Business Income Dynamics and Labor Market Fluidity*

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July 14, 2020

Abstract

Studies on the decline in U.S. labor market fluidity have focused on employees and excluded most of the rapidly growing population of the self-employed and business owners. In this paper, we use administrative records data to construct labor reallocation rates that include business income recipients. In our more inclusive definitions, declines in annual hire and separation rates from 1994 to 2014 were smaller by 1.3 to 1.4 percentage points (8.3% to 8.7%), mainly among jobs that were secondary sources of income or short in duration. We present evidence that worker transitions between wage and salary jobs and business income represent labor reallocation, as opposed to reclassification of employees as independent contractors.

^{*} We would like to thank Katharine Abraham, Andrew Foote, John Haltiwanger, William Kerr, Erika McEntarfer, Jesse Rothstein, James Spletzer, and participants at the U.S. Census Bureau research lunch, the 2018 American Economic Association conference, the 2018 IRS-Census workshop, the NBER Summer Institute 2018 CRIW Workshop, and the 2019 Comparative Analysis of Enterprise Data Conference for helpful comments and suggestions, and Claire Hou for assistance. Opinions and conclusions are those of the authors alone and do not necessarily represent the views of the U.S. Census Bureau. All results have been reviewed to ensure that no confidential data are disclosed (CBDRB-FY18-093, DRB-B0101-CDAR-20180628, DRB-B0046-CED-20190425, DRB-B0050-CED-20190509, CBDRB-FY20-CED006-0006, and CBDRB-FY20-CED006-0007). Early drafts of this paper were circulated under the title "Aggregate Labor Market Fluidity." JEL codes: J63, L26. Keywords: labor reallocation, employment dynamics, self-employment, business ownership.

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1 Introduction

Numerous data sources indicate that labor market fluidity in the U.S. has declined in recent decades, especially since 2000.¹ This evidence has received considerable interest because labor market fluidity is a key mechanism for efficiency-enhancing reallocation; and, if the pace of labor reallocation has slowed, then productivity growth may be lower as a result. One common feature of most available data on labor market fluidity is the use of employer-reported wage and salary records either through direct tabulations (e.g., BDS or QWI), or as the data frame (e.g., JOLTS). An implication of this reliance on employer-reported wage and salary records is that nearly all published series on labor reallocation exclude most of the self-employed and business owners.²

In this paper, we use administrative records data for the U.S. to demonstrate that the population of business income recipients matters for labor market fluidity. The share of the population receiving business income in administrative records has risen in recent decades such that nearly one-in-five income recipients in the U.S. received business income in 2014. We document that most business income recipients do not receive a wage or salary from their owned business; and thus, most business income recipients are omitted from standard measures of labor reallocation (which focus on wage and salary employees). Given this omission, we examine the relationship between recent changes in labor reallocation rates and contemporaneous increases in the share of the population that receives business income. We show that rates of entry and exit from business income receipt have been relatively constant in recent decades — which is in sharp contrast to the substantial decline in the movement into and out of wage and salary employment. Due to these differing trends, including business income recipients in measures of labor market fluidity reduces the measured decline in the annual hire (separation) rate by 8.3% (8.7%) between 1994 and 2014. We also consider the relationship between business income receipt and a worker's portfolio of tasks. When workers receive business income in a calendar year, they work in more industries.

¹These sources include the Business Dynamics Statistics (BDS), the Business Employment Dynamics (BED), the Current Population Survey (CPS), the Job Openings and Labor Turnover Survey (JOLTS), the Quarterly Workforce Indicators (QWI), and Job-to-Job Flows (J2J). See, among many others, Hyatt and Spletzer (2013), Davis and Halti-wanger (2014), and Molloy et al. (2016).

²Our data identify individuals receiving business income — which includes both business owners (irrespective of whether they contribute labor to the business) and the self-employed. As a result, we will refer to business owners and the self-employed collectively as "business income recipients." Some of our results distinguish between sole proprietors, partnerships, and S corporations, but this distinction is at best a crude approximation of this conceptual difference.

Our paper uses two datasets to examine the implications of including business income recipients in measures of labor market reallocation. The first dataset consists of survey records from the Annual Social and Economic (ASEC) Supplement to the Current Population Survey (CPS) linked to W-2 records on wage and salary earnings, as well as business income from Schedule SE filings (required for sole proprietors and owners of partnerships earning more than \$400 in self-employment income). This first dataset provides information from 1990 to 2016. The second dataset consists of universe-level income records, including W-2 records from 2005 onward, and person-level business income records from 2007 onward for three different legal forms of organization: sole proprietors, partnerships, and S corporations.³ Our universe-level dataset provides information through 2015.

We begin by using these two datasets to document the rise of self-employment in the U.S. in recent decades. We find that the share of the income recipients who receive business income rose 15% between 1991 and 2016. Our finding is consistent with the earlier work by Abraham et al. (2018) on the rising share of the population with business income. It also offers a counterpoint to studies such as Decker et al. (2014), Sedláček and Sterk (2017), and Pugsley and Şahin (2018) that emphasize the decline in the formation rate of new employer businesses in recent decades. The decline in the employer business entry rate has been small relative to the sustained growth in the number of nonemployer business entities.

We also assess whether business income recipients are counted as employees in administrative wage and salary records. We show that less than one-tenth of the owners of sole proprietors and partnerships received W-2 earnings, which is unsurprising given that in most cases the U.S. Internal Revenue Service (2017, 2019a) has discouraged or prohibited them from doing so. What is perhaps more surprising is that less than half of the owners of S corporations reported W-2 earnings at the firms they own. Because most published series on labor reallocation only consider the employees of businesses, this absence of W-2 earnings for most business owners suggests that the overwhelming majority of transitions into and out of business income receipt are omitted from measured labor reallocation rates. While this fact is important as motivation for our remaining analysis, it also provides a necessary correction to a small but rapidly growing strand of the lit-

³Ownership information for C corporations is much more limited and the extent to which owners are able to be identified is a subject of ongoing research, see Bayard et al. (2018). Ownership data on C corporations was not available for this analysis.

erature on entrepreneurship that, starting with Agarwal et al. (2016) and Kerr and Kerr (2017), assumes that business owners are commonly present in U.S. matched employer-employee data. Our empirical findings challenge this assumption.

The next part of our paper assesses the implications of excluding the large and growing share of the U.S. population with business income from labor reallocation rates. We first compute the rates of entry into and exit from business income receipt. We find differences in entry and exit rates by legal form of organization: sole proprietors have higher transition rates than owners of partnerships and S corporations. We also compare business income transition rates to employee hire and separation rates, and we find several differences. First, transition rates into and out of business income receipt have been lower than hire and separation rates, implying that self-employment provides greater stability than the typical wage and salary job. Second, entry and exit from business income receipt has been less cyclical than hire and separation rates of wage and salary employees. Third, there is no trend decline in transition rates into and out of business income receipt, in contrast to the dramatic declines in hire and separation rates of wage and salary workers. These facts imply that incorporating business income receipt into labor reallocation measures should lead to smaller measured declines in recent decades.

We then incorporate business income dynamics into our labor reallocation measures. The inclusion of business income has a substantial impact on hire and separation rates, reducing the measured declines by 1.3 and 1.4 percentage points respectively (accounting for 8.3% and 8.7% of the observed declines measured using only wage and salary earnings). We find much smaller offsetting effects for employer-to-employer transitions (0.2 percentage points), employment-tononemployment transitions (0.1 percentage points), and nonemployment-to-employment transitions (0.1 percentage points).

Most of the offsetting effect of including business income recipients in hire and separation rates can be attributed to jobs that are secondary sources of income, or short in duration. Among such jobs, inclusion of business income recipients offsets 1.1 and 1.2 percentage points of the measured decline in the incidence of hires and separations, respectively. Given that these secondary and short-duration jobs had the most dramatic measured declines across the various measures of labor reallocation, this finding suggests that the increase in the population receiving business income has played a role in the long-run decline in short-duration wage and salary jobs. The few existing studies on the declining incidence of short-duration jobs by Hyatt and Spletzer (2017), Hall and Schulhofer-Wohl (2018), and Pries and Rogerson (2019) do not consider this margin of adjustment.

Because short-duration and secondary wage and salary jobs may have shifted to "gig" employment, we assess the extent to which business income receipt reduces the number of wage and salary jobs held by the recipient. The degree to which business income receipt displaces wage and salary jobs depends on whether the business income is the dominant source of income for the individual. When individuals earned more from self-employment than from wage and salary employment, then business income displaced wage and salary jobs about one-for-one. If instead the business income was a secondary source of income (as in the case with most "gig" workers), then receiving business income had a much smaller effect on the number of wage and salary jobs — reducing the average number of wage and salary jobs held in a year by only one-tenth of a job. Furthermore, we find that this displacement of wage and salary jobs by secondary business income receipt has weakened over time.

Finally, we explore the extent to which growing self-employment represents reclassification of tasks from within a business entity to external self-employed contractors. To do so, we examine how the number of industries adjusts as an individual moves into and out of business income receipt. When business income provided the secondary source of income, the average number of industries worked rose by 0.392. If, instead, the business income was the individual's dominant source of income, then the average number of industries worked rose by 0.508. These results suggest that business income receipients tend to work in new industries rather than the same industries from which they receive wage and salary payments. The rise in business income receipt has therefore led to true increases in labor reallocation rather than mere reclassification of wage and salary jobs to self-employment. We conclude that rising share of the U.S. workforce that receives business income is important for understanding the declines in labor reallocation rates considered by Hyatt and Spletzer (2013), Davis and Haltiwanger (2014), and many others.

2 Data

We measure the relationship between business income and labor market fluidity using two administrative records datasets. First, CPS records enhanced with longitudinal administrative records provide information on years 1990 to 2016. These administrative records data cover wage and salary jobs as well as self-employment spells for sole proprietors and partners. Our second data source consists of universe-level administrative records. W-2 wage and salary job records provide information starting in 2005. Administrative business income records provide information on all sole proprietors, partnerships, and S corporations (but excluding C corporations) starting in 2007. Using the survey-based dataset, we obtain a much longer time series. The universe-level data allows us to consider a broader population of owners, as well as to distinguish between sole proprietors, partnerships, and S corporations. We describe each of our two datasets below.

2.1 CPS-DER

Our first dataset is the Annual Social and Economic (ASEC) Supplement to the Current Population Survey (CPS) linked with administrative records on wage and salary earnings, as well as business income. In each year, sole proprietors and partners who receive (from all combined ownership activity) net annual earnings above \$400 are required to file a Schedule SE (Self-Employment Tax). The Master Earnings File (MEF) database maintained by the Social Security Administration contains information on business income from Schedule SE filings as well as information on wage and salary earnings from each Form W-2 a person has received during the year. Our analysis uses an extract (called the Detailed Earnings Record, or DER) that includes MEF records for each CPS respondent for whom a Protected Identification Key (PIK), a person-level longitudinal identifier, is available. Note that the CPS-DER lacks identifiers for the businesses that Schedule SE filers own. Therefore, the only distinction we can make is whether or not an individual receives business income, not the dynamics associated with the particular businesses that they own.

By using the sample weights from the CPS ASEC in combination with DER administrative information for the same year, our dataset can be used to estimate population-level characteristics of Schedule SE recipients as well as W-2 jobholders in each year. It is worth noting that this linked dataset provides information (including weights) for only those CPS respondents who have received a PIK and are therefore linkable with administrative records data sources. Because not all respondents receive a PIK, we create sample weights that adjust for the likelihood of receiving a PIK.⁴ Although the DER includes the entire W-2 and Schedule SE history of CPS respondents

⁴We use the base weights for the CPS multiplied by the inverse of the estimated probability of receiving a PIK from

(back to 1978), we only select jobs observed in the DER in the CPS reference year. This selection ensures that the survey-weighted sample is representative of the U.S. population. Because we produce statistics on entry and exit from wage and salary work and business ownership, for each wage and salary job and self-employment spell observed in the DER in the relevant year, we also include information on that job or spell from the preceding and following year. We use CPS ASEC responses from 1991 to 2016, linked with administrative records from 1990 through 2016.

2.2 Universe-level W-2, Schedule C, and Schedule K-1 records

We construct our second dataset from universe-level sources of wage and salary earnings and business income for several legal forms of organization. We use the universe of W-2 records for wage and salary jobs starting in 2005. We obtain sole proprietor ownership spells from the U.S. Census Bureau non-employer and employer business registers. The universe of Schedule K-1 data provides information on partnership and S corporation ownership spells.⁵ Among sole proprietors, we have access to information on the owner (from 1040 Schedule C) from 2007 through 2015. For pass-through entities covered by Schedule K-1, each entity may have multiple owners. We have access to information on these entities linked to their owners for years 2007 through 2015.

Our universe-level dataset offers several advantages over the CPS-DER. First, it is universelevel data rather than a sample of the population. Second, it includes S corporations and distinguishes between sources of business income (by legal entity). Third, by obtaining ownership information directly from tax reports, we gain a wealth of information on the entities they own and operate. If an individual has multiple businesses, each with a distinct Employer Identification Number (EIN), then ownership is considered separately for each business. This contrasts with the CPS-DER, which pools together all sole proprietor and partnership earnings for the person in a calendar year and so does not allow this distinction. Fourth, we obtain information on sole proprietors and owners of partnerships who did not file a Schedule SE. Finally, the business identifiers in our universe-level data, along with the universal coverage of wage and salary payments for

a linear probability model. Observable characteristics used to estimate this probability are age categories, education categories, gender, race group indicators (black alone, white alone, and any other race), marital status, a foreign born indicator, and indicators for state of residence.

⁵See Garcia-Perez et al. (2013) for further description of sole proprietor linkages. See Goldschlag, Kim, and McCue (2017) for further descriptions of the Schedule K-1 data and linkages.

these businesses, allow us to assess the extent to which business owners receive wage and salary payments.

2.3 Measuring labor market transitions

We construct individual-level annual measures of employment, hires, and separations using the concepts developed by Abowd et al. (2009).⁶ If an individual has no earnings from an employer in one year but positive earnings in the next, we infer a hire in the second year. Similarly, if an individual has positive earnings from an employer in one year, but has no earnings from the employer in the next, we infer a separation in the first year.

Following Hyatt et al. (2014), we also construct annual measures of worker transitions into and out of employment, as well as transitions between dominant employers from one year to the next. We identify a worker's dominant employer at the start of a given year as the employer that: (1) paid the worker positive earnings in both the current and previous year (indicating that the worker was employed at the start of the current year), and (2) paid the highest cumulative earnings, over those two years, among the set of firms that employed the worker at the start of the current year. If an individual's dominant start-of-year employer changes from one year to the next, then we infer an employer-to-employer transition. If a person is employed at the beginning of a year but not at its end, we infer a employment-to-nonemployment transition. Finally, if a person is employed at the end of a year but not at its beginning, we infer a nonemployment-to-employment transition.

3 Prevalence of business income receipt in the U.S.

To measure the prevalence of business income receipt, we categorize workers into one of four types of employment based on their receipt of wage and salary income and/or business income in a given year. Our first category of income recipients consists of workers who only have wage and salary income. Our second category consists of individuals who only receive business income. Our third and fourth categories consist of those individuals who receive both business income and wage and salary income — these two categories distinguish whether business income exceeds wage and

⁶See Appendix A for formal definitions.

Figure 1: Shares of the population with income in a calendar year, by income source (a) CPS-DER W-2 and Schedule SE



Notes: Authors' calculations of the CPS-DER W-2 and Schedule SE records. , as well as universe-level W-2, Schedule C, and Schedule K-1 records. Schedule C covers sole proprietors, and K-1 covers partnership and S corporation income. Wage and salary income comes from W-2 records. See text for additional details.

salary income. Our calculations here exclude the nonemployed, so these four shares sum to 100% in each year and are exhaustive of the population with positive wage and salary or business income.

We measure the trends in business income receipt over time. Figure 1(a) shows that the share of individuals receiving any business income in the CPS-DER records rose 15% between 1991 and 2016. The share with only wage and salary income fell from 90.1% in 1991 to 88.5% in 2016, which corresponded with an increase in the share with any business income from 9.9% to 11.5%.

A larger fraction of the population received business income in the universe-level data than in the CPS-DER. As shown in Figure 1(b), according to the universe-level W-2 and business income records, 19.6% of income recipients received some business income in 2014. Furthermore, business income was the dominant source of income for 12.7% of income recipients in 2014 (and the sole source of income for 10.9%). In the CPS-DER, (see Figure 1(a)), a smaller share (11.5%) of income recipients had business income in 2014. The higher share of the population with positive business income in the universe-level data is likely due to the fact that our CPS-DER sample only includes sole proprietors and owners of partnerships who file a Schedule SE. The CPS-DER thus excludes owners of S corporations as well as sole proprietors and owners of partnerships who earn less than \$400 in business income.

Many business income recipients also receive wage and salary income. Figure 1(a) shows that, in the CPS-DER, the share of these individuals with secondary business ownership income increased from 2.7% in 1991 to 3.5% in 2016. The share of the population with secondary wage and salary income grew from 1.1% in 1991 to 1.5% in 2016. Many new business income recipients rely on this income exclusively: the share of individuals with only business income grew from 6.0% of all income recipients in 1991 to 6.5% in 2016.

As shown in Figure 1(b), the universe-level administrative records data indicate that the share of individuals with only wage and salary income declined from 81.3% in 2007 to 80.4% in 2014. The ordinal ranking of the shares of business income categories was similar in both the CPS-DER sample and the universe-level sample. In the universe-level administrative records data, 9.5% of income recipients had only business income in 2007, while 10.9% did in 2014. The share of income recipients receiving more wage and salary income than business income declined slightly between 2007 to 2014 (falling from 7.2% in 2007 to 6.7% in 2014). In the universe-level administrative records data, the share of income recipients with business income and secondary wage and salary income grew from 1.6% in 2007 to 1.8% in 2014.

These seemingly modest changes of a couple of percentage points in the share of the population that receives business income reflect important changes in employment and the number of business entities in the U.S. in recent decades. Each percentage point increase in the share of this population corresponds with more than one million new business income recipients, and roughly as many new business entities.⁷ This finding has important implications for studies that emphasize the decline in the employer business entry rate — which almost exclusively rely on data from the BDS (e.g. Decker et al. (2014), Sedláček and Sterk (2017), and Pugsley and Şahin (2018)). Data from the U.S. Census Bureau (2006, 2019) indicates that the total number of nonemployer business entities increased from 19.5 million in 2004 to 24.8 million in 2016, a proportional increase of about 24%. BDS aggregates for those same years indicate slower growth the number of employer business entities: from 5.0 million to 5.2 million, a much more modest change of only about 4%. A complete picture of total employment and business dynamism in the U.S. needs to consider both employer and nonemployer businesses.

4 Do business owners receive wage and salary payments?

We now examine how often business owners receive wage and salary payments from the businesses they own. To do so, we merge our universe-level business ownership data with the W-2 wage and salary records based on the EIN of the employer and the PIK of the owner. A match on EIN and PIK indicates that, in addition to the business income, the owner also receives wage and salary payments from the owned business.

As shown in Table 1, relatively few business owners received wage and salary income from their businesses.⁸ Only 7.4% of sole proprietors and 1.1% of the owners of partnerships received wage and salary income from their businesses. The owners of S corporations were more frequently employees of the businesses that they own, with 40.4% of the owners of S corporations receiving wage and salary income from their owned business. These results indicate that most business owners are not present in employer-reported wage and salary records — and thus business income recipients are generally omitted from labor reallocation statistics.

These results also have implications for the literature that, starting with Agarwal et al. (2016) and Kerr and Kerr (2017), attempts to identify business founders, owners, and entrepreneurs using

⁷To see this, note that the employed population in the U.S. was 157 million in June of 2019, according to the Bureau of Labor Statistics (2019). We also explore this further in Appendix Figure B1, which presents totals for the data sources we consider. The number of individuals with business income in the CPS-DER increased from 13.0 million in 1996 to 19.4 million in 2016. In the universe-level administrative records data, it increased from 16.7 million in 2008 to 18.9 million in 2015.

⁸The sole proprietor businesses that do not have EINs (which are a majority of sole proprietors) are beyond the scope of Table 1.

	Schedule C	Form K-1	Form K-1
	Sole Proprietors	Partnerships	S Corps.
Share of owners:			
Owners of EIN firms in W-2	7.4%	1.1%	40.4%
Owners of employer firms in W-2	12.5%	11.2%	70.3%
Owners of employer firms, top 3	10.3%	7.9%	59.7%
Share of firms:			
Any owner of EIN firms in W-2	7.5%	3.8%	48.8%
Any owners of employer firms in W-2	13.2%	24.3%	84.8%
Any owners of employer firms, top 3	11.0%	20.2%	78.5%

Table 1: Share of business owners who receive wage and salary payments

Notes: Authors' calculations of universe-level W-2, Schedule C, and Schedule K-1 records. Percentages indicate the share of businesses owners who receive a W-2 record of the wage and salary income from a business of a particular legal form that they own, as well as the share of businesses that have at least one owner with a W-2 record. "Top 3" indicates that the owner is among the top 3 highest earners at that employer.

matched employer-employee wage and salary records. This approach labels an individual as a business founder if the individual was one of the three highest earning employees in the year that the business hired its first employee. The popularity of this approach is somewhat surprising given that the U.S. Internal Revenue Service (2017, 2019a) discourages and in many cases prohibits sole proprietors and the owners of partnerships from paying themselves a salary.⁹ Given this legal prohibition, this method is unlikely to successfully identify the owners of sole proprietor and partnership businesses. Conversely, owners of S corporations who provide labor services to their firms are "officers" who are legally required to receive wage and salary payments.¹⁰

Our linked datasets allow us to assess the accuracy of labeling high earning employees as business founders. We first measure the frequency with which owners of employer businesses receive wage and salary income from their owned business (and thus appear as employees in matched employer-employee linked datasets). The second row in Table 1 shows that only 12.5% of employer sole proprietors, and 11.2% of the owners of employer partnerships received wage and salary payments. A higher fraction of the owners of employer S corporations, 70.3%, received

⁹The Tax Cuts and Jobs Act of 2017 changed the tax treatment of sole proprietor businesses. As a result, since 2018 sole proprietors have been able to deduct some wage and salary payments to owners.

¹⁰According to the U.S. Internal Revenue Service (2019b), "an officer of a corporation is generally an employee; however, an officer who performs no services or only minor services, and neither receives nor is entitled to receive any pay, isn't considered an employee." In practice, the U.S. Internal Revenue Service (2019c) asks that only businesses that have \$500,000 or more in revenue report officer compensation directly. This reporting requirement may also influence the extent to which owners receive wage and salary payments.

wage and salary payments. The fifth row of the table provides the share of firms in which at least one owner is present in the W-2 data. 13.2% of employer sole proprietors and 24.3% of the owners of partnerships have at least one owner present in the W-2 data. In 84.8% of employer S corporations, at least one owner received wage and salary payments.

Next, we measure the accuracy of the the entrepreneur identification strategy of Kerr and Kerr (2017) by focusing on the top three earners at employer businesses.¹¹ The third row of Table 1 shows that the share of owners who were among the top three wage and salary earners was 10.3% for sole proprietors, 7.9% for partnerships, and 59.7% for S corporations.¹² The sixth row of the table shows that the top three highest earning employees include at least one owner in 11.0% of sole proprietors, 20.2% of partnerships, and 78.5% of S corporations.

These results suggest that researchers should not assume that the owner of a particular business is present in U.S. matched employer-employee data. In the case of sole proprietor and partnership employer businesses, a much better approximation of the data can be obtained by assuming that no owner is present.¹³ The top three earners of an S corporation include at least one owner nearly four-out-of-five times, but most of the top three wage and salary earners at S corporations are not owners. Thus, even for S corporations these approaches yield at best a proxy for ownership.

5 Incorporating business owners into labor reallocation

We now assess the implications of excluding business income recipients from measures of labor market fluidity. To do so, we first compute standard measures of labor reallocation using wage and salary payments alone. We then re-compute these labor reallocation measures treating the receipt of business income as though it were a wage and salary job. To the extent that business income receipt captures labor conducted by the recipient on behalf of the owned business, this allows us to

¹¹In Appendix Table C1, we present results derived from wage records collected by U.S. states as part of unemployment insurance payroll taxation, and results are quite similar to Table 1. In Appendix Table C2, we incorporate commonly used sample selection techniques: including only businesses that start as single establishment firms, and requiring that owners receive wage and salary payments in the year the business hires its first employee. Results are broadly similar for sole proprietor and partnership businesses. For S corporations, these sample selection techniques identify fewer owners.

¹²Azoulay et al. (2020) report that 90% of S corporation owners who are also employees of their firms are among the top three highest earning employees. Table 1 implies something comparable: 59.7/70.3 = 84.9%.

¹³While most studies that identify business founders or entrepreneurs implement their method without distinguishing between different legal forms of organization, a recent exception is Azoulay et al. (2020).

measure the effect of using more inclusive definitions of labor reallocation.¹⁴

5.1 Entry and exit from business income receipt

We measure transition rates into and out of business income receipt, and compare these with hire and separation rates for wage and salary employees. Results are shown in Figure 2. Figure 2(a) shows three facts which indicate that incorporating business income receipt into our measures of hire and separation rates should result in series that are lower in magnitude, exhibit less cyclicality, and have less of a trend decline. First, prior to the 2007-2009 recession, the wage and salary hire and separation rates were consistently higher than business income entry and exit rates, and this difference peaked at 24.8 percentage points immediately prior to the 2001 recession. Second, business income entry and exit rates were less cyclical than wage and salary hire and separation rates. Third, business income entry and exit rates did not show evidence of trend declines. The entry rate was virtually unchanged from 45.9% in 1991 to 44.6% in 2016. The exit rate increased slightly from 42.9% in 1990 to 46.7% in 2015.

Using universe-level business income records, we find that the rate of business income entry and exit differed substantially by legal form of organization. Figure 2(b) shows that sole proprietor businesses had entry and exit rates between 38% and 52%, whereas the business income entry and exit rates for partnerships and S corporations were about two to four times lower (ranging from 9% to 19%).¹⁵ These findings confirm that including business income dynamics should lower the magnitude and cyclicality of labor reallocation rates.

¹⁴This is the antecedent upon which the interpretation of our exercises in this Section depends. If many business income recipients are performing labor, then it makes sense to include them in a labor reallocation measure. If they rather take no active role in the business other than providing investment or capital, then it makes less sense to include them. It is likely that the income of sole proprietors is more indicative of labor than that of the owners of partnerships or S corporations.

¹⁵Goetz et al. (2017) present entry and exit rates for sole proprietors in 2009. The entry and exit rates for business income for sole proprietors are broadly consistent with, but somewhat lower than, the entry and exit rates calculated for Schedule SE filers from the CPS-linked sample. Only sole proprietors earning more than \$400 from the business are required to file a Schedule SE.



Figure 2: Hire and separation rates vs. business income entry and exit rates (a) CPS-DER W-2 and Schedule C

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 records. Entry indicates zero business income in the previous year and positive business income in the current year. Exit indicates positive business income in the current year but zero business income in the subsequent year. Denominators for each rate are the average of total number of owners of the given entity type at the beginning and end of the year.

5.2 Hire and separation rates

We now include business income recipients when we calculate hire and separation rates. For comparison, we also show hire and separation rates using W-2 data alone.¹⁶ Figure 3(a) shows results using the CPS-DER. Because was business income recipients constituted only 6% to 9%

¹⁶Additional evidence on the hire and separation rates from W-2 data is shown in Appendix Figure E1 and the associated discussion. Note that the W-2 hire and separation rates that we report in Figure 3 range from 41% to 69%. These hire and separation rates are much higher than most published series because our series are calculated on an annual basis, whereas most other series are published on a monthly or quarterly basis. In Appendix Figure E2, we compare the annual hire and separation rates with other available data by summing monthly or quarterly rates for any given year.

of consecutive year employment in the CPS-linked records, the more inclusive hire and separation series mimic those created using only W-2 records of wage and salary income.¹⁷

Because business income entry and exit rates were lower than wage and salary hire and separation rates up until the 2007-2009 recession, the more inclusive hire and separation rates tended to be lower than those that relied on the W-2 wage and salary records only. The differences between these hire and separation rates were largest in the 1990s, peaking in 2000 at 1.8 percentage points. As the wage and salary hire and separation rates fell during the 2001 and the 2007-2009 recessions, the relative acyclicality of the business income entry and exit rates caused the gap between the combined and W-2 only hire and separation rates to shrink — such that from 2008 onward the difference never exceeded 0.6 percentage points.

Including business income receipt in the calculation of hire and separation rates also reduces the measured decline from 1990 to 2016, since the large gap observed in the 1990s between the business income entry and exit rates versus the W-2 hire and separation rates had largely disappeared by 2009. This dampening effect was largest for the separation rate — with the separation rate calculated using both business income and wage and salary income falling 1.2 percentage points less from 1990 to 2015 relative to the separation rate calculated using only wage and salary records (the combined separation rate fell 13.0 percentage points, whereas the wage and salary separation rate fell 14.2 percentage points over this period). The combined hire rate, on the other hand, fell by 5.3 percentage points from 1991 to 2016, which was 0.2 percentage points less than the 5.5 percentage point decline in the hire rate measured using only wage and salary income.

The universe-level dataset shows that including business income entrants and exits in hire and separation rates results in both a more substantial lowering of the hire and separation rates and smaller cyclical fluctuations in these labor reallocation measures. Hire rates were 2.7 to 6.5 percentage points lower, and separation rates 0.5 to 5.3 percentage points lower, when including business income recipients from the universe-level records.

We can compare these labor reallocation series starting in 2008 to assess how including business income records from the universe-level dataset (which include a broader set of business income recipients) affects the measured decline in hires and separations during and after the 2007-

¹⁷For a time series of the business income share of consecutive year employment, as well as its share of dominant (maximal income) consecutive year employment, see Appendix Figure D1.



Figure 3: Hires and separations

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 records. See text for additional details.

2009 recession. According to the universe-level records, including business income entrants results in a 17% smaller decline in the hire rate from 2008 to 2010, whereas the CPS-linked records indicate that including business income recipients only dampens the fall in the hire rate by 5.5%. The universe-level and CPS-DER records largely agree on the effect of including business income on the separation rate. In the CPS-DER, including business income exits generates a 4.5% smaller decline in the combined separation rate from 2008 to 2010 relative to the wage and salary separation rate. In the universe-level records, including business income exits in the separation rate results in a 6.5% smaller cyclical decline in separations from 2008 to 2010 relative to the separation rate measured using only wage and salary records.

5.3 Employer-to-employer and nonemployment transitions

In Figure 4, we consider how including business income can affect the employer-to-employer transition rate, as well as the rate at which workers transition into and out of nonemployment. We find that including business income has little effect on these reallocation rates.¹⁸

The employer-to-employer transition rate evolved in a basically identical manner whether we include or exclude the self-employed, as shown in Figure 4(a). In the CPS-DER, differences between the employer-to-employer transition rate using W-2 data alone relative to that which includes business income recipients were at most 0.2 percentage points. As a result, the long-term evolution of the employer-to-employer transition rates including and excluding business income recipients are also virtually identical. Both series increase by about 2.0% from 1991 to 2015.

Figure 4(a) shows more substantial differences in the employer-to-employer transition rate using universe-level administrative records, but the range of years available is much smaller. The employer-to-employer transition rate was 0.4 to 1.2 percentage points higher for the population of wage and salary employees and business income recipients. The employer-to-employer transition rate had a smaller decline during the 2007-2009 recession using universe-level administrative records. According to both series, there was a sharp decline of about 2.4 to 2.5 percentage points from 2008 to 2009. The W-2 data alone reached a minimum of 12.0% in both 2009 and 2010, before rising to 12.9% in 2011. The employer-to-employer transition rate including business income recipients increased from 12.4% in 2009 to 13.1% in 2010, and then to 13.7% in 2011. The recovery in the employer-to-employer transition rate was therefore more U-shaped in the W-2 data alone, and more V-shaped when including business income recipients. This albeit limited evidence suggests that the omission of business income recipients may affect the slow measured labor market recovery to the 2007-2009 recession in employer-reported administrative records data. More generally, more research on the role of self-employment and business ownership in labor market recoveries is needed.

Results for employment-to-nonemployment transitions are shown in Figure 4(b). Again, the

¹⁸Additional detail on our employer-to-employer and transition series is shown in Appendix Figure E3 and the associated discussion. In Appendix Figure E4, we benchmark these transition rates against the similar estimates from the CPS, JOLTS, and J2J. The CPS ASEC also asks how many non-overlapping employers a worker had in the previous year, see Molloy et al. (2016) and Hyatt et al. (2018). In Appendix Figure E5, we compare the CPS-DER W-2 employer-to-employer transition rate with responses to the CPS-ASEC.



Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 records. See text for additional details.

rates are similar whether including or excluding business income recipients. In the CPS-DER, transition rates including only wage and salary employment to nonemployment were consistently

higher than those same transition rates that include business income recipients, by 0.1 to 0.5 percentage points. Such evidence is consistent with transitions between wage and salary work and self-employment being omitted from employer-reported administrative records. Using employerreported administrative records alone, such moves may appear to be into or from nonemployment. However, the universe-level administrative records data suggests that this explanation may not be the dominant factor. Omission of the self-employed necessarily excludes transitions between self-employment and nonemployment. The transition rates excluding business income recipients are sometimes lower, and sometimes higher than when they are included. Thus the universe-level administrative records data therefore suggests that there are many omitted transitions between selfemployment and nonemployment, which by definition are not captured in employer-reported wage and salary records.

Results for nonemployment-to-employment transitions, shown in Figure 4(c), tell a broadly similar story. Transition rates excluding business income recipients were almost always higher than those that include them, by as much as 0.6 percentage points (the exception is 2009, when they were lower by less than 0.1 percentage point). The relative rankings of nonemployment-to-employment transitions are somewhat mixed in the universe-level administrative records data.

5.4 Secondary and short-duration jobs

The decline in labor reallocation rates in recent decades was concentrated among jobs that provided secondary income or were short in duration. Hire and separation rates capture transitions into and out of all wage and salary jobs. The dominant employer-based measures of worker reallocation, on the other hand, capture only transitions to or from an individual's dominant employer as of the start of the year. Thus, these dominant employer-based measures of worker reallocation omit both: (1) coincident, but lower earning jobs ("secondary jobs"), and (2) jobs that start and end within the same calendar year ("short-duration jobs"). As formalized by Hyatt and Spletzer (2013), the residual between the hire and separation rates and these dominant employer-based measures are precisely these secondary and short-duration jobs.¹⁹

¹⁹In addition to Hyatt and Spletzer (2013), a few other papers have considered the role of short-duration jobs in declining labor market fluidity. Pries and Rogerson (2019) showed the importance of short-duration jobs for labor market fluidity using the Quarterly Workforce Indicators. Hyatt and Spletzer (2017) explored this using the Longitudinal Employer-Household Dynamics microdata which underlie those measures. Hall and Schulhofer-Wohl (2018)



Figure 5: Secondary and short-duration job hires and separations (a) Hires

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 records. See text for additional details.

We plot these transitions in Figure 5, both using W-2 data alone, as well as using data on business income. The results are similar to those in Figure 3. Secondary and short-duration jobs followed the "stair-step" pattern noted by Hyatt and Spletzer (2013). These jobs declined during the 2001 recession, were stable between 2002 and 2006, and then declined again during the 2007-2009 recession. Secondary and short duration jobs were the main driving force in the stair-step pattern in overall hires and separations, consistent with the findings of Hyatt and Spletzer (2017). Including business income recipients in the calculation of secondary and short-duration hires and separations lowers labor reallocation rates by nearly a percentage point in the 1990s, but by only 0.2 to 0.3 percentage points in more recent years. This differential shows that the inclusion of

explored declining short-duration jobs using data from the Current Population Survey.

business income recipients has a substantial offsetting effect on the measured decline in the hire and separation rates for secondary and short-duration jobs.

5.5 Business income dynamics and labor market fluidity

The inclusion or exclusion of business owner income affects the empirical measures that are used to assess changes in labor market fluidity and the extent of its decline. We summarize our findings in Table 2, which considers our results from the CPS-DER for the years 1994 and 2014. We choose these particular years to assess the trend decline in labor market fluidity for three reasons. First, 1991 was a recession year, and so its low transition rate is likely due to cyclical factors rather than any trend change. Second, 2015 is the last year which can be calculated given available data, and so is most likely to be subject to revision with updates to the administrative records data. Third, the increase in the transition rates from 1993 to 1994 was the largest in most of our labor reallocation time series. These large increases were also coincident with a redesign of the CPS, and so may not have been driven by economic conditions.

Because business income entry and exit rates were stable over time, and, during the 1990s much lower than wage and salary hire and separation rates, their inclusion lowers the total measured decline in the hire (separation) rate by 1.3 (1.4) percentage points. The inclusion of business income receipt therefore offsets 8.3% (8.7%) of the decline in the hire (separation) rate. The inclusion of business income receipt also offsets some of the measured declines in the employer-to-employer, employment-to-nonemployment, and employment-to-nonemployment transition transition rates, but the magnitude of these declines was smaller in magnitude (only offsetting 0.1 to 0.2 percentage points change in each measure). The residual comparing these measures with each other is secondary and short-duration jobs, which (mechanically) end up being the dominant mechanism by which rising business income receipt changes the measured declines in hire and separation rates. Inclusion of business income receipt changes the measured decline in the secondary and short-duration job hire rate by 1.1 percentage point (from 13.9% to 12.8%) and its separation rate by 1.2 percentage points (from 13.8% to 12.6%).

These findings highlight the importance of increasing business income receipt on understanding changes in labor market fluidity, both on the underlying economic phenomena as well as the

	-	-		
Reallocation	Decline	Decline in W-2		Percent
Measure	in W-2	& Schedule SE	Difference	offset
Hires	15.7	14.4	1.3	8.3%
Separations	16.1	14.7	1.4	8.7%
Employer-to-employer	0.7	0.6	0.2	23.8%
Employment-to-nonemployment	1.8	1.7	0.1	4.1%
Nonemployment-to-employment	1.3	1.1	0.2	15.6%
Secondary / short-duration hires	13.9	12.8	1.1	7.4%
Secondary / short-duration separations.	13.8	12.6	1.2	8.9%

Table 2: Measured decline in reallocation including vs. excluding self-employed: 1994 vs. 2014

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records. The "Difference" and "Percent offset" columns may not exactly correspond with the difference between the the "Decline in W-2" and "Decline in Schedule SE" columns due to rounding. See text for additional details.

measurement issues. Declines in measured labor market fluidity are concentrated among hires and separations into and from jobs that either provide relatively little income, or do not last very long. Secondary and short-duration business income receipt have become more prevalent in the U.S., while over these same decades wage and salary jobs have disappeared. The magnitudes of these changes are, however, very different. Therefore, the offsetting effect is about an order of magnitude smaller than the decline in hire and separation rates.

6 Labor reallocation vs. reclassification

The preceding section provided evidence that the inclusion or exclusion business income affects the measured decline in labor reallocation rates. In this section, we follow workers as they transition between wage and salary work and business income receipt to further refine our understanding of the economic significance of these flows.

We first explore the extent to which business income receipt displaces workers' wage and salary jobs; as well as how this has changed over time. In some cases, business income receipt may indicate that workers switched their labor activity from wage and salary jobs to self-employment, resulting in workers having fewer wage and salary jobs in years when they receive business income. Alternatively, when engaging in self-employment work, workers may tend to retain their existing jobs while expanding into new sources of earnings.

We seek to assess whether transitions between wage and salary employment and business in-

come receipt represent efficiency-enhancing labor reallocation. For example, if such a transition occurs because a company changed the legal structure of the employment agreement with a worker from being a wage and salary employee to being a self-employed contractor, then this transition does not improve the efficiency of the labor market. We therefore explore to what extent the self-employment activity takes place within the same broad industry sector as the wage and salary job(s). For example, if the employment impact of ridesharing platforms is largely to encourage wage and salary taxi drivers to switch to being self-employed drivers on the ridesharing platforms, then these changes may be thought of as within-industry reclassification rather than efficiency-enhancing labor reallocation. If, instead, business income receipt tends to diversify the industry composition of the work performed by an individual, then the self-employment activity is more likely to represent efficiency-enhancing labor reallocation.

6.1 Does business income substitute for wage and salary jobs?

To examine the degree of displacement of wage and salary jobs by business income receipt, we conduct a regression analysis. Our specifications exploit within-person, over time variation in self-employment status. Our outcome variable of interest is the number of wage and salary jobs held in a given calendar year. If movement into self-employment represents an expansion of the intensity of individuals' labor activity, then we would expect less than a one-for-one fall in wage and salary jobs.²⁰

Results are shown in Table 3. We regress the number of wage and salary jobs held by an individual on an indicator for whether the individual is self-employed in that year. We report estimation results for regressions performed both with and without worker-specific fixed effects. We consider all individuals who worked in wage salary or self-employment, and include an indicator as to whether a worker ever received self-employment income. In a typical year, workers who are

²⁰Our data do not permit us to observe workers holding more than one self-employment job in any given year. Instead, we can observe only the presence and amount of total combined self-employment earnings. Thus, it is possible that business income reported on an individual's Schedule SE may represent activities that would have generated multiple W-2s for a wage and salary worker. For example, in previous decades a particular worker might have received multiple W-2s for short stints as a taxi driver and as an office temp worker, whereas in today's "gig economy," that same worker may instead work short stints via a ridesharing platform and a clerical services platform. If this is the case, then our accounting method used in the previous section, which allows at most one Schedule SE or sole proprietor job (and hence business owner entry and exit) per person per year may understate the true amount of worker reallocation. In this case, our estimate of the offsetting effect of self-employment dynamics on labor reallocation would be a lower bound.

	CPS-DER		Univ	verse
Intercept	1.420***		1.409***	
	(0.000)		(0.000)	
Ever BI	0.074***		0.081***	
	(0.000)		(0.001)	
Ever BI* BI	-0.143***	-0.101***	-0.088***	0.013***
	(0.001)	(0.001)	(0.001)	(0.001)
Ever BI* BI*BI dom	-1.002***	-0.818***	-1.010***	-0.817***
	(0.001)	(0.001)	(0.002)	(0.001)
Worker fixed effects	Ν	Y	Ν	Y
R^2	0.096	0.446	0.151	0.652

Table 3: Number of wage and salary jobs worked regressed on business income

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 data. Regressions estimate how the number of jobs worked differ based on an individual's Business Income (BI). See text for additional details.

ever self-employed work a small fraction (0.074 to 0.081) of a job more than workers who only ever have wage and salary earnings.

In years with positive earnings, workers received earnings from almost one and a half employers if they received only wage and salary income. Specifically, they worked at 1.409 to 1.420 employers if never self-employed, and 1.490 (=1.409+0.081) to 1.494 (=1.420+0.074) if ever selfemployed. Self-employment income displaced wage and salary jobs, but there were substantial differences depending on whether it constituted a worker's dominant form of income. If selfemployment provided secondary income, it displaced at most a small fraction (-0.013 to 0.143) of a wage and salary job. In contrast, if self-employment was a worker's dominant form of income, it displaced about one (0.804=8.817-0.013 to 1.145=1.002+0.143) wage and salary job. This suggests that the secondary self-employment generates increased labor activity.²¹

Has the relationship between business income receipt and wage and salary work changed over time? To address this question, we estimate a regression specification in which we allow for the effect of the presence of business income to vary by year. Separately by year, we estimate the effects of business income receipt, as well as the business income being the individual's dominant source of income. We additionally control for year effects. Results of this specification are presented in

²¹In Appendix Table F1 we consider the relationship between self-employment and total jobs worked without distinction on whether it constituted a worker's dominant form of income. We find that, overall, self-employment income is associated with 0.313 to 0.883 fewer wage and salary jobs. This suggests that there is less than one-to-one displacement of wage and salary jobs when workers receive self-employment income.





Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records. Plots show point estimates from a regression with the number of wage and salary jobs worked in a given year. Specifically, the year interactions for primary and secondary income from self-employment in a given calendar year. See text for additional details.

Figure 6.

A substantial change occurred in the relationship between business income receipt and the number of wage and salary jobs if the business income was the individual's secondary source of income. In the 1990s, secondary self-employment income was associated with 0.085 to 0.136 fewer wage and salary jobs. But in the years that followed the 2007-2009 recession, the presence of secondary self-employment income was associated with a reduction of only 0.032 to 0.039 fewer wage and salary jobs. Secondary income from self-employment never displaced a substantial number of wage and salary jobs and that this small degree of substitution has lessened over time. In contrast, if the business income was a person's dominant source of income, then business income receipt is associated with an additional eight-tenths fewer wage and salary jobs, and this differential was stable over time.

6.2 Within versus across industry reallocation

We now explore the extent to which self-employment is associated with individuals working in a more diverse set of sectors. If individuals work in more industries when receiving business income, this suggests that transitions into self-employment represent efficiency-enhancing labor reallocation. Specifically, we assess the extent to which business income receipt displaces wage and salary jobs where the outcome variables are both the total number of jobs held during the

	Total industries				
Intercept	1.258***	1.258***			
	(0.000)		(0.000)		
BI	0.444***	0.431***	0.384***	0.392***	
	(0.001)	(0.001)	(0.001)	(0.002)	
BI*BI dom			0.163***	0.116***	
			(0.002)	(0.002)	
Worker fixed effects	Ν	Y	Ν	Y	
R^2	0.053	0.406	0.055	0.407	

Table 4: Number of industries worked regressed on business income

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records. Regressions estimate how the number of industries worked differ based on an individual's Business Income (BI). See text for additional details.

year (wage and salary plus self-employment) and total number of unique North American Industry Classification System (NAICS) supersectors in which the person worked in the year. The key explanatory variables are variables indicating whether the person is receiving business income and whether the business income is the person's dominant (highest) source of earnings in the year. We have NAICS supersector information for the years 2002-2015, so we run regressions on this subsample of the CPS ASEC respondents with wage and salary earnings or self-employment earnings in the CPS-DER. The linked dataset includes individuals who are age 16 or older at the time of the CPS survey and who have received a Protected Identification Key (PIK). Individuals are included in this sample for all years 2002-2015 in which they are age 16+ and have (1) wage and salary earnings with at least one job with non-missing NAICS code in Business Register, or (2) self-employment earnings with non-missing NAICS supersector.

Results of this exercise are presented in Table 4. Workers were employed in 1.258 industries, on average, when working in wage and salary employment. Overall, the presence of business income was associated with 0.431 to 0.444 additional industries worked. We further distinguish the effect of business income on the number of industries worked based on whether it provided an individual's main source of income. Secondary self-employment income is associated with 0.384 to 0.392 more total industries. When self-employment was a person's dominant source of income, that person worked in 0.508 (= 0.392 + 0.116) to 0.547 (= 0.384 + 0.163) additional industries.

These findings suggest that self-employed individuals, whether self-employment is their dominant or secondary source of earnings, tend to work in new industries rather than the same industries as their wage and salary employment. More broadly, these findings provide further evidence that the observed rise in self-employment has led to reallocation of work across broad sectors of the economy and has been an important contributor to labor market fluidity.

7 Conclusion

Our paper documents that the rising share of the U.S. population that receives business income offsets a small but nontrivial portion of the apparent decline in labor market fluidity that occurred in recent decades. Most business owners do not receive wage and salary payments and are therefore outside the scope of published statistics on labor reallocation. Including business income recipients reduces the measured decline in the rate of hires and separations from 1994 to the early 2014 by 1.3 to 1.4 percentage points (about 8.3% to 8.7%). Secondary business income displaces few wage and salary jobs. Business income is associated with individuals working in more industries rather than fewer.

Our results imply that accounting for self-employment is important when analyzing labor market fluidity. The volatile tail of the labor market has undergone profound shifts in recent decades as the "gig economy" has provided more short-duration jobs. Wage and salary employment is certainly still the largest part of the labor market, but labor reallocation increasingly involves selfemployment.

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Appendices

A Definitions

A.1 Overview and employment concepts

This appendix provides definitions of employment concepts used in this paper and follows the notation in Abowd et al. (2009). Note that we are applying to our annual data concepts that were developed for quarterly data. Let w_{ijt} denote earnings for individual *i* from employer *j* in year *t*. If an individual has reported earnings from an employer in a given year and $w_{ijt} > 0$, then we infer the individual worked for the employer and call this employment relationship a job.

To count the set of individuals employed at a given point in time, we consider the subset of jobs that span two consecutive years. Formally, an individual is counted as employed at the beginning of year t if:

$$b_{ijt} = \begin{cases} 1, & \text{if } w_{ijt-1} > 0 \text{ and } w_{ijt} > 0 \\ 0, & \text{otherwise.} \end{cases}$$

Likewise, an individual is employed at the end of the year t if

$$e_{ijt} = \begin{cases} 1, & \text{if } w_{ijt} > 0 \text{ and } w_{ijt+1} > 0 \\ 0, & \text{otherwise.} \end{cases}$$

A.2 Hire and Separation Rates

A hire is dated to year t when earnings first appears at that employer. Formally,

$$h_{ijt} = \begin{cases} 1, & \text{if } w_{ijt-1} = 0 \text{ and } w_{ijt} > 0 \\ 0, & \text{otherwise.} \end{cases}$$

Similarly, an individual separates from an employer in year t if it is the last year that an employee

is observed at an employer. Formally,

$$s_{ijt} = \begin{cases} 1, & \text{if } w_{ijt+1} = 0 \text{ and } w_{ijt} > 0 \\ 0, & \text{otherwise.} \end{cases}$$

Following Abowd and Vilhuber (2011), the average of employment at the beginning and end of the year, i.e. $(b_{ijt} + e_{ijt})/2$, is used as the denominator for our hire and separation rates.

A.3 Employer-to-employer and nonemployment transitions

Employer-to-employer and nonemployment transitions are calculated at the person level following the employment transition concepts developed by Hyatt et al. (2014). We consider the jobs that span two consecutive years t - 1 and t. By definition, in such jobs the employee was employed at the beginning of year t, and so $b_{ijt} = 1$.

For any two-year pair, we disambiguate the data by considering jobs that are maximal earning among all jobs a worker holds at the beginning of year t. To do so, the job with the greatest (nominal) earnings summed across years t - 1 and t is identified, as follows:

$$domb_{ijt} = \begin{cases} 1, & \text{if } b_{ijt} = 1 \text{ and} \\ & w_{ijt} + w_{ijt-1} > w_{ikt} + w_{ikt-1} \forall k \\ & \text{s.t. } b_{ikt} = 1 \text{ and } j \neq k \\ & 0, & \text{otherwise.} \end{cases}$$

The set of jobs defined in $domb_{ijt}$ are those we use in all of our empirical analysis. Such jobs are unique at the person-year level. Average employment in the year, $(domb_{ijt} + domb_{ijt+1})/2$, serves as the denominator for employer-to-employer and nonemployment transition rates.

We consider within-year employer-to-employer transitions, as follows,

$$j2j_{ijkt} = \begin{cases} 1, & \text{if } domb_{ijt} = 1 \text{ and } domb_{ikt+1} = 1 \\ & \text{and } j \neq k \\ 0, & \text{otherwise,} \end{cases}$$

Flows into nonemployment in year t are defined as

$$en_{ijt} = \begin{cases} 1, & \text{if } domb_{ijt} = 1 \\ & \text{and } domb_{ilt+1} \neq 1 \forall l \\ 0, & \text{otherwise,} \end{cases}$$

Flows nonemployment into employment in year t are defined as

$$ne_{ikt} = \begin{cases} 1, & \text{if } domb_{ikt+1} = 1 \\ & \text{and } domb_{ilt} \neq 1 \forall l \\ 0, & \text{otherwise.} \end{cases}$$

A.4 Measuring secondary and short-duration jobs

We follow Hyatt and Spletzer (2013) in recovering secondary and short-duration jobs from hires, separations, employer-to-employer transitions, as well as transitions into and out of nonemployment. Total hires into secondary and short-duration jobs $h2_{ijt}$ are defined as:

$$\sum_{i,j} h_{ijt} = \sum_{i,j,k} j2j_{ijkt} + \sum_{i,k} ne_{ikt} + \sum_{i,j} h2_{ijt}$$

and total separations from secondary and short-duration jobs $s2_{ijt}$ can be defined analogously as

$$\sum_{i,j} s_{ijt} = \sum_{i,j,k} j2j_{ijkt} + \sum_{i,j} en_{ijt} + \sum_{i,j} s2_{ijt}.$$

Some additional arithmetic is required to measure the hire and separation rates associated with secondary and short-duration jobs because (job-level) hires and separation rates have slightly different denominators from the worker-level rates. The denominator for hires and separations are total jobs, that is, consecutive-year employer-employee combinations. The denominator for employerto-employer and nonemployment is total workers employed in any consecutive-year employeremployee combination. Therefore, we multiply these by the ratio of the denominators to express secondary and short-duration hires and separations in terms of total consecutive-year jobs.





Notes: CPS-DER numbers from 1996-2010 are from Figure 1 of Abraham et al. (2018). Other totals are authors' calculations on CPS-DER Schedule SE records, as well as universe-level Schedule C and Schedule K-1 records.

B Self-employed and business owners over time

We here assess the extent to which the universe-level administrative records capture a similar population as the CPS-DER. Population totals for the total number of business income recipients are shown in Figure B1. We calculate four totals. The first is the total number of CPS-DER Schedule SE filers, after applying weights to recover the population totals. Note that the CPS-DER Schedule SE totals from 1996-2010 are taken from Abraham et al. (2018). The total number of business income recipients in the CPS-DER increased steadily from 13.0 million in 1996 to 18.3 million in 2007, then declined to 17.3 million 2010, just after the 2007-2009 recession. In later years, it increased to 19.6 million in 2014 and fell slightly to 19.4 million in 2015.

The second group in Figure B1 is the total number of recipients of income from sole proprietor partnership businesses who file a Schedule SE in the universe-level administrative records. This includes a subset of those who report sole proprietor income, as well as those who report income from a partnership. There were 16.7 million Schedule SE filers in 2007, and this number fell to 15.1 million in 2009, and then increased to 18.9 million in 2015. The numbers of Schedule SE filers in the universe-level data were similar to those implied by the CPS-DER. The largest difference was about 15% in 2009, other years were different by less than 10%. The overall correlation between these two series is 0.78, but the correlation is much higher for 2010-2015, when it is 0.97.

Our third group is the total number of recipients of income from sole proprietor and partnership

	Schedule C	Form K-1	Form K-1
	Sole Proprietors	Partnerships	S Corps.
Share of owners:			
Owners of EIN firms in UI	7.3%	1.1%	39.4%
Owners of employer firms in UI	12.3%	9.3%	69.0%
Owners of employer firms, top 3	9.8%	6.9%	59.4%
Share of firms:			
Any owner of EIN firms in UI	7.8%	4.5%	48.6%
Any owners of employer firms in UI	13.6%	26.5%	83.0%
Any owners of employer firms, top 3	10.8%	22.4%	77.0%

Table C1: Share of owners who receive wage & salary payments (UI)

Notes: Authors' calculation of LEHD administrative records and universe-level Schedule C and Schedule K-1 records. This table shows the share of businesses owners who receive a unemployment insurance (UI) record of the wage and salary income from the businesses that they own, as well as the share of businesses that have at least one owner with a UI record. "Top 3" indicates that the owner is among the top 3 highest earners at that employer.

businesses, which was much larger than those who file a Schedule SE. 29.6 million reported such income in 2007, and this declined to 28.3 million in 2009, and then increased to 33.9 million in 2010. Overall, 55% to 57% of those who reported income from a sole proprietor or partnership business filed a Schedule SE.

Finally, the fourth group in Figure B1 is the total number of recipients of income from sole proprietor, partnership, and S corporations. Note that this group excludes business owner entities that are EINs (i.e., excluding businesses that own other businesses). There were 33.3 million business income recipients in 2007. This declined to 32.1 million in 2009, then increased to 37.6 million in 2015. Comparing the difference between this fourth group (all business income recipients) and the third group (all Schedule C sole proprietors and K-1 owners of partnerships) is the total number of owners of S corporations. The total number of owners of S corporations ranged from 3.6 to 3.8 million.

C How frequently are business owners employees?

Table C1 reports the relative frequency with which business owners are present in the Longitudinal Employer-Household Dynamics (LEHD) data. The LEHD data are derived from unemployment insurance (UI) wage and salary income records provided by U.S. states to the U.S. Census Bureau

	Schedule C	Form K-1	Form K-1
	Sole Proprietors	Partnerships	S Corps.
W-	2 Data		
Share of owners:			
Owners of employer firms in W-2	16.1%	14.3%	61.5%
Owners of employer firms, top 3	13.9%	11.4%	57.1%
Share of firms:			
Any owners of employer firms in W-2	17.0%	26.9%	74.6%
Any owners of employer firms, top 3	14.6%	23.5%	71.1%
LEI	HD Data		
Share of owners:			
Owners of employer firms in UI	15.5%	13.8%	59.6%
Owners of employer firms, top 3	13.5%	11.3%	55.4%
Share of firms:			
Any owners of employer firms in UI	16.4%	27.2%	72.7%
Any owners of employer firms, top 3	14.2%	23.8%	69.3%

Table C2: Share of owners who receive wage & salary payments: first year, single establishment

Notes: Authors' calculations of LEHD administrative records, as well as universe-level W-2, Schedule C, and Schedule K-1 records. Entries in the table indicate the share of businesses owners who receive a W-2 or have a LEHD unemployment insurance (UI) payroll record of the wage and salary income from the businesses that they own. Numbers are averages over time of annual averages from 2007-2014 for firms in their first year as employers that consist of a single establishment. "Top 3" indicates that the owner is among the top 3 highest earners at that employer.

as part of the Local Employment Dynamics federal-state partnership. The results are quite similar to Table 1, suggesting that owner wage and salary receipt is similar in federal income taxes (W-2 data), and state payroll taxes (UI wage records).

Table C2 reports the frequency that business owners are present in matched employer-employee data for a particular subset of firms. Following Kerr and Kerr (2017), we include only employer firms in their first year, where the firm had only one establishment in that year. For these employer firms, we identify the top three wage and salary earners at the firm and report the frequency that the business owner is one of these top three earners. Relative to Tables 1 and C1, this sample selection technique results in an increased fraction of business owners in sole proprietor businesses, and a lower share of business owners for S corporations. For example, 77.0% of S corporations employ at least one owner among their top three highest earning employees according to the UI wage records (Table C2), but in the year their S corporations hire their first employee, that share is only





Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records.

69.3% (Table C1). These results imply that many owners of S corporations start to receive wage and salary payments after they hire their first employee.

D Schedule SE share of consecutive year employment

Figure D1 shows the share of consecutive year employment attributable to Schedule SE filers in the CPS-DER. This is as a share of consecutive year employment, which is the denominator of the transition rates. At the start of 1991, self-employment constituted 7.2% of all employer-employee combinations, and 6.7% of all such that were dominant among a person's employer-employee combinations. The Schedule SE share of employment (dominant employment) declined to 7.1% (6.5%) in 2000, and then surged to 8.1% (7.5%) in 2005. Thereafter, the share of employment accounted for by Schedule SE filers was flat, and in 2016 Schedule SE files were 8.2% of employment and 7.5% of dominant employment.

E Reallocation rates for wage and salary employment

This paper contains the first estimates of labor reallocation rates using W-2 data. This appendix presents the basic trends in reallocation rates from our W-2 wage and salary data and benchmarks them against other available data sources.



Figure E1: Hire and separation rates: wage and salary (W-2) employment

Notes: Authors' calculations of CPS-DER W-2 records, as well as universe-level W-2 records. The denominator for all rates is the average of employment at the beginning and end of the year. See text for additional details.

E.1 Hire and separation rates in W-2 wage and salary data

E.1.1 Time trends

Figure E1 shows the time trend of hire and separation rates using W-2 data. After rising steadily during the 1990s, these rates fell sharply around the 2001 recession, stabilized briefly in the mid-2000s, and fell again during the 2007-2009 recession. This is the "stair-step" pattern noted by Hyatt and Spletzer (2013).

The W-2 based hire and separation from the universe-level files was consistently a few percentage points above those from the CPS-DER W-2 records. This difference may be due to sample selection and record linkage effects that are not completely accounted for by our linkage-adjusted CPS-DER weights. For example, people who responded to the CPS ASEC may have had slightly more stable employment histories than the population as a whole. Nevertheless, the rate calculated from the CPS-DER W-2 records is highly correlated with that of the universe-level W-2 records. The correlation between the hires series for years 2006-2015 is 0.978, and that of the separation series for years 2005-2015 is 0.984. At the end of our time series, hire and separation rates still had yet to recover to the levels observed during the 1990s. In 2015, the hire (separation) rate remained 13.9 (13.7) percentage points below the average hire (separation) rate from 1991-2000 of 64.5% (63.7%).

E.1.2 Benchmarking

We benchmark the W-2 hire and separation rates against the following data sources. JOLTS hires and separations were downloaded from https://www.bls.gov/jlt/. The national QWI hires and separations were taken downloaded from https://ledextract.ces.census.gov/static/data.html. For additional details on the construction of national QWI hires and separations, see Abowd and Vilhuber (2011).

For transitions from the CPS, we use data from two data sources. The CPS gross flows series, downloaded from https://www.bls.gov/webapps/legacy/cpsflowstab.htm, provides information on nonemployment-to-employment and employment-to-nonemployment transitions. Monthly rates of nonemployment-to-employment transitions are calculated in two steps. First, we sum unemployment to employment transitions (LNU07100000), not in the labor force transitions (LNU07200000), and other transitions to employment (LNU07300000). This sum is divided by total employment (LNU02000000). Analogously, we calculate monthly transition rate from employment-tononemployment as the sum of employment to unemployment (LNU07400000), employment to not in the labor force (LNU07800000), and other transitions out of employment (LNU08200000), divided by the same total employment. To measure employer-to-employer transitions, we use the EEhaz series calculated from CPS microdata by Fallick and Fleischman (2004), and downloaded from https://www.federalreserve.gov/pubs/feds/2004/200434/200434abs.html. Although data are available starting in 1994, we use these data starting in 1996 because published rates are missing for several months in 1995. Our CPS hire rate is the sum of the EEhaz series and our nonemployment-to-employment series. Our CPS separation rate is the sum of the EEhaz series and our employment-to-nonemployment series.

A challenge in attempting to benchmark our new data from the CPS-DER and universe-level administrative records against other data sources concerns the frequency of the underlying microdata. Our new transition rates are calculated using annual data, and so we calculate the rate of hire, separation, employer-to-employer, and nonemployment transitions on an annual basis. However, the other data sources are calculated at higher frequencies. The JOLTS and CPS are published at a monthly frequency, and the QWI and J2J are calculated at a quarterly frequency. We therefore approximate an annual rate from these data sources by taking the sum of the monthly or quarterly rates within a particular calendar year.

Our benchmarking exercises are affected by measurement issues, and these are somewhat different for hires and separations and the other transition rates. Employer-reported hire and separation rates from the QWI and JOLTS are calculated at a higher frequency than our new annual data, and therefore are more likely to record recalls as distinct separations and hires. In our annual data, if a separation and a recall occur in the same calendar year, we record it as a continuous employment spell because the employee's lapse in employment is not observed. This should occur less frequently in quarterly and monthly data.

The results of this exercise are shown in Figure E2. We include hire and separation rates from the CPS-DER linked with W-2 records, as well as universe-level W-2 records. We compare these with three additional measures of hires and separations from JOLTS, QWI, and CPS. The QWI has the highest hire and separation rates, and the JOLTS has the lowest hire and separation rates. The CPS has the least cyclical transition rates.

Figure E2(a) shows the results for hires, while E2(b) shows the results for separations. In 1993, the hire (separation) rate from the CPS-DER was 59.5% (58.1%). The QWI had a much higher hire (separation) rate of 102.1% (99%). In other words, total hires and separations in 1993 in the QWI were roughly the same as total employment, while in the W-2 data they were closer to three-fifths of employment. In 1996, the hire (separation) rate of 107.4% (102.6%) and the CPS-DER's rate of 65.8% (65.4%). In 2001, the annual hire (separation) rate from the JOLTS was the lowest at 47.2% (48.8%), that of the CPS-DER was higher at 59.0% (61.0%), the aggregates of CPS monthly data had a yet higher rate of 80.6% (81.7%), and the QWI was again highest at 100.9% (100.7%). In all data sources, separations exceeded hires in the recession year of 2001.

In 2006, we have hire and separation rates from the universe-level W-2 data, along with every other data series. Like the CPS-DER linked with W-2 data, our universe-level tabulations were consistently between the JOLTS and the monthly CPS. In 2006, the lowest hire (separation) rate was the JOLTS at 47.6% (46.0%), the CPS-DER had higher number at 55.4% (55.0%), the universe-level W-2 records had a rate of 63.7% (61.7%), the CPS monthly aggregates had a rate of 77.9% (75.7%), and the QWI had a rate of 94.2% (90.1%). At the end of the 2007-2009 recession, hires reached a local minimum in 2009 in each series: 35.6% in the JOLTS, 41.1% in the CPS-





Notes: Authors' calculations of CPS-DER W-2 records, universe-level W-2 data, and published aggregates. The denominator for the W-2 rates is the average of employment at the beginning and end of the year. Job Openings and Labor Turnover Survey (JOLTS), Quarterly Workforce Indicators (QWI), and CPS Gross Flows and Employer-to-Employer Transitions series sum rates within a calendar year. See text for additional details.

DER W-2 records, 45.3% in the universe-level W-2 records, 68.0% in the CPS monthly aggregates, and 67.8% in the QWI. Separations reached a local minimum in every series in 2010 except the aggregates of CPS microdata. In 2010, separations were 37.3% in the JOLTS, 43.4% in the CPS-DER W-2 records, 46.7% in the universe-level W-2 records, 68.5% in the QWI, and 69.6% in the aggregates of CPS monthly data. The separation rate in the CPS monthly data continued to decline until 2013, when it reached 68.5%. As document Fujita, Moscarini, and Postel-Vinay (2020), a change in the administration of the CPS resulted in a substantial, permanent decline in monthly transition rates starting in 2007.

The last year in which data from all series are available is 2014. In that year, the hire (separation) rate was 42.5% (40.2%) in the JOLTS, 49.7% (48.2%) in the CPS-DER W-2 records, 53.7% (51.5%) in the universe-level W-2 records, 70.5% (68.6%) in the aggregates of CPS monthly data, and 76.9% (72.5%) in the QWI. Most series increased in the expansion that followed the 2007-2009 recession. The exception is the aggregates of CPS monthly data, in which the hire (separation) rate was 69.2% (67.3%) in 2018.

Despite the level differences, our new hire and separation rates are highly correlated with these other series. The correlation between the hire (separation) rate in the CPS-DER W-2 records and the monthly aggregates from the CPS is 0.93 (0.91), 0.98 (0.98) with the QWI, and 0.97 (0.98) with the JOLTS. The correlation between the hire (separation) rate in the universe-level administrative records is 0.78 (0.81) with the monthly CPS aggregates, 0.93 (0.98) with the QWI, and 0.98 (0.97) with the JOLTS.

E.2 Employer-to-employer and nonemployment transitions in W-2 wage and salary records

E.2.1 Time trends

A challenge in attempting to benchmark our new data from the CPS-DER and universe-level administrative records against other data sources concerns the frequency of the underlying microdata. Our transition rates are calculated using annual W-2 data, and so we calculate the rate of hire and separation rates on an annual basis. However, the other data sources are calculated at higher frequencies. The JOLTS and CPS are published at a monthly frequency, and the QWI at a quarterly frequency. We therefore approximate an annual rate from these data sources by taking the sum of the monthly or quarterly rates within a particular calendar year.

These measurement issues affect the interpretation of our benchmarking exercises. In our annual data, if a separation and a recall occur in the same calendar year, we record it as a continuous employment spell because the employee's lapse in employment is not observed. More of these gaps in employment can be observed in quarterly and monthly data.

The evolution of the employer-to-employer, nonemployment-to-employment, and employmentto-nonemployment transition rates for our two data sources are shown in Figure E3. Results for





Notes: Authors' calculations of CPS-DER W-2 records, as well as universe-level W-2 records. The denominator for all rates is the average of dominant employment at the beginning and end of the year. See text for additional details.

the CPS-DER, shown in Figure E3(a), indicate that the employer-to-employer transition rate exhibited substantial cyclical declines around the 2001 and 2007-2009 recessions. It had recovered to 14.0% by 2015. The transition rates into and out of nonemployment also evolved cyclically. During expansions, the rate of nonemployment-to-employment transitions exceeded that of employment-to-nonemployment transitions, and so employment increased. During recessions, this patterns reversed as the employment-to-nonemployment transition rate surged. By 2015, both the employment-to-nonemployment and the nonemployment-to-employment transition rates were 1.5 percentage points below their 1991-2000 averages of 11.1% and 12.4% respectively.

Figure E3(b) shows that the transition rates in the universe-level W-2 data were somewhat higher than for the CPS-linked W-2 data. Transition rates evolved similarly in the shorter time

series of universe-level W-2 data, shown in Figure E3(b). In 2006, the employer-to-employer transition rate calculated from universe-level administrative records was 16.2%. The rate then fell to 12.0% in 2009, and recovered to 15.3% in 2014. In 2006, transitions out of nonemployment were 13.7% of average W-2 employment, while transitions into nonemployment were 11.9%. In the 2007-2009 recession, these rates reversed, as the employment-to-nonemployment transition rate reached 14.1% in 2009, when the nonemployment-to-employment transition rate was only 11.3%. In the years that followed, the nonemployment-to-employment rate recovered, and was 12.7% in 2014, when the employment-to-nonemployment rate was 11.8%. For example the employer-to-employer transition rate calculated from universe-level administrative records was 16.2% in 2006, compared to 13.9% in the CPS-DER. The universe-level W-2 records are highly correlated with the CPS-DER. The correlation between the employer-to-employer transition rates is 0.884, and the nonemployment-to-employment-to-employment rates is 0.908.

E.2.2 Benchmarking

We benchmark our W-2 employer-to-employer and nonemployment transition rates against other data sources. JOLTS layoffs and quits were downloaded from https://www.bls.gov/jlt/. Note that the JOLTS quit rate is not a direct measure of the employer-to-employer transition rate, but Hyatt et al. (2014) show it has a very high correlation with that of J2J. Similarly, the JOLTS layoff series is not a direct measure of employment-to-nonemployment transitions, but Hyatt et al. (2014) show that it is highly correlated with the J2J employment-to-nonemployment series. J2J data on employer-to-employer and nonemployment transitions, see Hyatt et al. (2014), were downloaded from https://lehd.ces.census.gov/data/j2j_beta.html. We use the J2JHires, NEPersist, and ENPersist for employer-to-employer transitions, nonemployment-to-employment, and employment-to-nonemployment at the beginning and end of every quarter. We aggregate CPS data to form employer-to-employer, employment-to-nonemployment, and nonemployment-to-employer transition rates as described in Appendix E.1.2.

Our W-2 measures of employer-to-employer and nonemployment transitions are calculated on an annual basis, and we seek to compare these to monthly data from the CPS and JOLTS, as well as quarterly J2J data. To do so, we sum the transititions by the calendar year in which they occurred. Note that our employer-to-employer and nonemployment measures consider transitions between dominant employers at particular points in time. In the CPS, such transitions occur between the months in which a person is surveyed. A person has at most one recorded transition, even if they had several transitions within a single month. Similarly, the J2J consider transitions between dominant employers from the beginning of a quarter to its end, and, again, employees can have at most one transition in a quarter. In our new annual series, people have at most one recorded transition between employers, or into or from nonemployment in a calendar year. In the J2J, an individual can have up to four transitions in a calendar year, and in the CPS an individual can have a transition between any pair of months, and so up to 12 transitions in a calendar year. Thus, the differences in the frequency with which the data are recorded can result in level differences between these different data sources.

We show the results of the benchmarking exercise in Figure E4. The transition rates using CPS-DER W-2 and universe-level W-2 records are consistently the lowest. We also find less agreement on the level of transitions into and from nonemployment than for hires, separations, or employerto-employer transitions.

Our employer-to-employer transition series are shown in Figure E4(a). In 1996, the CPS-DER W-2 records show that the annual employer-to-employer transition rate was 14.1%, while the sum of the monthly hazard rates from Fallick and Fleishman (2004) were more than twice that, at 31.3%. In 2001, we also compare the CPS-DER W-2 employer-to-employer transition rate with the employer-to-employer rate from J2J, as well as the quit rate from the JOLTS. There is also a conceptual relationship between these measures since employer-to-employer transitions usually involve a quit to take a new job. In 2001, the employer-to-employer transition rate was 13.8% in the CPS-DER W-2 records, 23.6% in J2J, and 29.8% when we aggregate the monthly CPS rates, while the JOLTS quit rates sum to 26.4%. In 2006, we can start to benchmark our rate from universe-level W-2 records, 16.2% in the universe-level W-2 records, 22.6% in the J2J, and 27.1% when we aggregate the monthly CPS rates, while the JOLTS quit rates are available is 2014. In that year, the CPS-DER W-2 records had an employer-to-employer for the sum of 26.4%.

to-employer transition rate of 13.1%, while the universe-level W-2 records show 15.3%, the CPS monthly rates sum to 19.9%, the J2J rate is 20.4%, and the JOLTS quit rate was 21.9%.

The CPS-DER W-2 and universe-level W-2 employer-to-employer transition rates are highly correlated with the other annual measures. The CPS-DER W-2 rate has a correlation of 0.85 with the monthly CPS rates, 0.98 with J2J, and 0.95 with the JOLTS quit rate. The rate from universe-level W-2 records has a correlation of 0.83 with the monthly CPS rates, 0.99 with J2J, and 0.97 with the JOLTS quit rate.

The employment-to-nonemployment transition rates, shown in Figure E4(b) exhibit considerable level differences. The rates from the CPS-DER W-2 and universe-level W-2s are consistently the lowest, and the sums of the monthly transition rates from the CPS Gross Flows are the highest. In 1991, the CPS-DER employment-to-nonemployment transition rate was 11.4%. In other words, somewhat more than one tenth of workers employed at the beginning of 1991 were employed at the end of 1991. The monthly transition rates from the CPS Gross Flows sum to 51.7%. The J2J employment-to-nonemployment rate and the JOLTS layoff rate were between the CPS-DER W-2 rates and the sum of the CPS monthly rates. In 2001, the CPS-DER W-2 data indicate that the employment-to-nonemployment transition rate was 11.9%, the J2J rates sum to 24.4%, the CPS monthly rates sum to 51.9%, and the JOLTS monthly layoff rates sum to 18.9%. These measures are also conceptually related because spikes in the employment-to-nonemployment rates that occur during recessions are driven by layoffs. The universe-level W-2 records yield an employment-tononemployment transition rate starting in 2006. In that year, the CPS-DER W-2 records had rate of 10.3%, the universe-level administrative records had a rate of 11.9%, the J2J rates sum to 21.7%, the CPS rates sum to 48.6%, and the JOLTS layoffs sum to 16.4%. In 2014, the last year for which data from all series are available, the CPS-DER had a rate of 9.4%, the universe-level administrative records had a rate of 11.8%, the J2J rates sum to 18.7%, the CPS-DER rates sum to 48.7%, and the JOLTS layoff rates sum to 15.2%.

The correlations between employment-to-nonemployment rates using W-2 records are highly correlated with the analogous rate from J2J: the CPS-DER has a correlation of 0.94 and the universe-level records has a correlation of 0.83. Both are also highly correlated with JOLTS layoff rate: the CPS-DER has a correlation of 0.89, and the universe-level records yield a correlation of 0.88. Correlations with the sum of the monthly employment-to-unemployment transition rates





(a) Employer-to-employer

Notes: Authors' calculations of CPS-DER W-2 records, as universe-level W-2 records, and published aggregates. The denominator for the W-2 based rates is the average of employment at the beginning and end of the year. Job Openings and Labor Turnover Survey (JOLTS), Quarterly Workforce Indicators (QWI), and CPS Gross Flows and Employer-to-Employer Transitions series sum rates within a calendar year. See text for additional details.

are much weaker: the CPS-DER has a correlation of 0.30, and the universe-level records yield a correlation of 0.72.

The nonemployment-to-employment transition rate series, shown in Figure E4(c), also exhibit considerable level differences. In 1991, CPS-DER W-2 records had a nonemployment-toemployment rate of 11.4%. The sum of the nonemployment-to-employment transitions in the CPS Gross Flows in that same year yields a far higher number, 51.7%. Starting in 2001, we have two additional data sources for comparison. The first is the nonemployment-to-employment rate from J2J. The second is the difference between the JOLTS hire rate and the JOLTS quit rate. This difference serves as a proxy for the number of workers who are hired who are not currently employed, as most quits involve a departure from one employer and a new job at another. In 2001, the nonemployment-to-employment transition rate from CPS-DER W-2 records was 11.5%, the J2J rates sum to 23.3%, and the differences between JOLTS hire and quit rates sum to 20.8%, and the rates from the CPS Gross Flows sum to 50.7%. Starting in 2006, there are also transition rates from the universe-level W-2 records. In that year, nonemployment-to-employment transition rate was 11.7% in the CPS-DER W-2 records, 13.7% in the universe-level W-2 records, the J2J rates sum to 23.4%, the CPS Gross Flows sum to 50.8%, and the differences between the JOLTS hire and quit rates sum to 21.3%. In 2014, the last year in which data for all series is available, the rate was 11.25% in the CPS-DER, 12.7% in the universe-level W-2 records, the J2J rates sum to 21.0%, the transitions from the CPS Gross Flows sum to 50.6\%, and the differences between the JOLTS hire and quit rates sum to 20.6%.

Our new nonemployment-to-employment measures from W-2 have lower correlations with other available data sources than our other transition rates. The correlation between the CPS-DER W-2 rate and that of J2J is 0.79, while the universe-level W-2 rate has a correlation of 0.74 with J2J. The correlation between the CPS-DER W-2 nonemployment-to-employment transition rate and the difference between the JOLTS hire and quit rates is 0.81, and the universe-level W-2 data has correlation with this JOLTS differential of 0.56. The lowest correlations are with the sum of the monthly nonemployment-to-employment transitions in the CPS Gross Flows, which have a correlation of only 0.14 with the CPS-DER W-2 transitions, and 0.29 with the universe-level W-2 transitions.





Notes: Authors' calculations of hires and separations using CPS ASEC respondents who report multiple nonoverlapping jobs in the prior year, as well as CPS-DER W-2 records. For the CPS ASEC responses, the denominator is those who were in scope of the multiple non-overlapping employers question. For the employer-to-employer transition rate derived from administrative records, the denominator is the average of employment at the beginning and end of the year.

E.2.3 Comparing W-2 employer-to-employer transitions with the March CPS

In addition to the benchmarking exercise conducted above, we here compare the employer-toemployer transition rate computed from the CPS-linked W-2 records with these same survey respondents' answers to the multiple non-overlapping employer question in the CPS ASEC. Since 1976, the CPS ASEC has asked respondents "For how many employers did [you or someone else in your household] work? If more than one at same time, only count it as one employer." Recent studies including Molloy et al. (2016) and Hyatt et al. (2018) have used the responses to this survey question to understand the long-run trends in reallocation. Survey response data was downloaded from iPUMS, see Ruggles et al. (2010).

These two alternative measures of the employer-to-employer transition rate exhibit striking differences. Figure E5 shows that the employer-to-employer transition rate calculated from CPS-linked W-2 records increased during the expansion between the 1990-1991 and 2001 recessions, whereas the rate calculated from survey responses was relatively stable. Furthermore, the precipitous decline in employer-to-employer transitions among CPS ASEC survey respondents that starts in 2001 is less severe and more cyclical in the W-2 records matched to these same survey respondents. Lastly, only the employer-to-employer transition rate calculated using the CPS ASEC responses remains depressed relative to its 1991-2000 average.

The differing long-term trends of these two measures of the employer-to-employer transition

rate provide further evidence of the importance of declines in short-duration jobs. The employerto-employer transition rate derived from CPS-linked W-2 data only identifies transitions where the start-of-year dominant employer of an individual changed from one year to the next. The CPS ASEC provides a more expansive measure of employer-to-employer transitions since it includes job transitions where one or more of the jobs starts and stops within the calendar year (both measures ignore multiple simultaneous jobs). Thus, the fact that only the employer-to-employer transition rate measured using the CPS ASEC responses exhibits a long-term decline is indicative that this decline is driven by a decline in short-duration jobs.

An alternative explanation for this discrepancy is survey effects. Survey effects are consistent with the unusual properties that the CPS ASEC series displays during the 1990s, as well as its feature that measured reallocation rates declined more in the 2001 recession than the much more severe 2007-2009 recession. Kaplan and Schulhofer-Wohl (2012) attribute the strong measured decline in interstate migration in the early 2000s in the CPS ASEC to survey effects, especially for imputed and proxy responses. Hyatt et al. (2018) note increasing divergence between CPS ASEC survey responses on interstate migration and administrative records for the same time period. It is possible that similar survey effects contribute to the measured decline in the multiple non-overlapping employers question in the CPS ASEC.

F Reallocation vs. reclassification: supplemental tables

Tables F1 and F2 contain point estimates from regressions of the number of jobs held by an individual on an indicator for whether the individual is self-employed in that year (in this table we only consider individuals who are ever self-employed). We have two dependent variables. Table F1 considers the number of wage and salary jobs. Table F2 considers the total number of jobs worked (including both wage and salary jobs and self-employment). We report estimation results for regressions performed both with and without worker-specific fixed effects. In Table F3 we add to our regression model an interaction term indicating whether a worker's self-employment earnings are dominant, i.e. the self-employment earnings are greater than the worker's earnings from any wage and salary employer in the year.

	CPS-DER	Universe	CPS-DER	Universe
Intercept	1.495***	1.490***		
	(0.000)	(0.001)		
BI	-0.883***	-0.757***	-0.596***	-0.313***
	(0.001)	(0.002)	(0.001)	(0.001)
Worker fixed effects	Ν	Ν	Y	Y
R^2	0.148	0.146	0.421	0.618

Table F1: Regression of number of wage and salary jobs worked on business income

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 records. See text for additional details.

Table F2: Regression of total number of jobs (including business income) on business income

	CPS-DER	Universe	CPS-DER	Universe
Intercept	1.495***	1.490***		
	(0.000)	(0.001)		
BI	0.117***	0.567***	0.404***	0.748***
	(0.001)	(0.001)	(0.001)	(0.001)
Worker fixed effects	Ν	Ν	Y	Y
R^2	0.003	0.026	0.323	0.757

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 records. Regressions estimate how the number of jobs worked differ based on an individual's Business Income (BI). See text for additional details.

	CPS-DER	Universe	CPS-DER	Universe
Intercept	1.456***	1.409***		
	(0.000)	(0.000)		
Ever BI	0.077***	0.081***		
	(0.001)	(0.001)		
Ever BI * BI	0.829***	1.251***	0.899***	1.805***
	(0.001)	(0.001)	(0.002)	(0.002)
Ever BI * BI * BI dom	-1.002***	-1.080***	-0.818***	-0.841***
	(0.001)	(0.001)	(0.002)	(0.002)
Worker fixed effects	Ν	Ν	Y	Y
R^2	0.028	0.074	0.351	0.768

Table F3: Regression of total number of jobs worked on business income, by dominant earnings

Notes: Authors' calculations of CPS-DER W-2 and Schedule SE records, as well as universe-level W-2, Schedule C, and Schedule K-1 data. Regressions estimate how the number of jobs worked differ based on an individual's Business Income (BI). See text for additional details.