

Job Search, Job Finding and the Role of Unemployment Insurance History

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Introduction

Standard search theory suggests:

- ① Returns to search $\uparrow \rightarrow$ Search intensity \uparrow
- ② Job search intensity $\uparrow \rightarrow$ Job finding probability \uparrow

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- Both **current** and **former UI recipients** search harder than **those never receiving UI**, but
- Both **current** and **former UI recipients** have a lower job finding rate than **those never receiving UI**

This Paper

To account for these empirical findings, I construct a search and matching model with endogenous search intensity, and

- Unproductive and/or inefficient job search

I then study the implications on

- Cyclical behaviour of search intensity
- Labour market dynamics (fluctuations and persistence)
- Effects of UI extensions (the Great Recession)

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Main findings: Unproductive/inefficient job search →

- ① Acyclical aggregate search intensity + underestimated matching efficiency
- ② Dampened labour market fluctuations and UI effects
- ③ More persistent unemployment

Literature

- **Job search of employed vs non-employed:** Faberman, Mueller, Şahin and Topa (2022)
- **Job search and unemployment duration:** Krueger and Mueller (2010, 2011), Guglielminetti, Lalive, Ruh and Wasmer (2018), Faberman and Kudlyak (2019), Marinescu and Skandalis (2021), DellaVigna, Heining, Schmieder and Trenkle (2022)
- **Job search and the business cycle:** Shimer (2004), DeLoach and Kurt (2013), Gomme and Lkhagvasuren (2015), Mukoyama, Patterson and Şahin (2018), Leyva (2018)
- **UI and job findings:**
 - Aaronson, Mazumder and Schechter (2010), Valletta and Kuang (2010), Farber and Valletta (2011), Rothstein (2011), Fujita (2011), Mazumder (2011), Nakajima (2012), Barnichon and Figura (2014)
 - Hagedorn, Karahan, Manovskii and Mitman (2019), Mitman and Rabinovich (2020), Rujiwattanapong (2022), Acosta et al. (2023), Birinci and See (2023)
- **The persistence of unemployment and worker heterogeneity:** Hornstein (2012), Kroft, Lange, Notowidigdo, and Katz (2015), Ravn and Sterk (2017), Hall and Schulhofer-Wohl (2018), Ahn and Hamilton (2020), Ahn (2022)

Data

- **Current Population Survey (CPS):** Basic monthly data + Occupational Mobility and Job Tenure supplements
- **Period:** 1998-2022
- **Sample:** Unemployed workers age 21-64
- **Job search intensity:**
 - ① Average number of job search methods used:
“What are all of the things you have done to find work during the last 4 weeks?”
 - ② Minutes of daily job search:
Imputed from American Time Use Survey à la Mukoyama et al. (2018)
- **Job finding probability:** Unemployment-to-employment transition probability

Data

- UI-related questions:
 - ① “Did you receive unemployment insurance benefits after that job ended?”
 - ② “Did you exhaust your eligibility for unemployment benefits?”

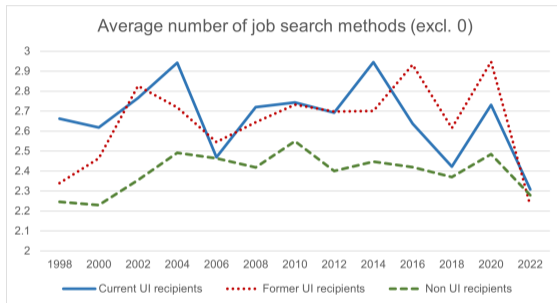
- Basic UI facts in the US:
 - ~70% of UI-eligible workers collect their benefits (Auray et al., 2009)
 - ~40% of all unemployed workers collect their benefits
 - 26 weeks of standard maximum UI duration
 - Some requirements:
 - Have earned enough wages during the base period
 - Be unemployed through no fault of your own

Job Search Methods

- ① Contacted employer directly/interview
- ② Contacted public employment agency
- ③ Contacted private employment agency
- ④ Contacted friends or relatives
- ⑤ Contacted school/university employment center
- ⑥ Sent out resumes/filled out application
- ⑦ Checked union/professional registers
- ⑧ Placed or answered ads
- ⑨ Other active
- ⑩ Looked at ads
- ⑪ Attended job training programs/courses
- ⑫ Nothing
- ⑬ Other passive

Data: Search Intensity

≡ number of job search methods used in a month



NB: Non-UI recipients = those who never received UI during current unemp. spell

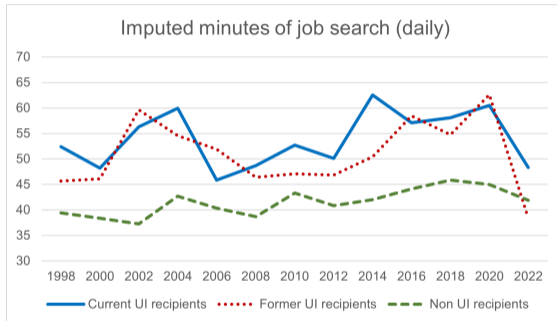
Source: CPS monthly data & January supplements

- Current and former UI recipients search harder than those never received UI (respectively 12% and 11% more) ▶ Including zero methods

Data: Search Intensity

≡ (imputed) minutes of daily job search

à la Mukoyama et al. (2018)'s 2-stage Heckman selection model



NB: Non-UI recipients = those who never received UI during current unemp. spell

Source: CPS monthly data, CPS January supplements, ATUS

- Current and former UI recipients search longer than those never received UI (respectively 30% and 23% more)

What explains search intensity?

Dependent variable: Job search intensity		
	(1)	(2)
	Number of methods	Imputed minutes
Current UI recipient	0.162*** (0.053)	2.605** (1.201)
Former UI recipient	0.127*** (0.042)	2.192** (0.955)
Unemployment duration (quartic)	✓	✓
Controls for worker characteristics, time and state fixed effects, recession dummy, and state unemployment rates	✓	✓
<i>N</i>	7,561	7,561
<i>R</i> ²	0.354	0.486

- Linear regression coefficients (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$)
- CPS monthly data and January supplements (1998-2022)
- Excluding 0 search methods (those on recalls)
- Worker characteristics include: race, education, gender, age (quartic), marital status, married-female, occupation, industry, recall expectation, potential UI exhaustion month, reason for unemployment, last job tenure, last job's weekly earnings

▶ More coefficients

Search Intensity

- **Current and former UI recipients** search *more intensely* than those never receiving UI
 - ▶ Who are they?
- **Current UI recipients:** UI eligibility and job search requirements (next slide)
- **Former UI recipients:** they now really want to be employed (?)
- How does search intensity translate into job finding probabilities?

UI Job Search Requirements in the US

- Every state must have a work search requirement but they design their own
- In **strict states** (e.g. Florida, Nebraska, Missouri, New Hampshire, and Utah):
 - Contact 4-5 new potential employers weekly
 - Submit a (bi-)weekly report
 - Use multiple methods (NH)
- In **lenient** states (e.g. California, Delaware, Massachusetts, South Carolina, West Virginia):
 - Broader definitions of work search activities
 - 1-2 activities each week (4 for WV but broad definitions)
 - Report upon request
 - If unemployment is high, no work search requirements (Michigan)
- Cross-state variations in the number of direct contacts and other activities required
 - An average state requires 1.2 direct contacts and a total of 2.8 work search activities each week (SD: 1.4 and 1.3)

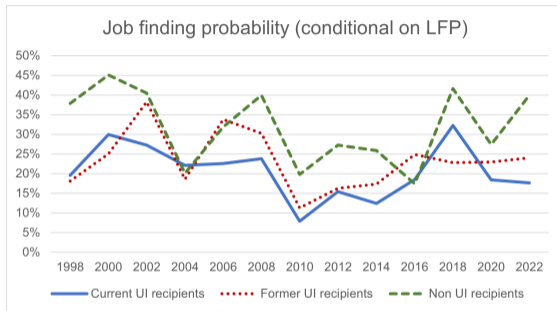
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	(1) Methods	(2) Minutes	(3) Methods	(4) Minutes	(5) Methods	(6) Minutes
Current UI recipient	0.162*** (0.053)	2.605** (1.201)				
Total search activities (required by state)			0.010*** (0.003)	0.178** (0.073)		
No. of direct contacts (required by state)					0.009* (0.005)	0.077 (0.107)
No. of other activities (required by state)					0.011*** (0.004)	0.246*** (0.090)
Former UI recipient	0.127*** (0.042)	2.192** (0.955)	0.106** (0.042)	2.056** (0.940)	0.106** (0.042)	2.064** (0.940)
<i>N</i>	7,561	7,561	7,561	7,561	7,561	7,561
<i>R</i> ²	0.354	0.486	0.347	0.486	0.347	0.487

- * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$.
- Other control variables include race, education, gender, age (quartic), marital status, female and married, occupation, industry, unemployment duration (quartic), recall expectation, potential UI exhaustion month, reason for unemployment, previous job's tenure, previous job's weekly earnings, a linear time trend, a recession dummy, state fixed effects, and state unemployment rates.

Data: Job Finding Probability

≡ monthly transition prob. from unemployment to employment



NB: Non-UI recipients = those who never received UI during current unemp. spell

Source: CPS monthly data & January supplements

- Those never receiving UI find jobs the quickest (32%)
- Current UI recipients and former UI recipients are slower to find jobs (21% and 23%) despite searching harder

What explains the job finding probability?

	Dependent variable: Job finding probability					
	(1)	(2)	(3)	(4)	(5)	(6)
Current UI recipient	-0.087*** (0.013)	-0.080*** (0.013)	-0.086*** (0.022)	-0.082*** (0.022)		
Total search activities (required by state)			0.000 (0.002)	0.000 (0.002)	-0.005*** (0.001)	-0.005*** (0.001)
Former UI recipient	-0.048*** (0.015)	-0.014 (0.016)	-0.047*** (0.015)	-0.014 (0.016)	-0.030** (0.015)	0.002 (0.015)
Search intensity	0.102* (0.058)	0.101* (0.057)	0.103* (0.058)	0.102* (0.057)	0.103* (0.058)	0.102* (0.057)
Unemployment duration (quartic)	NO	YES	NO	YES	NO	YES
<i>N</i>	5,693	5,693	5,693	5,693	5,693	5,693
<i>R</i> ²	0.064	0.076	0.068	0.079	0.066	0.077

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► Re-employment wages

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- Current UI recipients find jobs more slowly than non-UI recipients
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- This paper proposes that
 - ① **Current UI recipients** search harder due to UI search requirements, but they can *censor* their job search
 - Endogenously unproductive job search
 - Wang and Williamson (1996, 2002), Hall and Mueller (2018), Faberman et al. (2022)

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 - Wang and Williamson (1996, 2002), Hall and Mueller (2018), Faberman et al. (2022)
 - ② **Former UI recipients** search harder, but less efficiently
 - Exogenously inefficient job search
 - Duration dependence: employer screening, loss of networks, human capital, genuine duration dependence, etc. (Kroft et al., 2015; Jarosch and Pilossoph, 2019; Ahn and Hamilton, 2020)
 - By definition: long-term unemployed, Measure directly from the data

Model Outline

Built on standard DMP model with random search + heterogeneous workers from UI history

What's endogenous?

Deterministic model: **job search intensity** + vacancy creation

Stochastic model: above + job formations + job separations

Model Setup: Deterministic

- Time is discrete, runs forever
- Economy is populated by a continuum of **workers** of mass one and a large continuum of **firms**:
 - Infinitely-lived, risk neutral, ex ante identical

Workers:

- Either employed (E) or unemployed
- Unemployment statuses:
 - ① Current UI recipient: B
 - ② Former UI recipient: X (no UI + search less efficiently)
 - ③ Non-UI recipient: N (no UI)

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 - ③ Non-UI recipient: N (no UI)
- **Firms**: Either matched with one worker or unmatched
- **Output** of a worker-firm match is $y = z$ (TFP)

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Job search:

- Current UI recipients are required to exert " \underline{s}_B " but *optimally* censor it at rate $1 - \gamma_B \in (0, 1)$ (to lower disutility from search)
 - Current UI recipients' *effective* search intensity = $\gamma_B \cdot \underline{s}_B$

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- **Non-UI** and **former UI** recipients optimally choose s_N and s_X (resp.)
 - Non-UI recipients' *effective* search intensity = s_N

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 - $\gamma_X \in (0, 1)$: relative search efficiency

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Utility flows:

- **Employed** workers: wage w (Nash bargaining)
- **N-** and **X-**type workers: leisure flow h
- **B-**type workers: UI benefit b + leisure flow h

Model Setup: Deterministic

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- If **matched**: produce and earn y , pay wage w and lump-sum tax τ

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Matches:

- $M(s, v) = \frac{sv}{(s^\alpha + v^\alpha)^{\frac{1}{\alpha}}}$; $\alpha > 0$: CRS, concave
 - v = number of vacancies
 - s = aggregate search intensity
= number of searchers augmented by effective search intensity

Job finding rate per unit search: $\frac{M(s, v)}{s} \equiv p$

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- $Pr(X \rightarrow E) \equiv p_X(s_X) = \gamma_X \times s_X \times p$ (Former-UI)
- $Pr(B \rightarrow E) \equiv p_B(\gamma_B) = \gamma_B \times \underline{s}_B \times p$ (Current-UI)
 - γ_X : relative search efficiency of former UI to non-UI workers
 - γ_B : censored job search
 - \underline{s}_B : UI search requirement

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- **UI take-up**: Upon job separation, a fraction $1 - \psi \in (0, 1)$ of employed workers receive benefits, i.e.
 - $Pr(E \rightarrow B | \text{separation}) = 1 - \psi$
 - $Pr(E \rightarrow N | \text{separation}) = \psi$
 - $Pr(E \rightarrow X | \text{separation}) = 0$
 - Implications on eligibility, search cost heterogeneity, etc.

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- **UI exhaustion**: Current UI workers exhaust their benefits w/ probability $\phi \in (0, 1)$ and become former UI workers
 - $Pr(B \rightarrow B \mid \text{stay unemployed}) = 1 - \phi$
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- **UI budget:** Balanced each period by taxing producing firms

Value Functions: Unemployed

Respectively for current UI (B), former UI (X), and non-UI (N) recipients:

$$U_B = \max_{\gamma_B} b + h \underbrace{-c(\gamma_B)}_{\text{disutility from not censoring}} + \beta \left[\underbrace{p_B(\gamma_B)}_{Pr(B \rightarrow E)} W_B + (1 - p_B(\gamma_B)) \left((1 - \phi) U_B + \underbrace{\phi}_{Pr(\text{UI exhausted})} U_X \right) \right]$$

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$$U_X = \max_{s_X} \underbrace{h - c_X(s_X)}_{\text{disutility from search}} + \beta \left[p_X(s_X) W_X + (1 - p_X(s_X)) U_X \right]$$

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$$U_X = \max_{s_X} \underbrace{h - c_X(s_X)}_{\text{disutility from search}} + \beta \left[p_X(s_X) W_X + (1 - p_X(s_X)) U_X \right]$$

$$U_N = \max_{s_N} h - c(s_N) + \beta \left[p_N(s_N) W_N + (1 - p_N(s_N)) U_N \right]$$

- $W_j \equiv$ value of employment given previous status $j \in \{B, X, N\}$
- $\text{Pr}(N \rightarrow E) \equiv p_N(s_N) = s_N \times p$ (Non-UI)
- $\text{Pr}(X \rightarrow E) \equiv p_X(s_X) = \gamma_X \times s_X \times p$ (Former-UI)
- $\text{Pr}(B \rightarrow E) \equiv p_B(\gamma_B) = \gamma_B \times \underline{s}_B \times p$ (Current-UI)
- $c'(\cdot) > 0, c''(\cdot) > 0$
- Search cost parameter for X -type is different from others

Value Functions: Employed, Matched and Unmatched Firm

The value functions of employed workers with previous status $j \in \{E, B, X, N\}$:

$$W_j = w_j + \beta \left[\underbrace{(1 - \delta) W_E}_{\text{Pr(match survives)}} + \underbrace{\delta}_{\text{Pr(match destroyed)}} \left(\underbrace{(1 - \psi) U_B + \psi U_N}_{\text{Pr(not taking up UI)}} \right) \right]$$

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A matched firm with a worker whose previous status is $j \in \{E, B, X, N\}$ has the following value function:

$$J_j = y - w_j - \tau + \beta(1 - \delta)J_E$$

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The value function of an unmatched firm, assuming free entry, is:

$$0 = -\kappa + \beta q \left[\frac{\sum_{j \in \{B, X, N\}} \tilde{s}_j u_j J_j}{s} \right]$$

where \tilde{s}_j = effective search intensity of type- j workers,

u_j = stock of unemployed of type j , and

s = aggregate search intensity

► Nash Wage

► Search Cost

Stochastic Model

- Very few *UIx* workers in the steady state
- Will instead study a recession together with UI extensions
 - The Great Recession in the US
 - UI extensions from standard 26 weeks to (up to) 99 weeks
- 3 additional ingredients:
 - ① Aggregate productivity (z) as an $AR(1)$ process
 $\ln z_t = \rho_z \ln z_{t-1} + \varepsilon_t; \quad \varepsilon_t \sim N(0, \sigma_z^2)$

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 - UI benefit $b(m) \uparrow$ with m

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 - ③ State-dependent UI extension: $\phi(u)$ [▶ More on \$\phi\(u\)\$](#) [▶ More on policy functions](#)

Stochastic Model: Equilibrium Definition

A recursive competitive equilibrium is characterised by

- **value functions** $\{W_j(m; \omega), U_j(\omega), J_j(m; \omega), V(\omega)\}$;
- **market tightness** $\theta(\omega)$;
- **search policy** $s_j(\omega)$;
- **wage functions** $w_j(m; \omega)$;

for $j \in \{E(\tilde{m}), Bs(\tilde{m}), Bl(\tilde{m}), X, N\}$, such that, given initial distribution of workers, government's policy $\{\tau(\omega), \phi(\omega)\}$ and law of motion for z :

- ① Value functions and the market tightness satisfy the Bellman equations for workers and firms, and the free entry condition
- ② Search decisions satisfy the FOCs for optimal search intensity ▶ FOCs for search
- ③ Wage functions satisfy the FOCs for the Nash bargaining rule
- ④ UI's budget constraint is satisfied
- ⑤ The distribution of workers evolves according to the transition equations, consistent with the maximising behaviour of agents ▶ transitions

Macro Implications

① Unemployment-triggered UI extensions

- More of high- s types (those with a UI history) during recessions
- Countercyclical aggregate search intensity (?)

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 - UI recipients do not censor job search as much
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 - **Dampened labour market fluctuations**
- ③ Despite high search intensity, current UI workers' **censored job search** + former UI workers' **inefficient job search**
 - Persistently low job finding rates
 - **Unemployment persistence**

Calibration

- I calibrate a subset of parameters (12) to the US economy during 1948-2007 (SMM) including
 - Search cost parameter: a_X
 - Meeting function parameter
 - Match-specific productivity parameters
 - TFP-related parameters
 - UI-related parameters
- Other parameters are from the literature
- From Gruber (1997), set $b(m)$ and h to imply
 - 10% consumption drop from E to B (receiving $b(m) + h$); $\forall m$
 - 24% consumption drop from E to $\{X, N\}$ (receiving h)
 - Given 50% replacement rate

▶ Other fitted parameters

▶ Other fixed parameters

Targeted Moments

Moment	Data	Baseline	$\gamma_X = 1$
$E(u)$	0.0583	0.0569	0.0577
$E(\rho_{UE})$	0.4194	0.4414	0.4286
$E(\rho_{EU})$	0.0248	0.0252	0.0251
$E(\rho_{EE})$	0.0320	0.0322	0.0320
$E(udur)$	15.416	11.251	13.063
$E(u_B)$	0.0290	0.0302	0.0327
$std(u)$	0.1454	0.1123	0.1453
$std(\rho_{UE})$	0.0999	0.1035	0.1402
$std(\rho_{EU})$	0.0890	0.0517	0.0641
$std(LP)$	0.0131	0.0106	0.0104
$corr(LP, LP_{-1})$	0.7612	0.7609	0.7593
$E(\rho_{UE}^X / \rho_{UE}^N)$	0.7983	0.8012	1

- ρ_{UE} : job finding rate // ρ_{EU} : job separation rate // ρ_{EE} : job-to-job transition rate
 $udur$: mean unemployment duration (weeks) // LP : output per worker (quarterly)
- $\gamma_X = 1$: No duration dependence for former UI recipients
- **Relative job finding rates** much more realistic w.r.t. scenario without job search inefficiency

Application to the Great Recession in the US

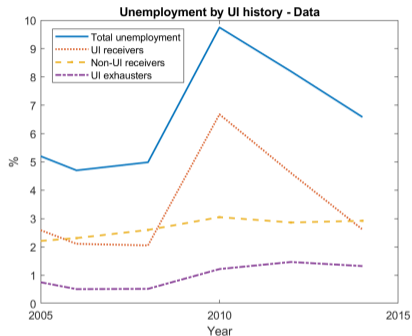
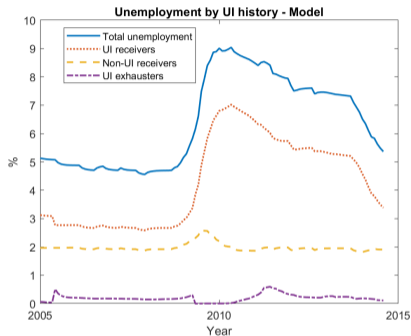
- I apply the model to the US economy (1948-2014) by **feeding in**:

① Historical UI extensions:

- Extension generosity varies with each recession
- Exogenous changes in ϕ_L to match observed changes in UI extension policies
- Agents rationally expect whether $\phi(u)$ will be ϕ or ϕ_L based on current u
- Agents assume ϕ_L never changes its value until observed otherwise
- Timings of extensions still endogenous - triggered when $u > \bar{u}$ ▶ Simulated UI extensions

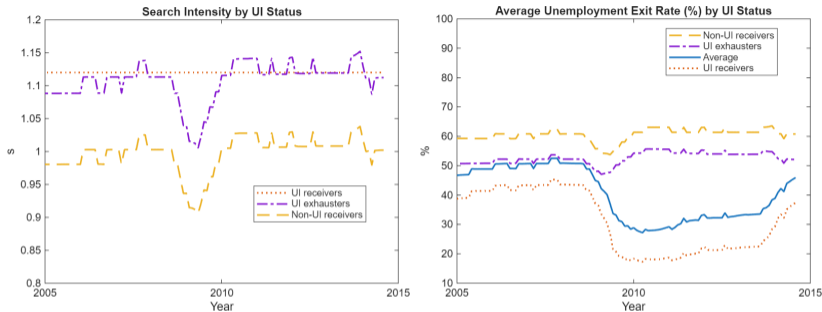
- ## ② A series of **productivity shocks** that match the deviations of output (GDP per capita) from HP trend

Unemployment Decomposition: Model vs Data



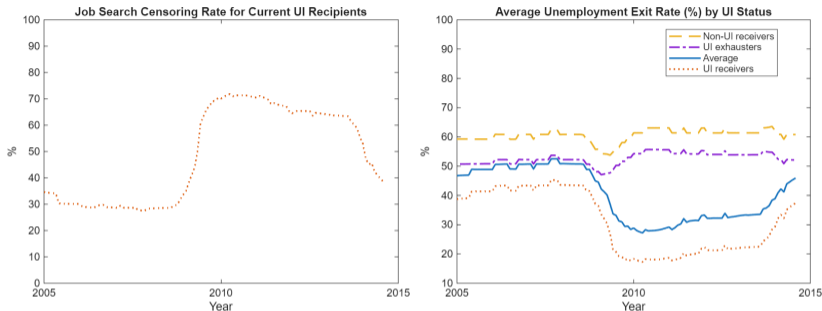
- Lower share of **UI exhausters** (X) in the model
- Slightly higher share of **UI receivers** (B) in the model
- The role of finer duration dependence? Unobserved heterogeneity?

Search intensity and Job Finding



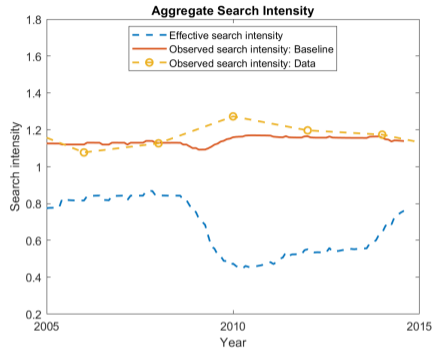
- UI exhausters' search inefficiency ($\gamma_X = 0.72$) is set such that $s_X/s_N = 1.11$
- Search is mildly procyclical for UI exhausters and non-UI receivers
- s_B is fixed at 1.12

Job Censoring



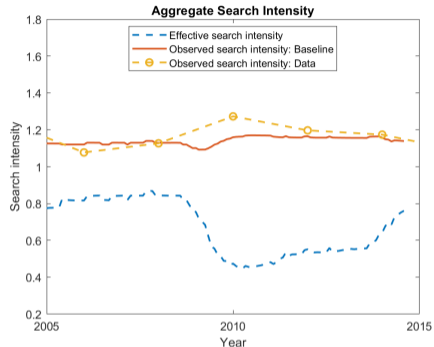
- **Current UI recipients'** job censoring rate ($1 - \gamma_B$): 37% on average
 - \uparrow UI extensions, \uparrow benefit amount $b(m) \rightarrow$ more censoring
 - $\uparrow z \rightarrow$ less censoring
 - Faberman et al. (2022): Unemployed workers reject 47% of best offers (52% of all offers)

Aggregate Search Intensity



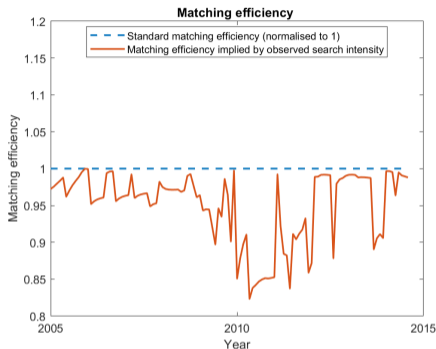
- **Observed** search intensity: acyclical (mildly countercyclical)
 - Composition effect: more workers with UI history (high s) during recessions
 - Falsely suggesting limited role of search intensity

Aggregate Search Intensity



- **Observed** search intensity: acyclical (mildly countercyclical)
 - Composition effect: more workers with UI history (high s) during recessions
 - Falsely suggesting limited role of search intensity
- **Effective** search intensity: smaller + procyclical
 - UI receivers' search censoring + UI exhausters' inefficient search
 - Even further with UI extensions (from 2009)

Matching Efficiency



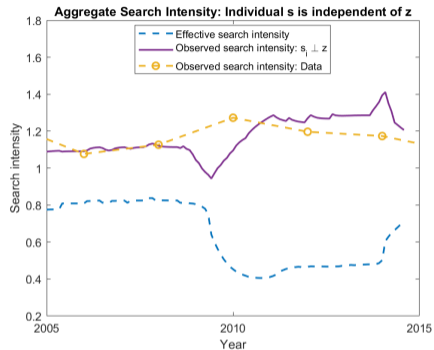
- $M = 1 \frac{sv}{(s^\alpha + v^\alpha)^{1/\alpha}} = A \frac{s_{obs}v}{(s_{obs}^\alpha + v^\alpha)^{1/\alpha}}$
- With **effective** search intensity, the matching efficiency is 1 (normalisation)
- With **observed** search intensity, matching efficiency is the residual A
 - Not corrected for censoring + inefficient search
 - Underestimate the actual matching efficiency up to 17%

Conclusion

- This paper shows empirically there is high search intensity but low job finding rates for workers with a UI history
- To reconcile with these findings, I propose a model of unproductive and inefficient job search
- Aggregate search intensity may be overestimated, esp. in recessions, leading to an underestimation in matching efficiency
- Accounting for effective search intensity implies
 - (1) smaller effects of UI extensions,
 - (2) dampened labour market fluctuations, and
 - (3) more persistent unemployment and its duration

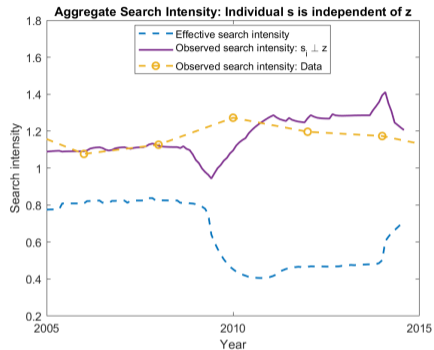
Thank you

Aggregate Search Intensity: Individual Job Search \perp Shocks



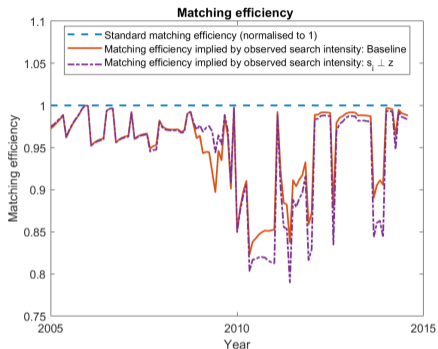
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Aggregate Search Intensity: Individual Job Search \perp Shocks



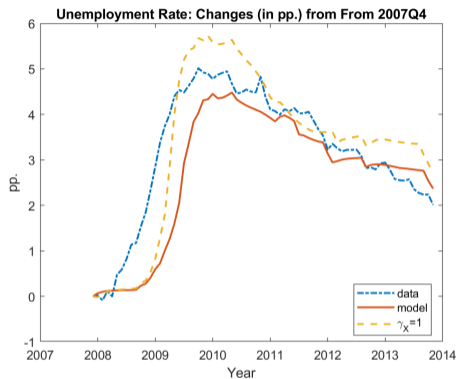
- **Effective** search intensity: smaller + procyclical
 - UI receivers' search censoring + UI exhausters' inefficient search
 - Even further with UI extensions (from 2009)
- **Observed** search intensity: Countercyclical but more volatile than the data
 - Assuming individual job search does not respond to shocks (z)

Matching Efficiency



- $M = 1 \frac{sv}{(s^\alpha + v^\alpha)^{1/\alpha}} = A \frac{s_{obs}v}{(s_{obs}^\alpha + v^\alpha)^{1/\alpha}}$
- With **effective** search intensity, the matching efficiency is 1 (normalisation)
- With **observed** search intensity, matching efficiency is the residual A
 - Underestimate the actual matching efficiency **up to 17%**
 - **Up to 21%** when individual search is assumed to be acyclical

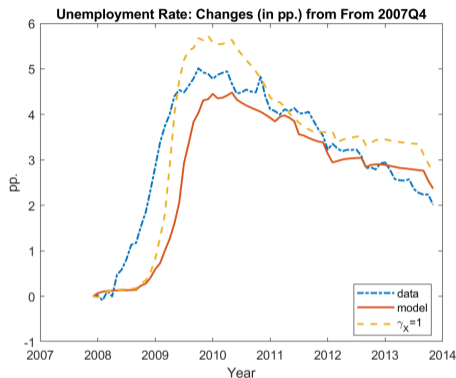
Unemployment



$\gamma_x = 1$: No duration dependence for former UI recipients

- Search inefficiency dampens the fluctuations in unemployment

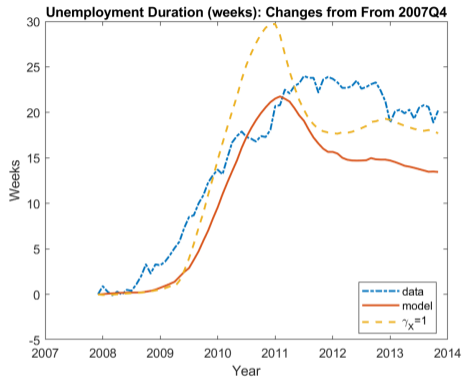
Unemployment



$\gamma_x = 1$: No duration dependence for former UI recipients

- Search inefficiency dampens the fluctuations in unemployment
- **Baseline model**: UI extensions account for 2pp \uparrow in unemployment
- **No search inefficiency**: overestimating the UI effect by 0.9pp ($\approx 50\%$)

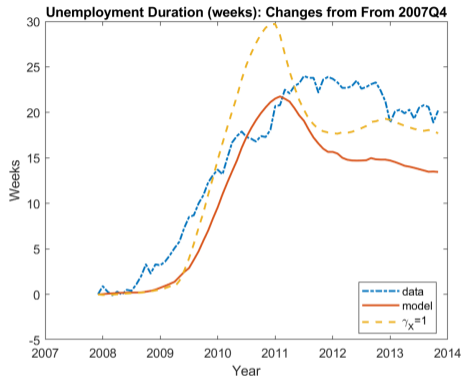
Average Unemployment Duration



$\gamma_x = 1$: No duration dependence for former UI recipients

- Search inefficiency dampens the response

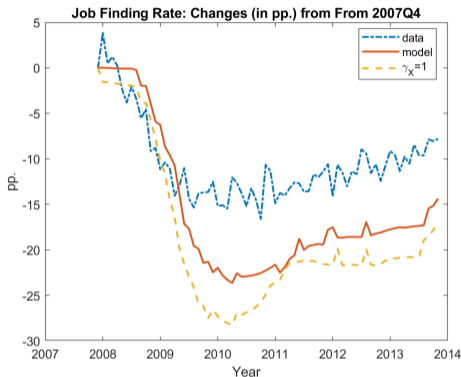
Average Unemployment Duration



$\gamma_x = 1$: No duration dependence for former UI recipients

- Search inefficiency dampens the response
- **Baseline**: UI extensions account for \uparrow 16 weeks in unemp. duration
- **No search inefficiency**: overestimating the UI effect by 9 wks ($\approx 56\%$)
- Average elasticity of unemp. duration to UI duration: **0.11** vs **0.15**

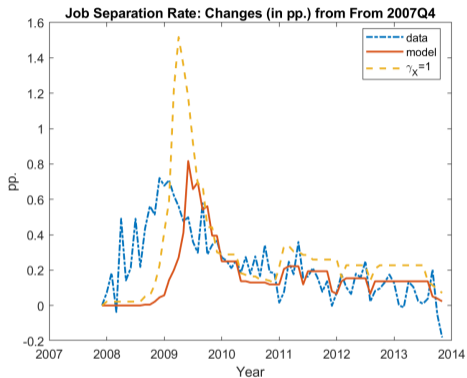
Job Findings



$\gamma_x = 1$: No duration dependence for former UI recipients

- **Baseline model** somewhat overshoots (negatively)
 - NB: negative trend in data
- But more moderate than a **model without inefficient search** where
 - No cost associated with being UI_x
 - UI workers reject even more jobs when UI is generous

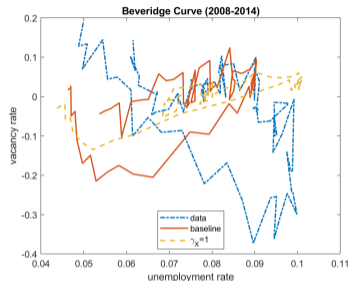
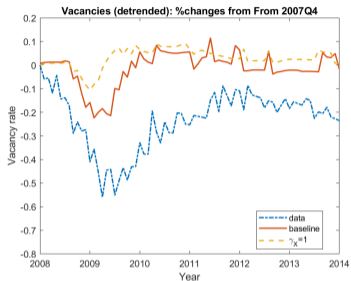
Job Separations



$\gamma_x = 1$: No duration dependence for former UI recipients

- **Baseline model:** realistic magnitude, despite lagged response
- **model without inefficient search:** heavily overestimating job separations (> 100%)

Vacancy rate and Beveridge Curve



$\gamma_x = 1$: No duration dependence for former UI recipients

- Vacancies less volatile than the data (50% for baseline and 10% for the alternative model)
- **Baseline** produces a negative correlation between u and v , whilst the model without inefficient search produces a positive one

Labour Market Persistence

Autocorrelation coefficients for 1-month lag

Moment	Data	Baseline	$\gamma_X = 1$
$\text{corr}(u, u_{-1})$	0.9921	(100%)	(99%)
$\text{corr}(\rho_{UE}, \rho_{UE, -1})$	0.9510	(103%)	(103%)
$\text{corr}(\rho_{EU}, \rho_{EU, -1})$	0.9616	(98%)	(98%)
$\text{corr}(udur, udur_{-1})$	0.9965	(100%)	(99%)

- ρ_{UE} : job finding rate // ρ_{EU} : job separation rate // $udur$: mean unemployment duration (weeks)
- $\gamma_X = 1$: No duration dependence for former UI recipients
- Little difference for 1-month lag

Labour Market Persistence

Autocorrelation coefficients for 12-month lag

Moment	Data	Baseline	$\gamma_X = 1$
$\text{corr}(u, u_{-12})$	0.7302	(82%)	(65%)
$\text{corr}(\rho_{UE}, \rho_{UE, -12})$	0.8119	(69%)	(60%)
$\text{corr}(\rho_{EU}, \rho_{EU, -12})$	0.7850	(30%)	(6%)
$\text{corr}(udur, udur_{-12})$	0.9220	(80%)	(69%)

- ρ_{UE} : job finding rate // ρ_{EU} : job separation rate // $udur$: mean unemployment duration (weeks)
- $\gamma_X = 1$: No duration dependence for former UI recipients
- Baseline noticeably performs better, particularly unemployment and unemployment duration

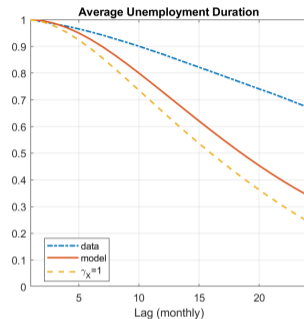
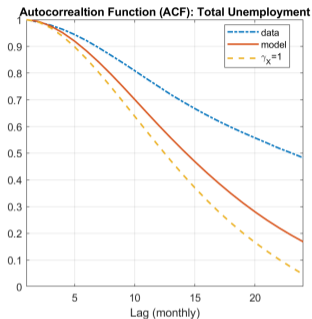
Labour Market Persistence

Autocorrelation coefficients for 24-month lag

Moment	Data	Baseline	$\gamma_X = 1$
$\text{corr}(u, u_{-24})$	0.4783	(38%)	(4%)
$\text{corr}(\rho_{UE}, \rho_{UE, -24})$	0.6468	(20%)	(6%)
$\text{corr}(\rho_{EU}, \rho_{EU, -24})$	0.7138	(8%)	(0.1%)
$\text{corr}(udur, udur_{-24})$	0.7909	(47%)	(30%)

- ρ_{UE} : job finding rate // ρ_{EU} : job separation rate // $udur$: mean unemployment duration (weeks)
- $\gamma_X = 1$: No duration dependence for former UI recipients
- Baseline fares much better especially for unemployment

Persistence of Unemployment



$\gamma_X = 1$: no duration dependence for former UI recipients

- Duration dependence improves the persistence of unemployment and its duration:
 - By construction + its dampening effects on current UI recipients
- A duration dependence model w/o UI status: Little persistence since all workers exit unemployment quickly from the start

Labour Market Fluctuations

Moment	Data	Baseline	$\gamma_X = 1$
$\text{std}(u)$	0.1454	0.1123 (77%)	0.1453 (99%)
$\text{std}(\rho_{UE})$	0.0999	0.1035 (103%)	0.1402 (140%)
$\text{std}(\rho_{EU})$	0.0890	0.0517 (58%)	0.0641 (72%)
$\text{std}(udur)$	6.9327	5.6412 (81%)	6.1954 (90%)

- ρ_{UE} : job finding rate // ρ_{EU} : job separation rate // $udur$: mean unemployment duration (weeks) // In parentheses are percentages w.r.t. empirical counterparts
- Labour market fluctuations dampened by 10-35% due to the duration dependence (esp. job finding rate)
 - ∴ In baseline, current UI workers less responsive to shocks + exit unemployment more quickly

What explains the re-employment wages?

Dependent variable: log re-employment weekly earnings		
	(1)	(2)
Current UI recipient	-0.057 (0.124)	-0.090 (0.124)
Former UI recipient	-0.478*** (0.148)	-0.589*** (0.154)
Search intensity	0.271 (0.356)	0.282 (0.353)
Unemp. duration (quartic)		✓
Controls for worker characteristics, time and state fixed effects, recession dummy, and state unemployment rates	✓	✓
<i>N</i>	303	303
<i>R</i> ²	0.411	0.411

- Linear probability model (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$)
- CPS monthly data and January supplements (1998–2022) excluding workers exiting the labour force next period
- Worker characteristics: race, education, gender, age (quartic), marital status, married-female, occupation, industry, recall expectation, search methods (-1), potential UI exhaustion month, reason for unemployment, last job tenure, last job's weekly earnings

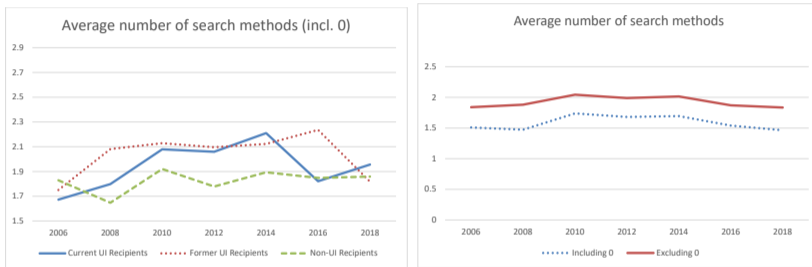
What explains job finding probability?

	Dependent variable: Job finding probability					
	(1)	(2)	(3)	(4)	(5)	(6)
Current UI recipient	-0.087*** (0.013)	-0.080*** (0.013)	-0.086*** (0.022)	-0.082*** (0.022)		
Total search activities (required by state)			0.000 (0.002)	0.000 (0.002)	-0.005*** (0.001)	-0.005*** (0.001)
Former UI recipient	-0.048*** (0.015)	-0.014 (0.016)	-0.047*** (0.015)	-0.014 (0.016)	-0.030** (0.015)	0.002 (0.015)
Search intensity	0.102* (0.058)	0.101* (0.057)	0.103* (0.058)	0.102* (0.057)	0.103* (0.058)	0.102* (0.057)
Unemployment duration (quartic)	NO	YES	NO	YES	NO	YES
<i>N</i>	5,693	5,693	5,693	5,693	5,693	5,693
<i>R</i> ²	0.064	0.076	0.068	0.079	0.066	0.077

- * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$.
- Other control variables include race, education, gender, age (quartic), marital status, female and married, occupation, industry, recall expectation, potential UI exhaustion month, reason for unemployment, previous job's tenure, previous job's weekly earnings, a linear time trend, a recession dummy, state fixed effects, and state unemployment rates.

[▶ Back to Linear Regression Model](#)

Data: Search Intensity (only active methods)

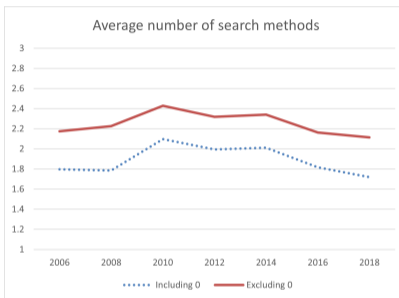
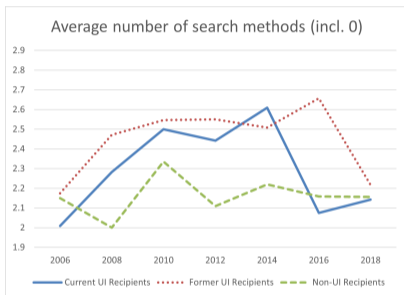


Non-UI recipients = those who never received UI

Source: CPS monthly data & January supplements

▶ [Back to average number of methods excl 0](#)

Data: Search Intensity

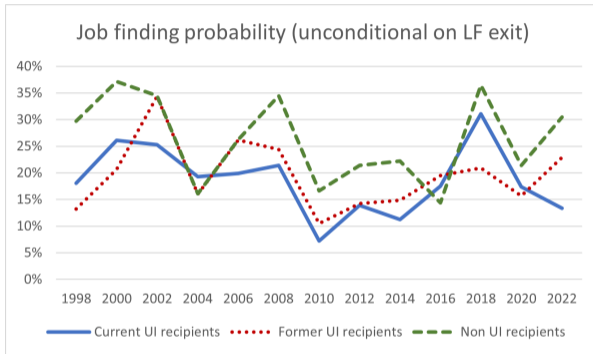


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▶ [Back to average number of methods excl 0](#)

Data: Job finding rate unconditional on LF exits

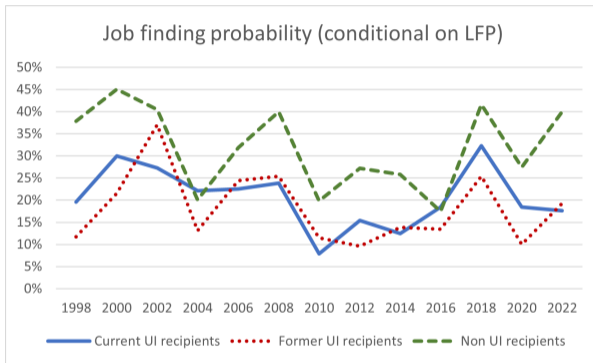


NB: Non-UI recipients = those who never received UI

Source: CPS monthly data & January supplements

▶ [Back to conditional UE rate](#)

Data: Job Finding Rate controlling for LTU for UIx workers



NB: Non-UI recipients = those who never received UI

Source: CPS monthly data & January supplements

▶ [Back to conditional UE rate](#)

Data: Job Finding Rate (uncond. on LFP)

Table: Dependent variable: Job Finding Probability

	(1)	(2)	(3)	(4)
Have received UI	-0.061*** (0.012)		-0.060*** (0.012)	
Current UI recipient		-0.076*** (0.012)		-0.078*** (0.012)
Former UI recipient		-0.024 (0.015)		-0.021 (0.015)
Long-term unemployed	-0.062*** (0.011)	-0.071*** (0.012)		
Unemp. duration (weeks)			-0.001*** (0.000)	-0.001*** (0.000)
Great Recession ('10-'12)	-0.064*** (0.011)	-0.059*** (0.011)	-0.063*** (0.011)	-0.056*** (0.011)
Worker characteristics	✓	✓	✓	✓
R^2	0.0775	0.0801	0.0784	0.0819

- Linear probability model (dependent variable = 1 if a worker moves from unemp. to emp., = 0 if remaining unemployed) (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$)
- CPS monthly data and January supplements: every 2 years between 2006-2018 (4,729 obs.)
- Worker characteristics include race, education, gender, age (quadratic), time trend (linear), occupation, industry, search method dummies
- Excluding 0 number of search methods

Data: Job Finding Rate

Table: Dependent variable: Job Finding Probability

	(1)	(2)	(3)	(4)
Number of search methods	-0.009* (0.005)	-0.009* (0.005)	-0.009* (0.005)	-0.009* (0.005)
Have received UI	-0.085*** (0.013)		-0.085*** (0.013)	
Current UI recipient		-0.104*** (0.014)		-0.105*** (0.014)
Former UI recipient		-0.034** (0.017)		-0.031* (0.018)
Long-term unemployed	-0.066*** (0.013)	-0.078*** (0.013)		
Unemp. duration (weeks)			-0.001*** (0.000)	-0.001*** (0.000)
Great Recession ('10-'12)	-0.077*** (0.013)	-0.070*** (0.013)	-0.075*** (0.013)	-0.067*** (0.013)
Worker characteristics	✓	✓	✓	✓
R^2	0.0775	0.0801	0.0784	0.0819

- Linear probability model (dependent variable = 1 if a worker moves from unemp. to emp., = 0 if remaining unemployed) (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$)
- CPS monthly data and January supplements: every 2 years between 2006-2018 (4,109 obs.)
- Worker characteristics include race, education, gender, age (quadratic), time trend (linear), occupation, industry
- Excluding 0 number of search methods and workers exiting the labour force next period

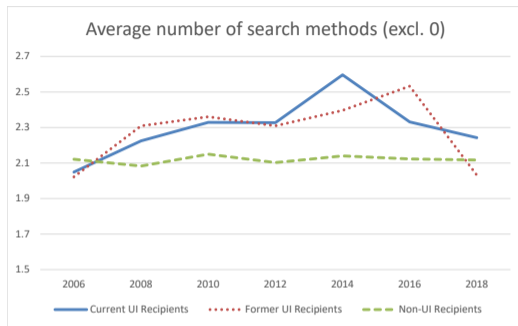
Data: UO Rate

Table: Dependent variable: UO Transition Rate (not filled)

	(1)	(2)	(3)	(4)
Have received UI	-0.061*** (0.012)		-0.060*** (0.012)	
Current UI recipient		-0.076*** (0.012)		-0.078*** (0.012)
Former UI recipient		-0.024 (0.015)		-0.021 (0.015)
Long-term unemployed	-0.062*** (0.011)	-0.071*** (0.012)		
Unemp. duration (weeks)			-0.001*** (0.000)	-0.001*** (0.000)
Great Recession ('10-'12)	-0.064*** (0.011)	-0.059*** (0.011)	-0.063*** (0.011)	-0.056*** (0.011)
Worker characteristics	✓	✓	✓	✓
R^2	0.0775	0.0801	0.0784	0.0819

- Linear probability model (dependent variable = 1 if a worker moves from unemp. to emp., = 0 if remaining unemployed) (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$)
- CPS monthly data and January supplements: every 2 years between 2006-2018 (4,109 obs.)
- Worker characteristics include race, education, gender, age (quadratic), time trend (linear), occupation, industry, search method dummies
- Excluding 0 number of search methods

Data: Search Intensity



NB: Non-UI recipients = those who never received UI

Source: CPS monthly data & January supplements

- Search intensity measured by the number of **active** job search methods (6 answers max.) of unemployed workers
- Excluding 0 search methods (workers expecting a recall)
- Current and former UI recipients generally search harder than those never received UI

Job Search Methods

- 1 Contacted employer directly/interview (*+)
- 2 Contacted public employment agency (*-)
- 3 Contacted private employment agency
- 4 Contacted friends or relatives
- 5 Contacted school/university empl center (*-)
- 6 Sent out resumes/filled out application (**-)
- 7 Checked union/professional registers
- 8 Placed or answered ads
- 9 Other active
- 10 Looked at ads
- 11 Attended job training programs/courses
- 12 Nothing
- 13 Other passive

Job Search Methods: Current UI Workers

- 1 Contacted employer directly/interview
- 2 Contacted public employment agency
- 3 Contacted private employment agency
- 4 Contacted friends or relatives
- 5 Contacted school/university empl center
- 6 Sent out resumes/filled out application (*-)
- 7 Checked union/professional registers
- 8 Placed or answered ads
- 9 Other active
- 10 Looked at ads
- 11 Attended job training programs/courses
- 12 Nothing
- 13 Other passive

▶ Back to average number of methods excl 0

Job Search Methods: Exhausted UI Workers

- ① Contacted employer directly/interview
- ② Contacted public employment agency
- ③ Contacted private employment agency (**+)
- ④ Contacted friends or relatives
- ⑤ Contacted school/university empl center
- ⑥ Sent out resumes/filled out application
- ⑦ Checked union/professional registers (**-)
- ⑧ Placed or answered ads
- ⑨ Other active
- ⑩ Looked at ads
- ⑪ Attended job training programs/courses
- ⑫ Nothing
- ⑬ Other passive

▶ Back to average number of methods excl 0

Job Search Methods: Non UI Workers

- 1 Contacted employer directly/interview
- 2 Contacted public employment agency (*-)
- 3 Contacted private employment agency
- 4 Contacted friends or relatives (*-)
- 5 Contacted school/university empl center
- 6 Sent out resumes/filled out application
- 7 Checked union/professional registers
- 8 Placed or answered ads
- 9 Other active
- 10 Looked at ads
- 11 Attended job training programs/courses
- 12 Nothing
- 13 Other passive

▶ Back to average number of methods excl 0

Stochastic Model: Value functions: Firms

- The value of a matched firm with a worker whose work history is $j \in \{E(\tilde{m}), Bs(\tilde{m}), Bl(\tilde{m}), X, N\}$ is

$$\begin{aligned} J_j(m; \omega) = & y(m; \omega) - w_j(m; \omega) - \tau(\omega) + \beta E_{\omega'|\omega} \left[\dots \right. \\ & (1 - \delta)(1 - \lambda) \left[(1 - p_E(m; \omega)(1 - F(m))) J_{E(m)+}(m; \omega') \right] \\ & + (1 - \delta)\lambda E_{m'} \left[(1 - p_E(m; \omega)(1 - F(m'))) J_{E(m)+}(m'; \omega') \right] \\ & \left. + \delta V(\omega') \right] \end{aligned}$$

- $J_{E(m)+}(m'; \omega') \equiv \max\{J_{E(m)}(m'; \omega'), V(\omega')\}$

◀ Back to value function for the employed

Stochastic Model: Value Functions: Unemployed

- Define ω as the set of state variables
- For $j \in \{X, N\}$:

$$U_j(\omega) = \max_{s_j(\omega)} h - c_j(s_j(\omega)) + \beta E_{m' \omega' | \omega} \left[p_j(\omega) \underbrace{\max \{W_j(m'; \omega'), U_j(\omega')\}}_{\text{job formation/rejection decision}} + (1 - p_j(\omega)) U_j(\omega') \right]$$

▶ Back to policy functions

Stochastic Model: Value Functions: Unemployed

- Define ω as the set of state variables
- For $B\ell$ -workers with benefit $b(\tilde{m})$ and long unemp. duration:

$$\begin{aligned}
 U_{B\ell}(\tilde{m}; \omega) &= \max_{\gamma_{B\ell}(\tilde{m}; \omega)} b(\tilde{m}) + h - c(\gamma_{B\ell}(\tilde{m}; \omega) \underline{s}_B) \\
 &+ \beta E_{m' \omega' | \omega} \left[p_{B\ell}(\tilde{m}; \omega) \max \left\{ W_{B\ell}(\tilde{m})(m'; \omega'), \dots \right. \right. \\
 &\quad \left. \left. \underbrace{(1 - \phi(u))(1 - \xi)}_{\text{keep UI | meeting a firm}} U_{B\ell}(\tilde{m}; \omega') + \underbrace{(\phi(u) + (1 - \phi(u))\xi)}_{\text{lose UI | meeting a firm}} U_X(\omega') \right\} \right. \\
 &\quad \left. + (1 - p_{UI\ell}(\tilde{m}; \omega)) \left((1 - \phi(u)) U_{B\ell}(\tilde{m}; \omega') + \phi(u) U_X(\omega') \right) \right]
 \end{aligned}$$

$\xi \equiv \text{Pr}(\text{losing UI eligibility} \mid \text{rejecting job offers})$

► Back to policy functions

Stochastic Model: Value Functions: Unemployed

- Define ω as the set of state variables
- For Bs -workers with benefit $b(\tilde{m})$ and short unemp. duration:

$$\begin{aligned}
 U_{Bs}(\tilde{m}; \omega) = & \max_{\gamma_{Bs}(\tilde{m}; \omega)} b(\tilde{m}) + h - c(\gamma_{Bs}(\tilde{m}; \omega) \underline{s}_B) \\
 & + \beta E_{m' \omega' | \omega} \left[p_{Bs}(\tilde{m}; \omega) \max \left\{ W_{Bs}(\tilde{m})(m'; \omega'), \dots \right. \right. \\
 & \underbrace{(1 - \phi(u))(1 - \xi)}_{\text{keep UI | meeting a firm}} \left. \left. ((1 - \gamma_\ell) U_{Bs}(\tilde{m}; \omega') + \gamma_\ell U_{Be}(\tilde{m}; \omega')) + \dots \right. \right. \\
 & \underbrace{(\phi(u) + (1 - \phi(u))\xi)}_{\text{lose UI | meeting a firm}} \left. \left. U_X(\omega') \right\} \right] \\
 & + (1 - p_{Bs}(\tilde{m}; \omega)) \times \left(\dots \right. \\
 & \left. (1 - \phi(u)) \left((1 - \gamma_\ell) U_{Bs}(\tilde{m}; \omega') + \gamma_\ell U_{Be}(\tilde{m}; \omega') \right) + \phi(u) U_X(\omega') \right) \left. \right]
 \end{aligned}$$

$\xi \equiv \text{Pr}(\text{losing UI eligibility} \mid \text{rejecting job offers})$

▶ Back to policy functions

Stochastic Model: Value Function: Employed

- The value of a type- i employed worker with last period's employment status and associated benefit level $j \in \{E(\tilde{m}), Bs(\tilde{m}), Bl(\tilde{m}), X, N\}$:

$$\begin{aligned} W_j(m; \omega) = & \max_{s_E(m; \omega)} w_j(m; \omega) - c_E(s_E(m; \omega)) + \beta E_{\omega' | \omega} \left[\right. \\ & (1 - \delta)(1 - \lambda) \left[(1 - p_E(m; \omega)(1 - F(m))) W_{E(m)+}(m; \omega') \right. \\ & \left. \left. + p_E(m; \omega)(1 - F(m)) E_{m' | m' > m} [W_{E(m)+}(m'; \omega')] \right] \right. \\ & \left. + (1 - \delta)\lambda E_{m'} \left[(1 - p_E(m; \omega)(1 - F(m'))) W_{E(m)+}(m'; \omega') \right. \right. \\ & \left. \left. + p_E(m; \omega)(1 - F(m')) E_{m'' | m'' > m'} [W_{E(m)+}(m''; \omega')] \right] \right. \\ & \left. + \delta \left((1 - \psi) U_{Bs}(m, \omega') + \psi U_N(\omega') \right) \right] \end{aligned}$$

$$W_{E(m)+}(m'; \omega') \equiv \max\{W_{E(m)}(m'; \omega'), (1 - \psi) U_{Bs}(m, \omega') + \psi U_N(\omega')\}$$

Stochastic Model: Free Entry

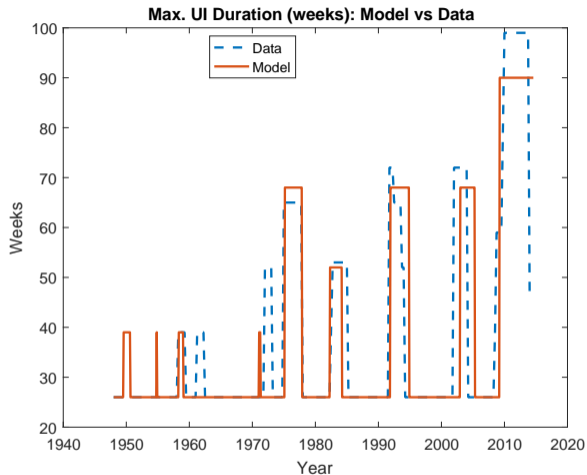
- The value of an unmatched firm is

$$\begin{aligned}
 V(\omega) = & -\kappa + \beta q(\omega) E_{\omega'|\omega} \left[\dots \right. \\
 & \sum_m \zeta_E(m; \omega) (1 - F(m)) E_{m'|m' > m} [J^{E(m)+}(m'; \omega')] \\
 & + \sum_m \zeta_B(m, \omega) E_{m'} [J^{UI(m)+}(m'; \omega')] \\
 & \left. + \zeta_N(\omega) E_{m'} [J^{UU+}(m'; \omega')] + \zeta_X(\omega) E_{m'} [J^{UX+}(m'; \omega')] \right]
 \end{aligned}$$

$$\begin{aligned}
 \text{where } \zeta_E(m) &= \frac{(1 - \lambda) s_E(m) E(m) + \lambda f(m) \sum_m s_E(m) E(m)}{s} \\
 \zeta_B(m) &= \frac{\gamma_B(m) \underline{s}_B U_B(m)}{s}; \quad \zeta_N = \frac{s_N U_N}{s}; \quad \zeta_X = \frac{s_X U_X}{s} \\
 s &= \sum_m \left((1 - \lambda) s_E(m) E(m) + \lambda f(m) \sum_m s_E(m) E(m) + \gamma_B(m) \underline{s}_B U_B(m) \right) \\
 &+ s_N U_N + s_X U_X
 \end{aligned}$$

- Free entry condition implies $V(\omega) = 0, \forall \omega$

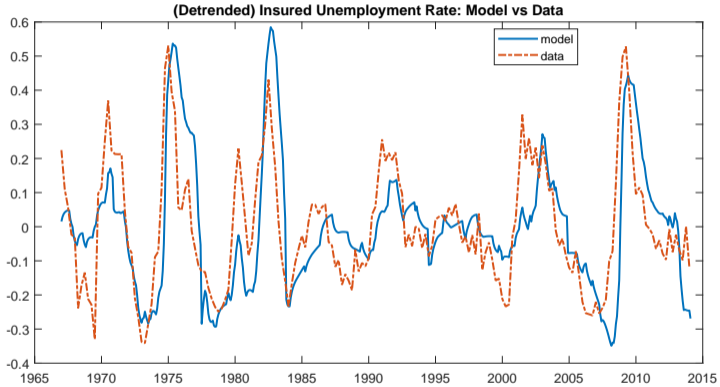
UI Extensions



- UI extension **triggered by high unemployment** rate above 6%
- The Great Recession: Emergency Unemployment Compensation (EUC08) + Extended Benefits programmes

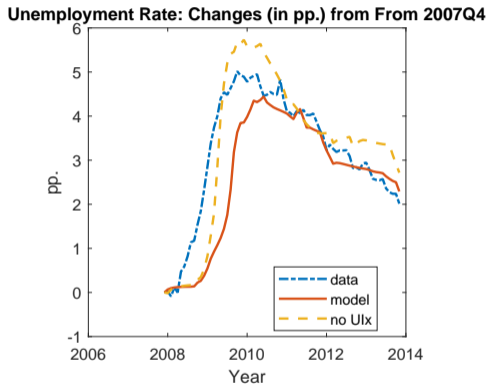
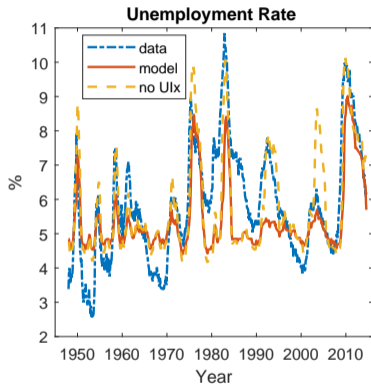
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Insured Unemployment Rate



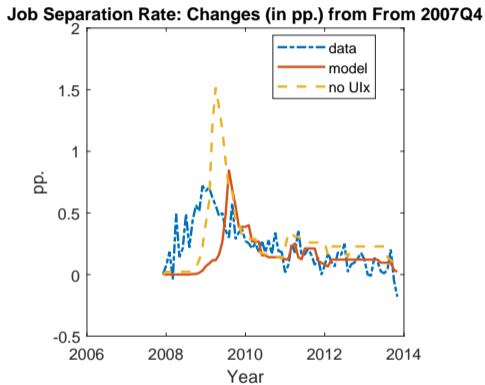
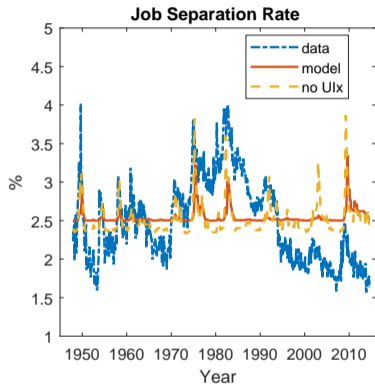
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Unemployment: Longer Horizon



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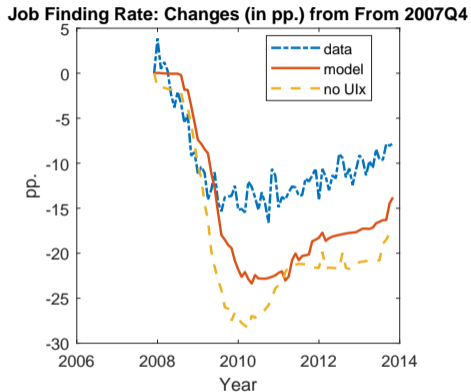
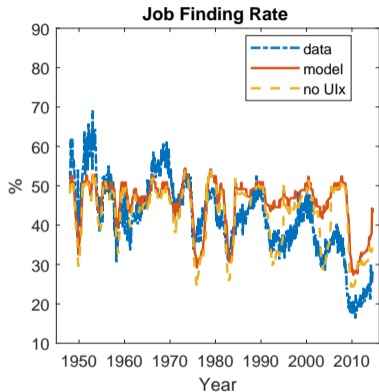
Job Separations: Longer Horizon



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Job Finding: Longer Horizon



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Fitted parameters (monthly)

Parameter	Description	Value
α	Matching function	0.51
δ	Exogenous separation	0.023
λ	Redrawing new m	0.50
ψ	Losing UI after becoming unemp.	0.49
ξ	Losing UI after meeting firm	0.50
a_E	Search cost function	0.15
a_{UIx}	Search cost function	0.0717
\underline{m}	Lowest match-specific prod.	0.396
β_1	Match-specific prod. distribution	2.55
β_2	Match-specific prod. distribution ▶ Dist. of m	5.26
ρ_z	Persistence of TFP	0.9562
σ_z	Standard deviation of TFP shocks	0.0075

Moments to match

- Mean and SD of unemployment, UE and EU transition rates, and average unemployment duration
- Persistence and SD of labour productivity, correlation between output and labour productivity, mean insured unemployment, and mean job-to-job transition rate
- $M(s, v) = sv / (s^\alpha + v^\alpha)^{1/\alpha}$

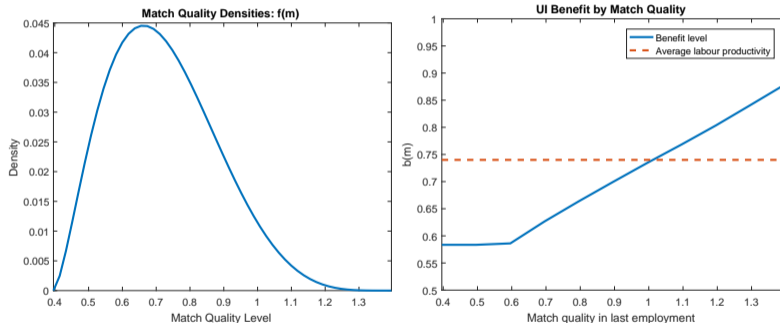
Fixed parameters (monthly)

- $b(m)$ and h imply consumption drop of 24% for U_{UU} and 10% for U_{UI} at 50% replacement rate (Gruber 1997)

Parameter	Description	Value	Source/Remarks
β	Discount factor	0.9967	Annual interest rate of 4%
κ	Vacancy posting cost	0.0392	Fujita & Ramey (2012)
μ	Worker's bargaining power	0.5	Den Haan, Ramey & Watson (2000)
ϕ	UI exhaustion rate	1/6	6 months max UI duration, ETA
ϕ_{L1}	UI exhaustion rate	1/9	9 months max UI duration, ETA
ϕ_{L2}	UI exhaustion rate	1/12	12 months max UI duration, ETA
ϕ_{L3}	UI exhaustion rate	1/16	16 months max UI duration, ETA
ϕ_{L4}	UI exhaustion rate	1/21	21 months max UI duration, ETA
\bar{u}	UI policy threshold	0.06	ETA
a_U	Search cost function	0.1116	Normalisation
d_u, d_e	Search cost function	1	Christensen et al. (2004), Yashiv (2000)
h	Leisure flow	0.5835	Gruber (1997)
γ_{UIx}	Search inefficiency of UIx	0.72	Relative search intensity of UIx wrt UU , CPS

Benefit Levels and Match Quality

- The distribution of match quality is crucial for the heterogeneity of the unemployment exit rates amongst U_{UI}



► Back to calibration

What explains search intensity?

Dependent variable: Job search intensity		
	Number of methods	Imputed minutes
Recall expected	-2.486*** (0.046)	-35.528*** (1.032)
Male	0.015 (0.044)	10.536*** (0.985)
Married female	-0.147** (0.059)	-24.939*** (1.325)
State unemp. rate	0.037*** (0.013)	0.530* (0.283)
dur	0.012* (0.007)	0.337** (0.158)
dur2	-0.000 (0.000)	-0.013** (0.006)
dur3	0.000 (0.000)	0.000** (0.000)
dur4	-0.000 (0.000)	-0.000** (0.000)
R^2	0.354	0.486

- Linear regression coefficients (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$), $N = 7,561$
- CPS monthly data and January supplements (1998-2022) [▶ Back to intro: search intensity](#)
- Worker characteristics include: race, education, gender, age (quartic), marital status, married-female, occupation, industry, recall expectation, potential UI exhaustion month, reason for unemployment, last job tenure, last job's weekly earnings

What explains search intensity?

Dependent variable: Job search intensity		
	Number of methods	Imputed minutes
Potential UI exhaustion month	0.022 (0.055)	0.142 (1.238)
High School	0.120*** (0.046)	6.411*** (1.034)
Some College	0.219*** (0.049)	21.165*** (1.105)
College	0.386*** (0.059)	48.213*** (1.332)
Last job's weekly earnings (Unit: USD1000)	0.001** (0.000)	0.017*** (0.009)
R^2	0.354	0.451

- Linear regression coefficients (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$). $N = 7,561$. CPS monthly data and January supplements (1998-2022). Excluding 0 search methods (those on recalls) [▶ Back to OLS](#)
- Worker characteristics include: race, education, gender, age (quartic), marital status, married-female, occupation, industry, recall expectation, potential UI exhaustion month, reason for unemployment, last job tenure, last job's weekly earnings

Summary statistics by UI history

	UI receivers	UI exhausters	Non-UI receivers
Share of total unemp.	39.21%	20.51%	40.27%
Age (year)	42.68	44.49	38.27
Female	39.56%	41.65%	40.95%
Married female	20.54%	20.63%	17.80%
Black	10.02%	15.99%	16.88%
<High school	10.59%	11.61%	19.67%
High school	34.20%	36.30%	36.12%
Some College	22.26%	21.15%	19.28%
College and more	32.95%	30.95%	24.93%
Job loser/on lay off	14.10%	8.96%	13.46%
Other job loser	69.51%	60.93%	43.25%
Temp job ended	9.34%	11.86%	16.42%
Job leaver	1.75%	2.71%	8.05%
Re-entrant	5.26%	15.41%	18.65%
New entrant	0.03%	0.13%	0.16%
Unemp. Duration (weeks)	23.3	42.26	19.4
Fraction of LTU	31.97%	63.83%	24.96%
Given/expecting a recall	14.10%	8.96%	13.46%
Last job's tenure (year)	5.72	5.53	3.15
Last job's weekly earning	854.08	789.31	589.18

- CPS monthly data and January supplements: every 2 years between 1998-2022 (7561 obs.)

[◀ Back to intro](#)

Model ► FOCs

◀ Back to Model: Equilibrium

$$\begin{aligned} c'_E(s_{E(\tilde{m})}(m; \omega)) &= -\beta(1 - \delta)M(\theta(\omega))E_{\omega'|\omega} \left[\dots \right. \\ &\quad \left. (1 - \lambda)(1 - F(m)) \left(WS_{E(m)+}(m; \omega') - E_{m'|m'>m} [WS_{E(m)+}(m'; \omega')] \right) \right. \\ &\quad \left. + \lambda E_{m'} \left[(1 - F(m')) \left(WS_{E(m)+}(m'; \omega') - E_{m''|m''>m'} [WS_{E(m)+}(m''; \omega')] \right) \right] \right] \end{aligned}$$

$$\begin{aligned} c'(\gamma_{B(m)}(\omega)) &= \beta M(\theta(\omega)) E_{m'\omega'|\omega} \left[\max\{WS_{B(m)}(m'; \omega'), 0\} \dots \right. \\ &\quad \left. - \xi(1 - \phi)US(m, \omega') \right] \end{aligned}$$

$$c'(s_N(\omega)) = \beta M(\theta(\omega)) E_{m'\omega'|\omega} \left[\max\{WS_N(m'; \omega'), 0\} \right]$$

$$c'_X(s_X(\omega)) = \beta M(\theta(\omega)) E_{m'\omega'|\omega} \left[\max\{WS_X(m'; \omega'), 0\} \right]$$

Dependent variable: Job Finding Probability

	(1)	(2)	(3)	(4)
Have received UI	-0.082*** (0.013)		-0.067*** (0.013)	
Current UI recipient		-0.100*** (0.014)		-0.087*** (0.014)
Former UI recipient		-0.035** (0.018)		-0.019 (0.018)
Long-term unemployed	-0.064*** (0.013)	-0.076*** (0.013)		
U duration			-0.015*** (0.003)	-0.014*** (0.003)
U duration ²			0.000*** (0.000)	0.000*** (0.000)
U duration ³			-0.000*** (0.000)	-0.000*** (0.000)
U duration ⁴			0.000*** (0.000)	0.000** (0.000)
Great Recession (2010-2012)	-0.077*** (0.013)	-0.070*** (0.013)	-0.069*** (0.013)	-0.061*** (0.013)
R ²	0.0784	0.0819	0.0775	0.0801

- Linear probability model (dependent variable = 1 if a worker moves from unemp. to emp., = 0 if remaining unemployed) (* p<0.05, ** p<0.01, *** p<0.001)
- CPS monthly data and January supplements: every 2 years between 2006-2018 (4,109 obs.)
- Other regressors include: worker characteristics (race, education, gender, age (quartic), occupation, industry, recall expectation, search methods), linear time trend and state FE
- Excluding zero search methods and workers exiting the labour force next period

Dependent variable: Job Finding Probability

	(1)	(2)	(3)	(4)
Have received UI	-0.082*** (0.013)		-0.067*** (0.013)	
Current UI recipient		-0.100*** (0.014)		-0.087*** (0.014)
Former UI recipient		-0.035** (0.018)		-0.019 (0.018)
Long-term unemployed	-0.064*** (0.013)	-0.076*** (0.013)		
U duration			-0.015*** (0.003)	-0.014*** (0.003)
U duration ²			0.000*** (0.000)	0.000*** (0.000)
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Great Recession (2010-2012)	-0.077*** (0.013)	-0.070*** (0.013)	-0.069*** (0.013)	-0.061*** (0.013)
R ²	0.0784	0.0819	0.0775	0.0801

- Linear probability model (dependent variable = 1 if a worker moves from unemp. to emp., = 0 if remaining unemployed) (* p<0.05, ** p<0.01, *** p<0.001)
- CPS monthly data and January supplements: every 2 years between 2006-2018 (4,109 obs.)
- Other regressors include: worker characteristics (race, education, gender, age (quartic), occupation, industry, recall expectation, search methods), linear time trend and state FE
- Excluding zero search methods and workers exiting the labour force next period

Dependent variable: Job Finding Probability

	(1)	(2)	(3)	(4)
Have received UI	-0.082*** (0.013)		-0.067*** (0.013)	
Current UI recipient		-0.100*** (0.014)		-0.087*** (0.014)
Former UI recipient		-0.035** (0.018)		-0.019 (0.018)
Long-term unemployed	-0.064*** (0.013)	-0.076*** (0.013)		
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Model ► Transitions I: Unemployment

◀ Back to Model: Equilibrium

$$U_{B,t+1}(m) = \underbrace{(1 - \psi)\rho_{eu,t}(m)e_t^{post}(m)}_{\text{separated match, not losing UI}} + \underbrace{(1 - \phi_t)(1 - p_{B,t}(m))U_{B,t}(m)}_{\text{no meeting, not losing UI}} + \underbrace{\chi_{B,t}(m)(1 - \phi_t)(1 - \xi)p_{B,t}(m)U_{B,t}(m)}_{\text{bad meeting, not losing UI}} \quad (1)$$

$\chi_{B,t}(\tilde{m})$ is the rate the meetings with $U_B(\tilde{m})$ are unproductive. Namely,
 $\chi_{B,t}(\tilde{m}) = \int \mathbf{1}\{S_{B(\tilde{m})t}(m) \leq 0\}f(m)dm.$

Model ► Transitions I: Unemployment (Cont.)

◀ Back to Model: Equilibrium

$$\begin{aligned}
 U_{N,t+1} &= \underbrace{\phi_t(1 - p_{B,t}(m))U_{UI,t}(m)}_{\text{no meeting, losing UI}} + \underbrace{\chi_{UI,t}(m)\left(\phi_t + (1 - \phi_t)\xi\right)p_{B,t}(m)U_{B,t}(m)}_{\text{bad meeting, losing UI}} \\
 &\quad + (1 - \rho_{ue,tN})U_{N,t} + \underbrace{\psi\rho_{eu,t}e_t^{post}}_{\text{separated match, losing UI}} \\
 u_{t+1} &= \sum_m U_{B,t+1}(m) + U_{N,t+1}
 \end{aligned}$$

$\chi_{B,t}(\tilde{m})$ is the rate the meetings with $U_B(\tilde{m})$ are unproductive. Namely,
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Model ► Transitions II: Employment

◀ Back to Model: Equilibrium

$$\begin{aligned} e_{m,it+1} = & \left((1 - \delta)(1 - \lambda)(1 - p_{Em,it} + p_{Em,it}F(m))e_{m,it} \right. \\ & + (1 - \delta)(1 - \lambda)f(m) \sum_{m' < m} p_{Em',it}e_{m',it} \\ & + (1 - \delta)\lambda f(m) \sum_{m'} (1 - p_{Em',it} + p_{Em',it}F(m))e_{m',it} \\ & \left. + (1 - \delta)\lambda \left(\sum_{m' < m} f(m') \right) f(m) \sum_{m'} p_{Em',it}e_{m',it} \right) \mathbf{1}\{S_{Em,\omega_{it+1}} > 0\} \\ & + \sum_{\tilde{m}} f(m)(U_{B,t}(\tilde{m})p_{B,t}(\tilde{m})) \mathbf{1}\{S_{m,\omega_{t+1}}^{B(\tilde{m})} > 0\} \\ & + f(m)(U_{N,t}p_{N,t}) \mathbf{1}\{s_{Nm,\omega_{t+1}} > 0\} \end{aligned}$$

The sum in the 2nd and 4th lines is over employed workers with match qualities worse than m who hence make a job-to-job transition.

Model ► Transitions III: Job Separation Rate

◀ Back to Model: Equilibrium

$$\rho_{eu,t}(m) = \begin{cases} \delta & \text{if } S_{Em,\omega_{t+1}} > 0, \\ 1 & \text{otherwise} \end{cases}$$

$$\rho_{eu,t} = \frac{\delta \int_{\{m: S_{Em,t+1} > 0\}} e_{m,t}^{post} dm + \int_{\{m: S_{Em,t+1} \leq 0\}} e_{m,t}^{post} dm}{e_t}$$

$$(\equiv \rho_{eu,t}^{exo} + \rho_{eu,t}^{endo})$$

where $e_{m,t}^{post} \equiv$

$$\begin{aligned} & (1 - \lambda)(1 - p_{Em,t} + p_{Em,t}F(m))e_{m,t} \\ & + (1 - \lambda)f(m) \sum_{m' < m} p_{Em',t} e_{m',t} \\ & + \lambda f(m) \sum_{m'} (1 - p_{Em',t} + p_{Em',t}F(m))e_{m',t} \\ & + \lambda \left(\sum_{m' < m} f(m') \right) f(m) \sum_{m'} p_{Em',t} e_{m',t} \end{aligned}$$

Model ► Transitions IV: Job Finding Rate

◀ Back to Model: Equilibrium

For $j = \{B(\tilde{m}), N\}$

$$\rho_{j,t}^{ue} = \int \rho_{j,t}^{ue}(m) f(m) dm$$
$$\rho_t^{ue} = \frac{\sum_{\tilde{m}} (u_{B(\tilde{m}),t} \rho_{B(\tilde{m}),t}^{ue}) + u_{N,t} \rho_{N,t}^{ue}}{u_{UI,t} + U_{N,t}}$$

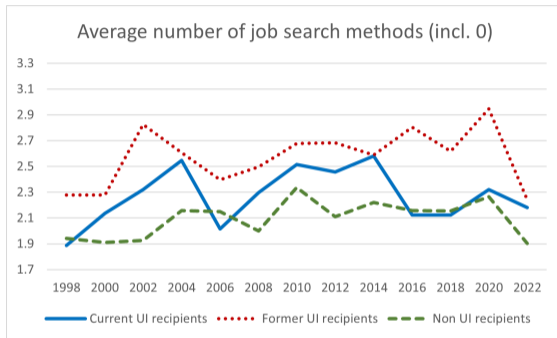
$$\text{where } \rho_{j,t}^{ue}(m) = \begin{cases} p_{\omega_t}^j & \text{if } S_{m',\omega_{t+1}}^j > 0, \\ 0 & \text{otherwise} \end{cases}$$

Job-to-job transition rate is

$$\rho_{m,t}^{ee} = (1 - \delta) \left((1 - \lambda) p_{m,t}^E (1 - F(m)) E_{m' > m} [\mathbf{1}\{S_{m',\omega_{t+1}}^{e(m)} > 0\}] \right. \\ \left. + \lambda \sum_{m'} p_{m,t}^E f(m') (1 - F(m')) E_{m'' > m'} [\mathbf{1}\{S_{m'',\omega_{t+1}}^{e(m)} > 0\}] \right)$$
$$\rho_t^{ee} = \frac{\sum_m \rho_{m,t}^{ee} e_{m,t}}{e_t}$$

Data: Search Intensity

≡ number of job search methods used



NB: Non-UI recipients = those who never received UI

Source: CPS monthly data & January supplements

- **Current** and **former UI recipients** search harder than **those never received UI** (respectively **8%** and **23%** more)

Stochastic Model: UI Extensions

State-dependent UI extension:

- via UI exhaustion rate, $Pr(B \rightarrow X)$, as a function of **unemployment**

$$\phi(u_t) = \begin{cases} \phi & \text{when } u_t < \bar{u} \\ \phi_L & \text{when } u_t \geq \bar{u} \end{cases}$$

- Assume $\phi_L < \phi \Leftrightarrow \frac{1}{\phi_L} > \frac{1}{\phi}$
- Recall $\frac{1}{\phi(u_t)}$ = expected UI duration
- \bar{u} = threshold above which UI extension is triggered
- Consistent with UI extension policies in the US (countercyclical)
- Due to UI extensions, some *UI* workers can be long-term unemployed $\rightarrow \{Bs, Bl\}$

Stochastic Model: Policy Functions

Joint worker-firm (total match) surplus varies with

- $m(+)$: match-specific productivity
- $z(+)$: total factor productivity
- $u(-)$: unemployment rate (UI extension policy)
+ cross-sectional distribution of workers (to predict u')

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Search intensity/censoring varies with

- $z(+)$: total factor productivity
- $u(-)$: unemployment rate (UI extension policy - only UI workers)
+ cross-sectional distribution of workers (to predict u')
- $b(m)(-)$: UI benefit amount (only B -type workers)
- $m(-)$: match-specific productivity (only E workers)

Optimal Search/Job Censoring

Search inefficiency discourages workers from searching harder:

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UI workers' optimal search censoring (γ_B):

- Compare the job finding rate given \underline{s}_B vs. optimal s_B^*
 - s_B^* : what current UI workers would have optimally exerted
- $\gamma_B^* \cdot \underline{s}_B \cdot p \equiv s_B^* \cdot p$

▶ Back to the Bellman Equations

Optimal Search: Quadratic search cost

Assume $c_j(s) = a_j \cdot s^2$ à la Christensen et al. (2015)

Optimal search intensity: FOCs:

$$\frac{\partial c(s_N)}{\partial s_N} = 2 \cdot a \cdot s_N = \beta \cdot p \cdot (W_N - U_N)$$
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- Assume the limiting case: $\lim_{\delta \rightarrow 0} \frac{W_X - U_X}{W_N - U_N} = 1$ (generally < 1)

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$$\rightarrow \frac{a_X}{a} = \gamma_X < 1$$

- But since empirically $s_X > s_N$

$$\rightarrow \frac{a_X}{a} < \gamma_X < 1$$

Wage

- Wages are negotiated using a generalised Nash bargaining rule
- Workers with previous employment status $j \in \{E, B, X, N\}$ receive

$$w_j = \operatorname{argmax} \left(WS_j \right)^\mu \left(J_j \right)^{(1-\mu)}$$

- μ = worker's bargaining power
- WS_j = Surplus of employed workers with history j
- Total match surplus $S_j \equiv WS_j + J_j$

→ $WS_j = \mu S_j$ and $J_j = (1 - \mu) S_j$

What explains search intensity?

Dependent variable: (1) Number of job search methods (2) Imputed minutes of job search à la Mukoyama et al. (2018)					
Strict UI search requirements	(1)	(2)	Lenient UI search requirements	(1)	(2)
Florida	0.263*	3.070	California	-0.100	-2.062
	(0.142)	(3.198)		(0.130)	(2.919)
Missouri	0.200	2.984	Delaware	-0.249	-3.880
	(0.158)	(3.560)		(0.198)	(4.447)
Nebraska	0.237	8.188*	Massachusetts	-0.011	-4.011
	(0.200)	(4.492)		(0.167)	(3.760)
New Hampshire	0.287*	4.134	South Carolina	-0.296*	-4.043
	(0.165)	(3.700)		(0.171)	(3.828)
Utah	0.324*	0.410	West Virginia	-0.408**	-7.612*
	(0.191)	(4.286)		(0.174)	(3.911)
<i>N</i>	7,561	7,561	<i>N</i>	7,561	7,561
<i>R</i> ²	0.354	0.486	<i>R</i> ²	0.354	0.486

- Linear regression coefficients (* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$)
- Alabama is the reference state (3 work search contacts weekly)
- CPS monthly data and January supplements (1998-2022), excluding 0 search methods (those on recalls)
- Worker characteristics include: UI status, race, education, gender, age (quartic), marital status, occupation, industry, recall expectation, potential UI exhaustion month, reason for unemployment, last job tenure, last job's weekly earnings