the previous period have lower adjustment costs. This represents the idea that large firms might maintain a separate division to handle hiring and firing within the firm, and so these administrative duties would be carried out more efficiently and thus at a lower cost.

3.5 Government

The government in the economy provides unemployment insurance. As previously mentioned, the baseline UI policy replaces a fraction b^E of wages for traditional employees after they lose their job, and the replacement rate for contingent workers is $b^C = 0$. The policy analysis in subsection 6.3 will examine the effects of extending the same replacement rate to contingent workers. In both scenarios, the government also levies tax rate τ on labor income, UI benefits⁸, and dividends in order to balance its budget each period:

$$UI = \tau(w^E L^E + w^C L^C + UI + \pi A) \quad (12)$$

where L^E denotes aggregate effective labor hours provided by traditional employees, L^C denotes aggregate effective labor hours provided by contingent workers, and A is aggregate stock shares.

3.6 Equilibrium

Definition 3.1. A *recursive competitive equilibrium* in the economy is value functions and policy functions for individuals and firms, prices $\{w^E, w^C, q\}$, an income tax rate τ , distributions Ω^E of traditional employees and Ω^C of contingent workers over states $x \equiv (a, s, z; \theta)$, distribution Ω^U of unemployed individuals over states (x, b^k) , and a distribution Ψ of firms over states (z_{-1}^F, ℓ_{-1}^E) , such that:

1. given prices and the distributions, the value functions and policy functions solve the individuals' problems in (2), (3), (4) and (5), and the firms' problems in (8) and (9);

⁸I assume that the government taxes UI benefits as this is the current policy in the U.S. Since taxes in the model are proportional (not progressive), this is equivalent to a policy with replacement rate $(1 - \tau)b$ and untaxed benefits.

2. the distributions Ω^E , Ω^C , and Ω^U are derived from individuals' policy functions, initial stock shares a_1 , and the exogenous processes of productivity shocks z , taste shocks ϵ over job type, retirement $\{d_s\}_{s=1}^S$, job finding $\{p^j\}_{j \in \{E,C\}}$ and separation $\{\delta^j\}_{j \in \{E,C\}}$, UI benefit expiration δ^U , and individual types θ ;
3. the distribution Ψ is derived from firms' policy functions and the exogenous process of TFP productivity shocks z^F ;
4. the wage w^E clears the market for traditional employees;
5. the wage w^C clears the market for contingent workers;
6. the price q clears the stock market (the stock shares sum to 1);
7. and the tax rate τ balances the government's budget each period.

4 Discussion of Mechanisms

The main force that drives an individual's decision over job type is the tradeoff they face between higher wages as a traditional employee and the flexibility to marginally adjust their labor supply if they decide to be a contingent worker. In addition, they also consider the probabilities that they will find a job and later separate from their job, and the UI benefits that they will receive once they lose that job. For firms, they must make hiring decisions that take into account 1) the marginal rate of substitution between the two types of workers and the relative wages, 2) the risk involved in choosing traditional employees before observing their TFP shock and then hiring less productive contingent workers afterward if needed, and 3) administrative costs incurred for adjusting traditional employment from its level in the last period and potential adjustment costs in the following period. Each of these forces are

analyzed below.

4.1 Individuals: Flexibility versus Wages

In Stage 1 of the model, individuals choose whether to be a traditional employee or a contingent worker. The main factors they consider is that they will receive a higher wage if they choose to be a traditional employee, but they will be able to choose labor hours as a contingent worker⁹. Their decision will then depend on the wage premium between w^E and w^C , their current productivity shock z , and their idiosyncratic weight θ on the disutility from work.

In order to illustrate how these factors affect the job choice, I assume for this discussion that individuals' utility function is slightly different than the calibration in subsection 3.3.1:

$$u(c, n; \theta) = \frac{1}{1 - \gamma} \left(c - \theta \frac{n^{1+\nu}}{1 + \nu} \right)^{1-\gamma} \quad (13)$$

as in Greenwood, Hercowitz, and Huffman (1988). As before, the parameters $\gamma \geq 1$ and $\nu > 0$ are common among individuals while $\theta > 0$ is heterogeneous among the population but fixed across an individual's lifetime. GHH preferences give a closed form solution for the optimal labor supply that does not depend on assets, which is convenient for this discussion to illustrate individuals' tradeoffs. However, the same arguments hold true when utility is additively separable in consumption and labor as I assume for the main quantitative results of the paper.

Individuals who chose to be contingent workers will decide how many labor hours to supply by solving their problem (3). The first order conditions give the following (interior) solution

⁹For now, I assume the job finding and separation probabilities are the same for both job types and that no workers can receive UI benefits; subsection 4.2 discusses how these factors affect individuals' choices. In addition, individuals will also consider their taste shocks over job types. For this discussion, I assume that the variance of the distribution for taste shocks is so small that they have little influence on individuals' Stage 1 decisions.

for labor hours:

$$n^{C*} = \left(\frac{(1-\tau)w^C z}{\theta} \right)^{\frac{1}{\nu}} \quad (14)$$

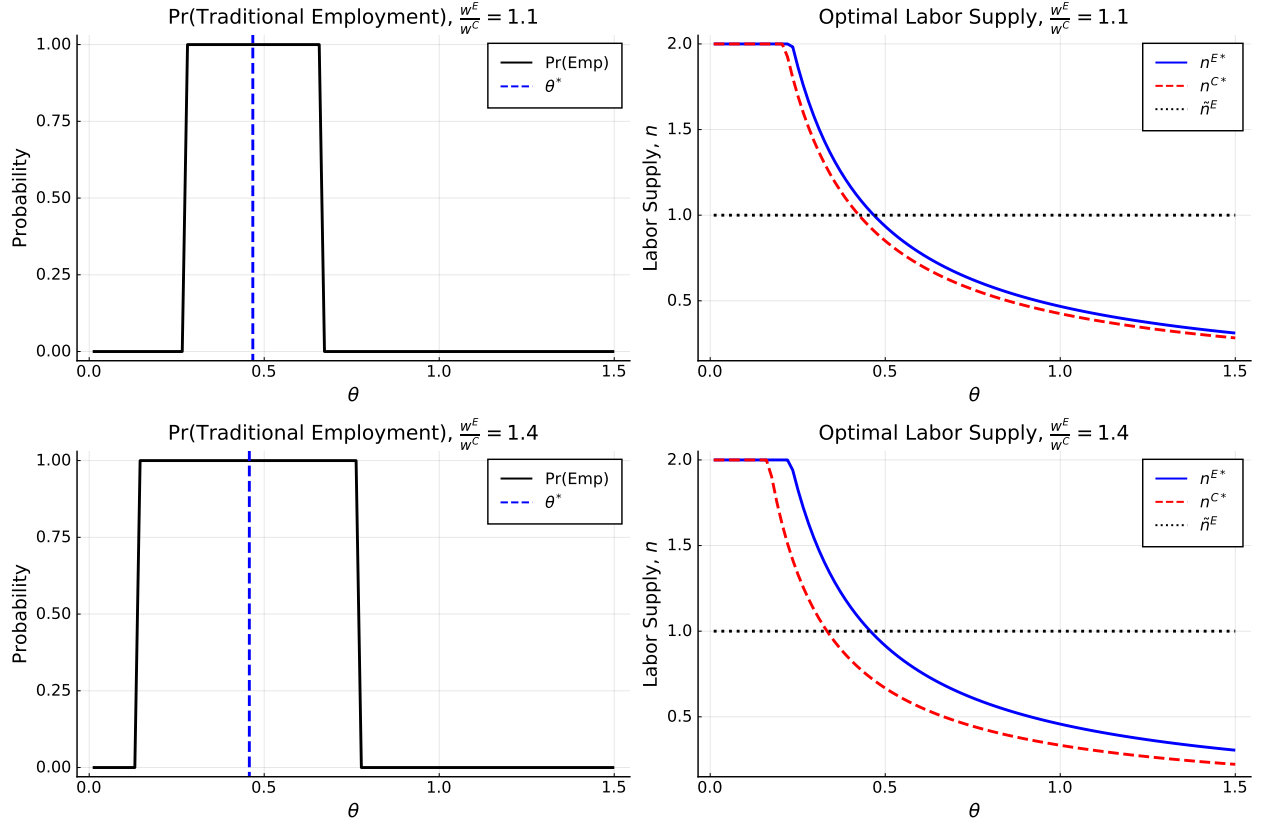
Since $\nu > 0$, a contingent worker's optimal labor supply is increasing in their wage w^C and productivity z and decreasing in θ . This makes sense since higher values of θ means the individual gets more disutility from work. Similarly, if traditional employees were able to choose their labor supply, the solution would be:

$$n^{E*} = \left(\frac{(1-\tau)w^E z}{\theta} \right)^{\frac{1}{\nu}} \quad (15)$$

Recall that \tilde{n}^E denotes the given labor hours that an individual must supply if they decide to be a traditional employee. When the productivity shock z and wage w^E is such that the desired n^{E*} that solves equation (15) is very close to \tilde{n}^E , then the individual would gain little extra value from being able to choose labor hours as a contingent worker. Thus, for a given productivity shock z , individuals with a $\theta^* = \frac{(1-\tau)w^E z}{(\tilde{n}^E)^\nu}$ will choose traditional employment. Individuals with θ above this θ^* would want to work fewer hours than \tilde{n}^E , while individuals with a lower θ would want to work more hours. However, since $w^E > w^C$, individuals with θ close enough to θ^* will still be willing to work the fixed number of hours \tilde{n}^E since the wage difference is enough to overcome their disutility from deviating from n^{E*} . For individuals with a substantially higher or lower θ , the utility loss from working a fixed number of hours different from their n^{E*} will be too large to compensate for the higher wage. As a result, there will be some cutoffs $\underline{\theta}, \bar{\theta}$ such that an individual will choose traditional employment if $\underline{\theta} \leq \theta \leq \bar{\theta}$ and contingent work otherwise. Furthermore, as the wage premium between w^E and w^C grows larger, the extra income will incentivize more individuals to choose traditional employment: $\underline{\theta}$ will decrease while $\bar{\theta}$ will increase. Figure 4 illustrates these ideas.

The solid black line in the left panels of Figure 4 show the logit probability that an individual

Figure 4: Effect of the Wage Premium on Job Choice



The right panels show the optimal labor supply at the traditional employment wage (solid blue line) and the contingent work wage (dashed red line) relative to the hours that traditional employees must supply (horizontal dotted black line). The solid black line in the left panels shows the logit probability of choosing traditional employment as a function of the weight on disutility from work, for a fixed state and with GHH preferences. The vertical dashed blue line shows the θ^* where n^{E*} and \tilde{n}^E intersect. The wage premium is larger in the bottom panels, which causes individuals to choose traditional employment for a larger range of θ .

chooses traditional employment¹⁰ as a function of their weight on the disutility of work θ , for a fixed state and with GHH preferences¹¹. The right panels of Figure 4 show the labor that traditional employees would supply if they were able to choose their hours (solid blue line) and contingent workers' labor supply (red dashed line), relative to traditional

¹⁰Under the specification and states considered here, individuals choose to search for a job with probability > 0.99 and so the left hand panels reflect the choice between traditional employment and contingent work. They are *probabilities* over this discrete choice due to the assumption that individuals receive taste shocks for job types. Here, I have set the variance of the Type 1 extreme value distribution to 0.01 so the taste shocks have little influence over the job type decision.

¹¹These comparative statics also follow for a utility function that is additively separable in consumption and labor.

employees' required labor supply $\tilde{n}^E = 1$ (black dotted line). The traditional employee wage is $w^E = 0.5$. The wage premium $\frac{w^E}{w^C}$ is 1.1 in the top panels and 1.4 in the bottom panels. The traditional employee's desired labor supply n^{E*} intersects with their required labor supply \tilde{n}^E at $\theta_z^* = 0.47$ (dashed blue line in the left panels). The top left panel shows that individuals with $\theta \in [0.3, 0.7]$ would choose traditional employment in order to receive the higher wage, even though they would have to work a sub-optimal number of hours. Individuals outside this range choose contingent work so they can work either substantially more or substantially fewer hours than \tilde{n}^E . The wage premium is larger in the bottom panels. As discussed above, this increases the range of θ over which individuals choose to be traditional employees.

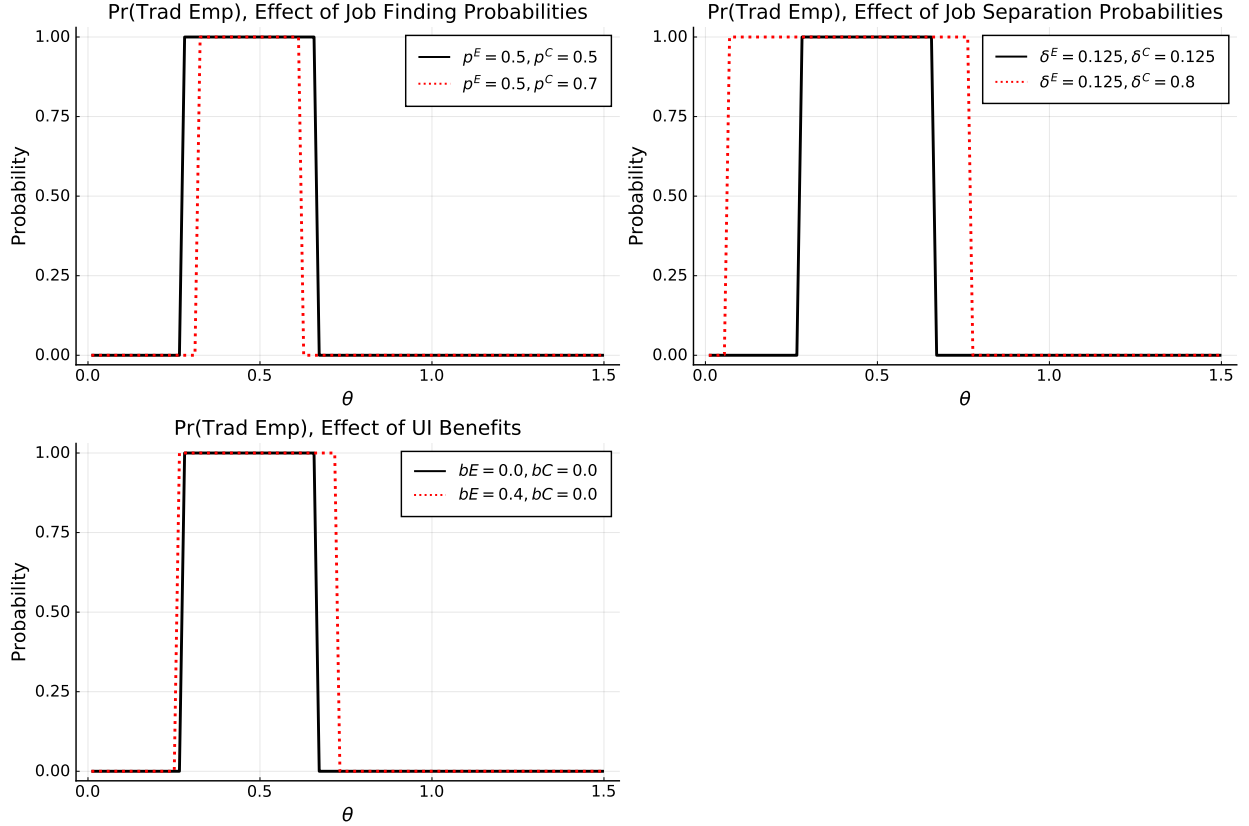
4.2 Individuals: Job Finding/Separation Probabilities and UI

In order to isolate the main tradeoff in the previous subsection, I assumed the job finding and separation probabilities are the same for both job types and that no workers can receive UI benefits. In this section, I consider how differences in these parameters by job type (to reflect differences observed in the data and the calibration used in the main analysis of section 6) affect job search decisions. The tradeoff between the wage premium for traditional employees and the hours flexibility for contingent workers still remains and so there will be a range of θ values for which individuals will choose traditional employment. This analysis illustrates how this range changes under higher job finding and separation rates for contingent workers and UI benefits for traditional employees.

Figure 5 shows the logit choice probability that an individual chooses traditional employment as a function of their weight on the disutility of work θ . As before, I assume that individuals have GHH preferences and I fix the traditional employee wage w^E to 0.5 and the wage premium to 1.1. In each panel, the solid black line is the same as in the top left panel

of Figure 4 (where the job finding rates are $p^E = p^C = 0.5$, the job separation rates are $\delta^E = \delta^C = 0.125$, and the UI replacement rates are $b^E = b^C = 0$). The red dotted line shows the logit choice probabilities after changing the parameter considered in each panel, *ceteris paribus*.

Figure 5: Effect of Job Finding/Separation Probabilities and UI on Job Choice



This figure shows how the job finding and separation rates (p^j and δ^j , respectively) and UI benefits affect the logit choice probability of choosing traditional employment as a function of the weight on disutility from work. The solid black line shows the logit probability when $p^E = p^C = 0.5$, $\delta^E = \delta^C = 0.125$, and there are no UI benefits ($b^E = b^C = 0.0$). The top left panel shows that fewer individuals search for traditional employment jobs when the job finding probability for contingent work increases. In the top right panel, contingent workers are more likely to lose their job ($\delta^C = 0.8$), which increases the range of θ over which individuals search for traditional employment. A similar effect occurs when UI benefits are offered to traditional employees with a replacement rate of $b^E = 0.4$.

In the top left panel, the job finding probability for contingent workers is higher at $p^C = 0.7$ than it is for traditional employees. This increases the expected value of searching for contingent work, which shrinks the range of θ over which individuals search for traditional

employment. In contrast, when the job separation rate δ^C increases from 0.125 to 0.8 in the top right panel, even individuals that successfully find a contingent job in the current period are likely to lose it by the following period. It is not ideal for them to lose their job, as evidenced by the fact that no individuals considered here choose to remain unemployed. As a result, individuals within a much wider range of θ search for traditional employment. Lastly, the bottom left panel shows the results when UI benefits are offered to traditional employees with a replacement rate of $b^E = 0.4$ but not to contingent workers. (I am considering the job choice of an individual who currently does not have UI benefits to make this analysis comparable to the others.) Because traditional employees will still receive some income after losing their job, the expected value of traditional employment increases. Thus, there is now a wider range of θ where individuals choose traditional employment.

4.3 Firms: Tradeoff between Productivity and Wages

This subsection discuss the tradeoffs firms face when choosing how many traditional employees and contingent workers to hire. Since all firms behave competitively, they weigh the marginal product of an additional unit of the input against its marginal cost. First, consider a simpler environment where firms can choose both traditional employment and contingent work hours after observing their TFP productivity parameter and without any labor adjustment costs. Then, a firm's profit maximization problem would be:

$$\tilde{V}^F(z^F) = \max_{\ell^E, \ell^C \geq 0} z^F (\ell^E + \lambda \ell^C)^{\alpha_2} - w^E \ell^E - w^C \ell^C + \beta \mathbb{E} [\tilde{V}^F(z_{+1}^F) \mid z^F] \quad (16)$$

In this specification, firms make both hiring choices simultaneously, and the value function no longer depends on the traditional employment level in the previous period because adjustment costs are zero. Because traditional employees and contingent workers are perfect substitutes up to the parameter $\lambda \in (0, 1)$, the first order conditions for ℓ^E and ℓ^C show that

firms hire only contingent workers if $\lambda > \frac{w^C}{w^E}$, and only traditional employees if $\lambda < \frac{w^C}{w^E}$.

Since the adjustment costs and timing assumptions (that traditional employees must be chosen before drawing the TFP shock) in the main model constrain the use of traditional employees but not contingent workers, if $\lambda > \frac{w^C}{w^E}$, firms continue to hire only contingent workers, regardless of the magnitude of adjustment costs or the distribution of firm shocks. Unlike in the simplified model, however, when $\lambda < \frac{w^C}{w^E}$ firms may hire a combination of traditional employees and contingent workers. For the remainder of this section, I assume that $\lambda < \frac{w^C}{w^E}$ and discuss the effects of the additional assumptions on firms' hiring decisions.

4.4 Firms: Timing of Hiring Decisions

Next, I consider a model where firms must hire traditional employees before observing their TFP shock z^F but can hire contingent workers afterward. Since I am assuming $\lambda < \frac{w^C}{w^E}$, firms would ideally like to hire only traditional employees. However, if they hire a large number of traditional employees and end up receiving a low TFP shock, they will not be able to adjust total labor downward. Thus, firms make their Stage 1 decision by balancing the higher marginal productivity of traditional employees against their inability to adjust labor downward in case they observe a low TFP shock. Knowing that they can always hire contingent workers to increase total employment in Stage 2 (if they observe a high TFP shock), firms choose to hire fewer traditional employees in Stage 1.

To see this, consider the firm's Stage 1 problem in this simplified model:

$$\hat{V}^{F1}(z_{-1}^F) = \max_{\ell^E \geq 0} \int \hat{V}^{F2}(z^F, \ell^E) dG_4(z^F \mid z_{-1}^F) \quad (17)$$

Their Stage 2 problem is the same as equation (9) in the full model, except that the contin-

uation value does not depend on ℓ^E :

$$\hat{V}^{F2}(z^F, \ell^E) = \max_{\ell^C \geq 0} z^F (\ell^E + \lambda \ell^C)^\alpha - w^E \ell^E - w^C \ell^C + \beta \hat{V}^{F1}(z^F) \quad (18)$$

First, note that under the assumption that $\lambda < \frac{w^C}{w^E}$, the firm hires at least as many traditional employees as they would if they were to receive the lowest TFP shock. Then, the the first order condition of the maximization problem in (18) shows that a firm's hiring decision for contingent workers is:

$$\ell^C(z^F, \ell^E) = \max \left\{ \frac{1}{\lambda} \left(\frac{\lambda \alpha z^F}{w^C} \right)^{\frac{1}{1-\alpha}} - \frac{\ell^E}{\lambda}, 0 \right\} \quad (19)$$

Since $\alpha, \lambda \in (0, 1)$, the hiring decision $\ell^C(z^F, \ell^E)$ is increasing in z^F and decreasing in ℓ^E . For low enough values of ℓ^E or high enough values of z^F , the firm hires a positive number of contingent workers, while for high values of ℓ^E or low values of z^F , the firm only hires traditional employees. For a fixed choice of ℓ^E , the cutoff \hat{z}^F such that $\ell^C(\hat{z}^F, \ell^E)$ is just 0 is given by:

$$\hat{z}^F(\ell^E) = \frac{(\ell^E)^{1-\alpha} w^C}{\lambda \alpha} \quad (20)$$

After substituting the solution for $\ell^C(z^F, \ell^E)$ into (18) and using the cutoff \hat{z}^F to split the integral in (17) based on whether the firm would hire only traditional employees (low TFP shocks) or a combination of worker types (high TFP shocks), the first order condition for ℓ^E becomes:

$$\int_{\hat{z}^F}^{\hat{z}^F(\ell^E)} \left[w^E - \frac{\alpha z^F}{(\ell^E)^{1-\alpha}} \right] dG_4(z^F | z_{-1}^F) = \int_{\hat{z}^F(\ell^E)}^{\hat{z}^F} \left[\frac{w^C}{\lambda} - w^E \right] dG_4(z^F | z_{-1}^F) \quad (21)$$

Equation (21) shows that firms balance the difference between traditional employees' wage and their expected marginal product in case the firm draws a low TFP shock (left hand side), against the difference between the effective wages of contingent workers and traditional

employees in the event that they receive a high enough TFP shock to hire contingent workers (right hand side).

4.5 Firms: Labor Adjustment Costs

Adjustment costs for traditional employees impose static and dynamic distortions. In the current period, firms consider the cost of adjusting ℓ^E from its value in the previous period, in addition to considering wages and productivity. The distortion is also dynamic because the hiring decision in the current period will affect the static distortion in the next period.

To see this, it is useful to consider again a model where firms hire both types of labor after observing their TFP shock but with adjustment costs for traditional employment. A firm that had traditional employment ℓ_{-1}^E last period and draws TFP shock z^F solves the following problem:

$$\begin{aligned} \bar{V}^F(z^F, \ell_{-1}^E) = & \max_{\ell^E, \ell^C \geq 0} z^F (\ell^E + \lambda \ell^C)^\alpha - w^E \ell^E - w^C \ell^C \\ & - \frac{\phi (\ell_{-1}^E - \ell^E)^2}{2 \ell_{-1}^E} + \beta \mathbb{E} [\bar{V}^F(z_{+1}^F, \ell^E) \mid z^F] \end{aligned} \quad (22)$$

where z_{+1}^F is a TFP shock in the following period. Combining the Envelope Condition and the first order conditions, the firm's choice for traditional employees solves the following nonlinear equation:

$$0 = \frac{w^C - \mu^C}{\lambda} - w^E + \underbrace{\phi \left(1 - \frac{\ell^E}{\ell_{-1}^E} \right)}_{\text{static distortion}} - \underbrace{\beta \frac{\phi}{2} \mathbb{E} \left[1 - \left(\frac{\ell_{+1}^E(z_{+1}^F, \ell^E)}{\ell^E} \right)^2 \mid z^F \right]}_{\text{dynamic distortion}} \quad (23)$$

where $\mu^C = w^C - \lambda \alpha z^F (\ell^E + \lambda \ell^C)^{\alpha-1}$ is the multiplier on the contingent worker non-negativity constraint and $\ell_{+1}^E(z_{+1}^F, \ell^E)$ is the policy function for traditional employment in the next period.

Suppose that a firm draws some high TFP shock, z_{high}^F . Since I am still assuming that $\lambda < \frac{w^C}{w^E}$, if $\phi = 0$ (no adjustment costs), then the firm would not hire any contingent workers and would hire $\bar{\ell}^E(z_{high}^F) = \left(\frac{\alpha z_{high}^F}{w^E}\right)^{\frac{1}{1-\alpha}}$ traditional employees. Further suppose that the firm's traditional employment from the previous period ℓ_{-1}^E is substantially lower than this amount. If $\phi > 0$, then the firm would incur administrative hiring costs to increase its traditional employment to its desired level. Consequently, the firm hires $\ell^E < \bar{\ell}^E(z_{high}^F)$ and supplements its total employment by hiring contingent workers.

The exact choice for ℓ^E solves equation (23) with $\mu^C = 0$. This equation balances the effective wage of a contingent worker ($\frac{w^C}{\lambda}$), the marginal cost of a traditional employee in the current period ($-w^E + \phi \left(1 - \frac{\ell^E}{\ell_{-1}^E}\right)$), and the expected marginal cost next period of adjusting traditional employment from its current level. The firm then chooses contingent workers $\ell^C = \lambda^{\frac{\alpha}{1-\alpha}} \left(\frac{\alpha z_{high}^F}{w^C}\right)^{\frac{1}{1-\alpha}} - \frac{\ell^E}{\lambda}$. In this case, contingent workers provide flexibility to the firm to increase its total employment level, while spreading the traditional employment adjustment costs across several periods.

Now, consider a firm that draws a low TFP shock, z_{low}^F . Without labor adjustment costs ($\phi = 0$), the firm would choose $\bar{\ell}^E(z_{low}^F) = \left(\frac{\alpha z_{low}^F}{w^E}\right)^{\frac{1}{1-\alpha}}$. If the firm came into the period with a higher level of traditional employment ℓ_{-1}^E and $\phi > 0$, then it would have to pay firing costs to adjust its employment level downward. As a result, it hires $\ell^E > \bar{\ell}^E(z_{low}^F)$ traditional employees. Since the employment level is already higher than the ideal level, the firm does not hire any contingent workers. After setting $\ell^C = 0$ and substituting in for the form of μ^C , equation (23) becomes:

$$\alpha z_{low}^F (\ell^E)^{\alpha-1} = w^E - \phi \left(1 - \frac{\ell^E}{\ell_{-1}^E}\right) + \beta \frac{\phi}{2} \mathbb{E} \left[1 - \left(\frac{\ell_{+1}^E(z_{+1}^F, \ell^E)}{\ell^E} \right)^2 \middle| z_{low}^F \right] \quad (24)$$

The hiring choice for traditional employment ℓ^E solves equation (24). It balances the

marginal product of traditional employees against the sum of the marginal cost in the current period and the expected marginal cost of adjusting traditional employment next period. Unlike in the first case with high TFP shocks, the ability to flexibly hire contingent workers does not help firms that draw low TFP shocks and are stuck with employment levels that are too high. Due to this asymmetric distortion, firms hire fewer traditional employees in the stationary equilibrium of the economy with labor adjustment costs.

5 Calibration

This section presents the calibration of the model parameters and functional forms. The model period is one quarter and I assume the discount factor is consistent with an annual interest rate of 4 percent. Individuals are born at age 25 and leave the model (die) with certainty after 200 periods (50 years). The measure of cohorts at birth is $\varphi = 0.25$ so that the total measure born in a given year is 1. I assume that $\theta \sim Uniform(0.5, 1.5)$, which I discretize with a grid of 9 evenly spaced points. The fixed labor supply of traditional employees is $\tilde{n}^E = 1$ and I allow contingent workers to supply up to twice as many hours. Table 3 summarizes the parameter values used to solve the model.

5.1 Individuals' Productivity Shocks

Each period, individuals receive a productivity shock z . The log of this shock is the sum of a deterministic component μ_s (a quadratic in age) and a stochastic component η that follows an AR(1) process with innovations distributed normally. When a new cohort is born, each individual draws their initial $\eta_{i,0}$ from a Normal distribution. The productivity shock

Table 3: Model Parameters

Parameter	Description	Value	Source or Target
γ	CRRA utility parameter	2.0	Standard value
ν	Elasticity of labor supply	1.0	Arellano, Bai, and Kehoe (2019)
θ	Weight on disutility from work	$\sim U(2, 3)$	Hours distribution
\tilde{n}^E	Fixed hours of traditional employees	1.0	Normalization
H	Max hours of contingent workers	2.0	NLSY79
φ	Measure of cohorts at birth	0.25	Measure 1 annually
β	Discount factor	0.99	4% annual interest rate
p^E, p^C	Job finding probabilities	0.5, 0.7	NLSY97
δ^E, δ^C	Job separation probabilities	0.125, 0.8	NLSY79
σ_ϵ	Variance of taste shocks	0.01	
μ_η	Individuals' prod. shocks, mean	1.0	NLSY79
σ_η	Individuals' prod. shocks, var.	0.1	
ρ_η	Individuals' prod. shocks, persistence	0.9	
α	Labor share	0.66	Standard value
λ	Contingent worker productivity	0.85	Wage gap
ϕ	Scale of adjustment costs	0.5	Ejarque and Portugal (2007)
μ_{z^F}	TFP shocks, mean	1.0	Arellano, Bai, and Kehoe (2019)
σ_{z^F}	TFP shocks, var.	0.12	
ρ_{z^F}	TFP shocks, persistence	0.9	

z follows the process:

$$\log(z_{i,s}) = \mu_s + \eta_{i,s} \quad (25)$$

$$\eta_{i,s} = \rho\eta_{i,s-1} + \xi_{i,s} \quad \text{for } s = 1, \dots, S \quad (26)$$

$$\xi_{i,s} \sim N(\mu_\xi, \sigma_\xi)$$

$$\eta_{i,0} \sim N(\mu_0, \sigma_0)$$

In order to solve the model, I discretize the AR(1) process for individual's productivity shocks using Rouwenhorst's (1995) method and 11 grid points. I assume that the initial distribution for $\eta_{i,19}$ has the unconditional mean $\mu_0 = \frac{\mu_\xi}{1-\rho}$ and variance $\sigma_0^2 = \frac{\sigma_\xi^2}{1-\rho^2}$ of the AR(1) process in equation (26).

5.2 Individuals' Taste Shocks

Following the literature on discrete choices as in [McFadden \(1973\)](#), I assume that unemployed individuals' taste shocks over job types are drawn i.i.d. from the Type 1 extreme value distribution with scale parameter σ_ϵ . For a given state $(a, s, z, b; \theta)$, the expectation over unemployed individuals' Stage 1 problem in (2) is given by the log-sum formula:

$$\mathbb{E}_\epsilon [V(a, s, z, b, \epsilon; \theta)] = \sigma_\epsilon \log \left[\exp \left(\frac{V^U(a, s, z, b; \theta)}{\sigma_\epsilon} \right) + \sum_{j \in \{E, C\}} \exp \left(\frac{\mathbb{E}_{p^j} [V^j(a, s, z, b; \theta)]}{\sigma_\epsilon} \right) \right]$$

where $\mathbb{E}_{p^j} [V^j(a, s, z, b; \theta)] = p^j V^j(a, s, z; \theta) + (1 - p^j) V^U(a, s, z, b; \theta)$ is the expected value of searching for job type $j \in \{E, C\}$. The logit choice probability that an individual chooses a given job type, conditional on their state, takes the following form:

$$P(j \mid a, s, z, b; \theta) = \frac{\exp \left(\frac{\mathbb{E}_{p^j} [V^j(a, s, z, b; \theta)]}{\sigma_\epsilon} \right)}{\exp \left(\frac{V^U(a, s, z, b; \theta)}{\sigma_\epsilon} \right) + \sum_{k \in \{E, C\}} \exp \left(\frac{\mathbb{E}_{p^k} [V^k(a, s, z, b; \theta)]}{\sigma_\epsilon} \right)}, \quad j \in \{U, E, C\}$$

I use these formulas to calculate the expected continuation value when solving the contingent worker's problem in (3), the traditional employee's problem in (4), and the unemployed individual's Stage 2 problem in (5).

5.3 Firm TFP Shocks

As discussed in Subsection 3.1, firm TFP shocks are persistent and i.i.d. across firms. Now, I further assume that the log of TFP shocks follow an AR(1) process with innovations distributed normally. This process is described as follows:

$$\log(z_t^F) = \rho_{z^F} \log(z_{t-1}^F) + \xi_t^F \tag{27}$$

$$\xi_t^F \sim N(\mu_{\xi^F}, \sigma_{\xi^F})$$

When solving the model, I discretize the AR(1) process for $\log(z^F)$ using Tauchen's (1986)

method and 51 grid points¹².

6 Results

I now turn to the main quantitative results from the model. I begin with a comparison of moments generated by the model to their empirical counterparts. The UI policy for these benchmark results provides a 40% replacement rate for traditional employees who lose their job but no UI benefits for contingent workers, since this was the policy in the U.S. until early 2020 when the Pandemic Unemployment Assistance program was passed. The model predicts that 14.3 percent of the labor force is engaged in contingent work, which is slightly higher than the share (10 percent) commonly reported in the data (Katz and Krueger, 2019; Lim, Miller, Risch, and Wilking, 2019). In addition, the equilibrium wage of contingent workers is about 13.1 percent lower than that of traditional employees. This is slightly larger than the 10.8 percent wage gap observed in the NLSY79 from subsection 2.1. Lastly, unemployment spells last 2.8 periods (36.4 weeks) on average in the model. In comparison, BLS (2021) data shows that unemployment spells lasted 30.8 weeks on average between 2009 and 2019, with the monthly averages ranging from 20 to 40 weeks.

Table 4: Model vs Data Moments

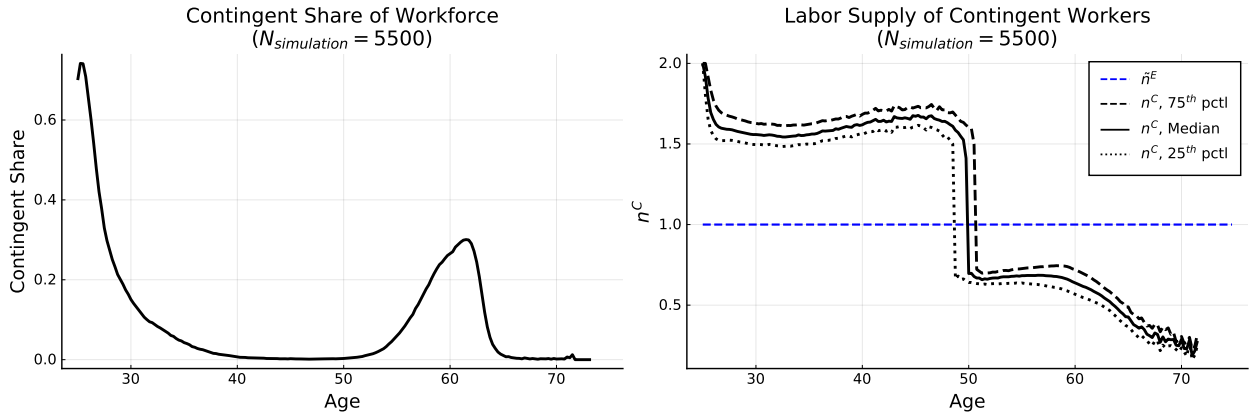
Moment	Model	Data
Contingent share of workforce	14.3%	10.0%
Wage gap $\frac{w^C}{w^E}$	0.87	0.89
Average duration of unemployment	36.4 weeks	30.8 weeks

¹²I use Tauchen’s (1986) method to discretize the AR(1) process for firms’ TFP shocks in order to have a fine grid (51 points with bounds 3 standard deviations above and below the unconditional mean) so I can find a reasonable approximation to the stationary distribution of firms. Meanwhile, using a course grid with 11 points for individuals’ productivity shocks does not significantly affect the results. Thus, I use Rouwenhorst’s (1995) method to discretize this process as Kopecky and Suen (2010) find that this method better matches the conditional and unconditional moments.

6.1 Contingent Work across the Life Cycle

This subsection examines the model dynamics of contingent work across the life cycle. The left panel of Figure 6 shows the share of individuals who choose to be contingent workers at each age after simulating the model for 5,500 individuals. According to this figure, the probability that an individual is a contingent worker is decreasing in age until about 50, at which point the probability again increases. This is consistent with the findings in Bidwell and Briscoe (2009) that individuals are more likely to work as independent contractors early in their careers or later as they near retirement.

Figure 6: Contingent Share and Labor Supply across the Life Cycle (Model)



The left panel shows the life cycle profile of the share of individuals who choose contingent work in the model. The right panel shows the median (solid line), 25th percentile (dotted line), and 75th percentile (dashed line) of hours supplied by contingent workers at each age relative to the required labor supply of traditional employees (horizontal line). The simulations for these figures includes 5,500 individuals.

The right panel of Figure 6 helps shed light on the life cycle profile of contingent work choices. This figure shows the median (solid line), 25th percentile (dotted line), and 75th percentile (dashed line) of contingent workers' labor hours at each age relative to traditional employees' required labor supply $\tilde{n}^E = 1$ (blue dashed line). Early in their careers, individuals who choose contingent work do so in order to work extra hours: since \tilde{n}^E represents a standard full-time, 40 hours per week job, then most contingent workers in the model are working

about 60 to 65 hours per week in most of their early years. This results from the wealth effects on the labor supply. Here, I assume that preferences are additively separable in consumption and labor as described in Subsection 3.3.1, which means the equation for intratemporal substitution between consumption and labor is:

$$n = \left(\frac{(1 - \tau)w^C z}{\theta c^\gamma} \right)^{\frac{1}{\nu}} \quad (28)$$

when the solution is interior. Since individuals are born with low asset levels and cannot borrow, they want to work more hours at the beginning of their career so they can 1) increase consumption and 2) build up stock holdings to insure against idiosyncratic productivity risk and unemployment shocks. As their assets increase over their lifetime (through age 50), individuals' ideal hours decrease and they are more satisfied with working a standard, full-time job. Towards the end of their career, the increase in the share of contingent workers is driven mainly by wealthy individuals who want to work fewer hours. This is consistent with the literature on how individuals use bridge jobs to phase into retirement (Cahill, Giandrea, and Quinn, 2011; James, Swanberg, and McKechnie, 2007).

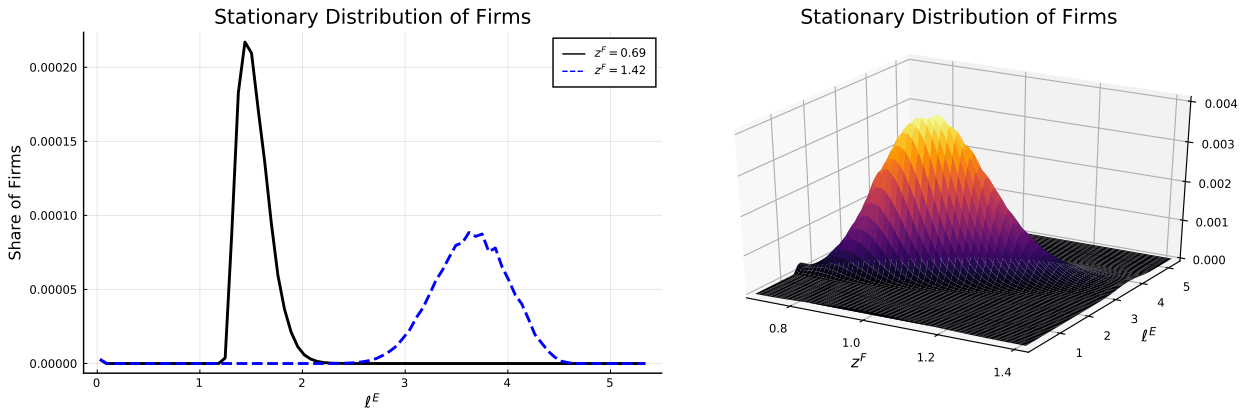
6.2 Stationary Distribution of Firms

I next look at the stationary distribution of firms that results in equilibrium. It is instructive to first consider the distribution that would result if firms were able to hire traditional employees *after* observing their TFP shock and did not have to pay adjustment costs. As discussed in subsection 4.4, all firms with a given shock z^F would choose the same traditional employment level $\ell^{E*}(z^F)$. For a z^F , the distribution would be degenerate with the entire mass of firms that receive z^F at $\ell^{E*}(z^F)$. Similarly, if firms had to hire traditional employees *before* observing their TFP shock but still had no adjustment costs, then all firms that came into the period with a given z_{-1}^F would choose the same traditional employment level

$\ell^{E*}(z_{-1}^F)$. For a given z^F slice of the distribution, there would be a mass of firms at $\ell^{E*}(z_{-1}^F)$ for each shock z_{-1}^F , where the share of firms in the mass would be the transition probability $\pi_{z_{-1}^F, z^F}$.

Once we assume that firms have labor adjustment costs, a firm's choice for traditional employment will depend on both their expected TFP shock and the traditional employment level in the previous period. The right panel of Figure 7 shows the stationary distribution of firms across realizations for TFP shocks and stocks of traditional employment. The left panel shows the slices of this distribution for the lowest TFP shock $z^F = 0.69$ (black solid line) and the highest TFP shock $z^F = 1.42$ (blue dashed line). Given the assumptions on the stochastic process of z^F , the unconditional probability of receiving these shocks is the same and so the integral under the curves is also the same. However, the distribution of firms across ℓ^E is much more disperse for the high shock than for the low shock.

Figure 7: Stationary Distribution of Firms across TFP Shocks and Stocks of Traditional Employment

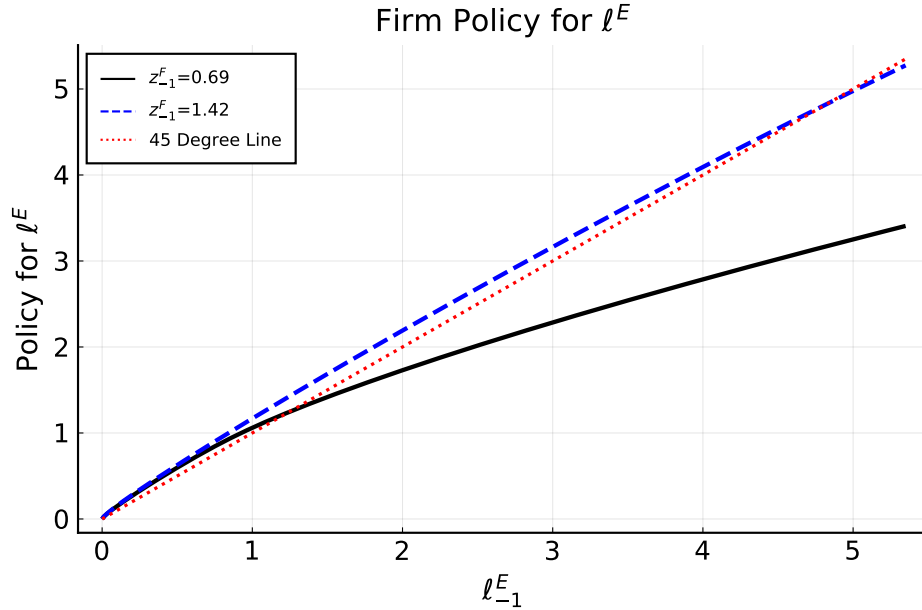


These figures show the stationary distribution of firms across TFP shocks and stocks of traditional employment. The left panel shows slices of this distribution for the lowest (solid black line) and highest (dashed blue line) shocks.

To understand this difference in the dispersion, consider the policy function of firms in different situations (depicted in Figure fig: Firm Policy Functions). A firm with a stock

of traditional employees $\ell_{-1}^E = 3$ that gets the $z_{low}^F = 0.69$ shock is over-employed since at $\ell_{-1}^E = 3$ its policy function (the solid black line in Figure 8) is below the 45-degree line. If the firm continued to receive the z_{low}^F shock, it would eventually converge to $\ell_{low}^{E*} = 1.2$ (where the black line intersects the 45-degree line). Although adjusting traditional employment is costly, the z_{low}^F policy function shows that even firms that start out with a high level of traditional employment will end up close to ℓ_{low}^{E*} after just a few periods with the low shock. Consequently, the z_{low}^F slice of the distribution in Figure 7 is fairly concentrated near ℓ_{low}^{E*} .

Figure 8: Firm Policy Function for Traditional Employment



This figure shows the firm policy function for how many traditional employees to hire based on their previous traditional employment level ℓ_{-1}^E if they had the lowest (solid black line) or highest (dashed blue line) TFP shock in the previous period. The dotted red line is the 45-degree line.

Now, consider another firm with $\ell_{-1}^E = 3$ but who receives the $z_{high}^F = 1.42$ shock. They are underemployed as the steady state employment level with the high shock is $\ell_{high}^{E*} = 4.8$ (where the dashed blue line intersects the 45-degree line in Figure 8). Since shocks are persistent, it is likely that they will receive z_{high}^F against next period, so the firm will adjust toward ℓ_{high}^{E*} .

However, the firm can circumvent the convex costs of rapidly adjusting ℓ^E upward by hiring contingent workers. As a result, their traditional employee level adjusts more gradually than for firms with low productivity shocks and they can spread the adjustment cost over many periods. As a result, the distribution of firms with the high z^F shocks is much more dispersed across traditional employment levels.

6.3 Policy Analysis: Extending UI to Contingent Workers

In the baseline policy, which has been used for the results thus far in this paper, UI replaces 40 percent of wage income for traditional employees after they lose their job, but contingent workers are not eligible for UI. This represents the prevailing policy rules in the U.S. from 1935 through early 2020; similarly, UI programs in many other developed countries also exclude self-employed and contingent workers. In addition, I set the UI expiration probability in the model to $\delta^U = 0.5$ since most states provide benefits for up to 26 weeks through their regularly-funded UI programs. The alternative policy that I consider here extends the same 40 percent replacement rate to contingent workers who become unemployed and has the same expiration rate for both types of workers. The results of the policy experiment are summarized in Table 5.

Comparing the results in columns 1 and 2 in Table 5, we see that extending UI benefits to contingent workers generates welfare losses of 2.8 percent in consumption-equivalent units. The main reason for these losses is that the UI benefits discourages individuals from looking for work, and so the UI bill decreases and total employment decreases. In addition, the share of the labor force in contingent work increases from 14.3 to 25.2 percent under the extended UI policy. Since fewer labor hours are supplied and a greater share of these hours are provided by contingent workers (who are less productive for firms by factor λ), GDP and ag-

Table 5: Results under Baseline and Extended UI Policies

	Baseline Policy	Extended Policy
Traditional employee UI replacement rate b^E	0.4	0.4
Contingent worker UI replacement rate b^C	0.0	0.4
Benefit expiration rate δ^U	0.5	0.5
Welfare, relative to baseline	-	-2.8%
GDP, relative to baseline	-	-3.0%
Wage gap $\frac{w^C}{w^E}$	0.87	0.86
Tax rate τ	2.6%	6.6%
Contingent share of workforce	14.3%	25.2%
Effective labor hours supplied, relative to baseline	-	-2.9%
UI bill relative to GDP	2.7%	7.0%

gregate consumption decrease by 3.0 percent. Consequently, although extending UI benefits to contingent workers helps more individuals smooth consumption through unemployment spells, the level decrease in consumption dominates and results in welfare losses.

7 Conclusion

In this paper, I studied the implications of extending UI benefits to contingent workers, while considering the incentives that drive individuals' choices between traditional employment and contingent work and firms' hiring decisions. I documented greater dispersion in hours worked by contingent workers than by traditional employees. I also documented larger changes in the weekly hours worked by contingent workers. In addition, I found that on average contingent workers' annual income is lower by 33 percent, their hourly wages are lower by 11 percent, and their job spells are 11 weeks shorter relative to those of traditional employees.

I developed a structural model where individuals choose to be either contingent workers (so they can flexibly choose hours) or traditional employees (to receive a higher hourly wage). Firms choose how many hours of each labor type to hire. When making this decision,

firms take into account 1) the marginal products of traditional employees versus contingent workers relative to their wages, 2) the risk involved in choosing traditional employees before observing their TFP shock and then hiring less productive contingent workers afterward if needed, and 3) administrative costs incurred for adjusting traditional employment from one period to the next. Based on the current calibration, the model generates a similar but slightly higher contingent share of the workforce and wage gap than I observe in the data. Using this model, I found that extending unemployment insurance to contingent workers would generate welfare losses of 2.8 percent and decrease GDP by 3.0 percent. These results suggest that it would not be beneficial overall to permanently extend the same UI benefits to contingent workers as is currently available for traditional employees. However, further research is required to examine whether providing partial UI to contingent workers would have similar effects in the economy.

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