

# A Congestion Theory of Unemployment Fluctuations

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LAEF

Labor Markets and Macroeconomic Outcomes

Why is job creation so unattractive in recessions?

We propose a **congestion** theory of unemployment fluctuations, based on two facts:

1. in recessions, *more* unemployed find jobs
2. limited capacity of firms to absorb increases in unemployment by expanding hiring
  - Congestion: diminishing returns in new hires' jobs (or convex hiring costs)

⇒ **Countercyclical congestion** (extra procyclicality of productivity in new jobs)

Provides a unified explanation for a range of macroeconomic patterns:

- strong amplification and propagation in the labor market
- relative cyclicity of new-hire wages (alternative calibration target)
- countercyclical labor wedge
- countercyclical earnings losses from job displacement and labor market entry
- ... all while featuring realistic insensitivity of hiring to policy changes such as UI

# Outline

## 1. Empirical evidence

- countercyclical shift of employment towards recently unemployed workers
- congestion in hiring

## 2. Model structure

- congestion mechanism
- embed in a standard DMP model
- calibrate to IRF of  $u$  and/or new hires' relative wage cyclical

## 3. Business cycle performance

- volatility and comovement of labor market variables
- estimate congestion unemployment

## 4. Three additional macroeconomic applications

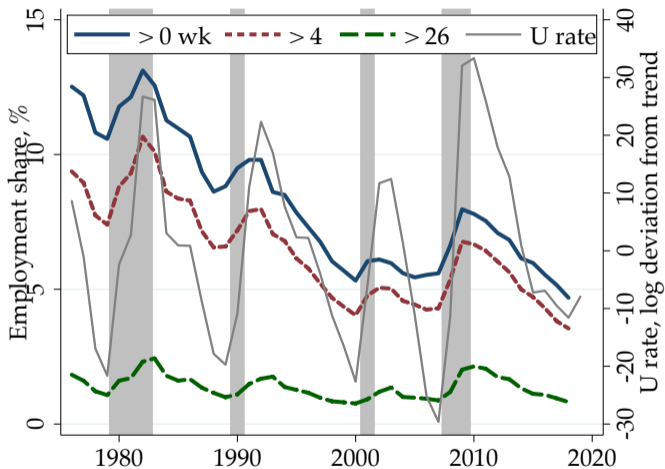
- labor wedge, earnings losses, sensitivity to policy changes

## 1. Empirical evidence

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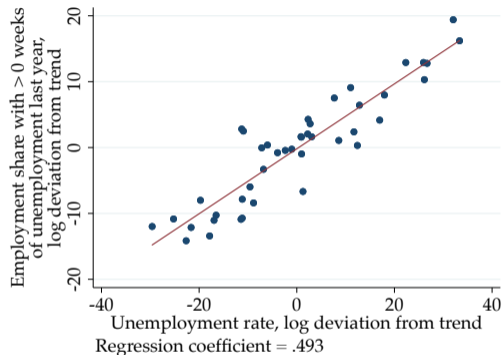
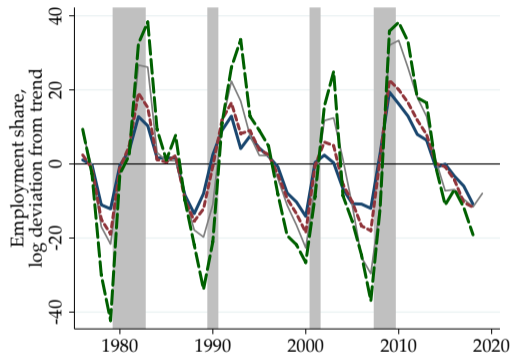
*Countercyclical hiring out of unemployment and congestion in hiring*

# 1. Employment share of workers with recent unemployment



CPS-ASEC (1976-2019): # of weeks workers spent in unemployment in previous year

# 1. Employment share of workers with recent unemployment

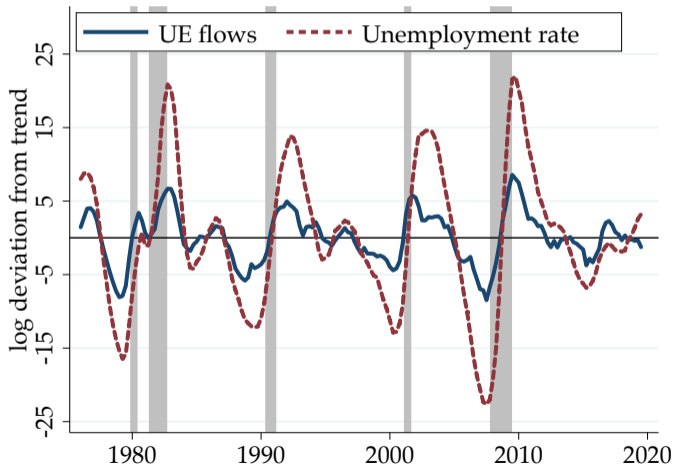


CPS-ASEC (1976-2019): # of weeks workers spent in unemployment in previous year

# 1. Flow origins of shifts in employment distribution

Unemployment-to-employment (UE) flows are strongly **countercyclical**

- see e.g. Burda and Wyplosz (1992), Fujita and Ramey (2009), Elsby et al. (2013)

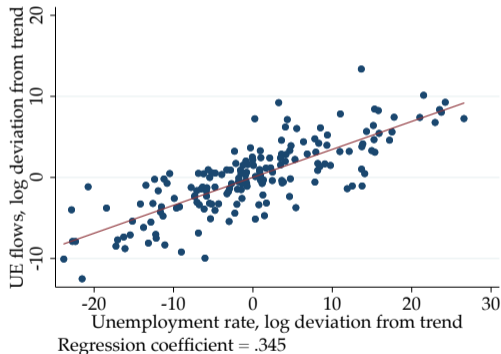
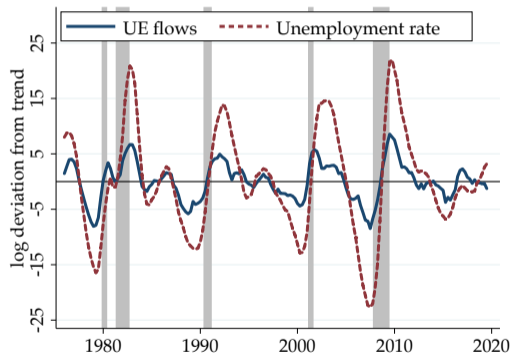


CPS (1976-2019): quarterly average of monthly UE flows, log deviations from trend

# 1. Flow origins of shifts in employment distribution

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## Why are UE flows countercyclical?

Use steady state expressions for unemployment,  $u = \delta / (\delta + f)$ , and UE flows =  $f \cdot u$

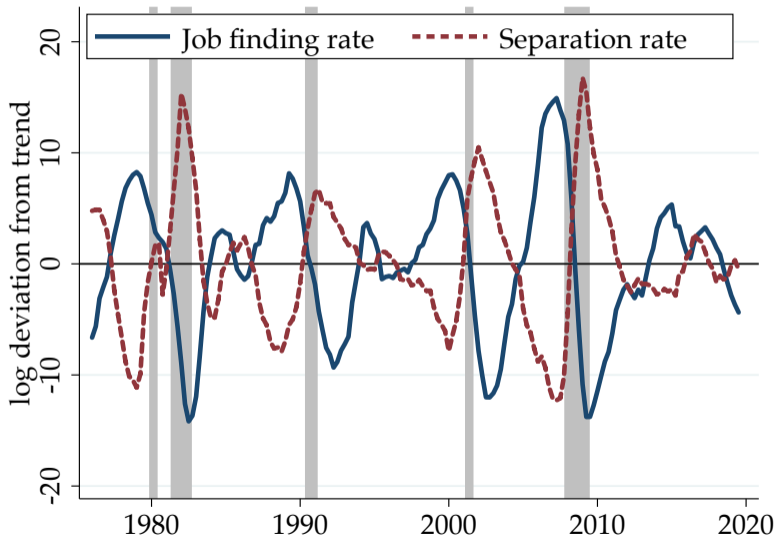
- $\delta$ : separation probability,  $f$ : job finding probability

$$\frac{dUE}{UE} = \frac{df}{f} + \frac{du}{u} \rightarrow \frac{dUE/UE}{du/u} = \frac{1}{(1-u) \left[ -1 + \frac{d\delta/\delta}{df/f} \right]} + 1$$

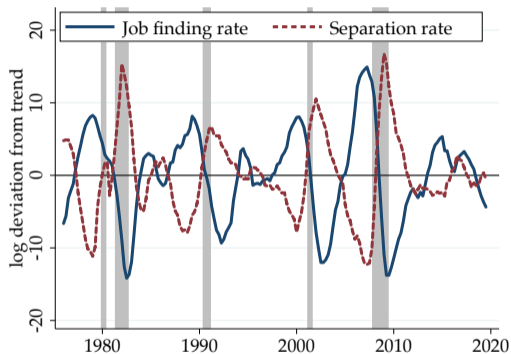
Relative cyclicity of  $\delta$  and  $f$  is key in explaining cyclicity of  $UE$ :

- if  $\frac{d\delta/\delta}{df/f} = 0$ , then  $\frac{dUE/UE}{du/u} = -\frac{u}{1-u} \rightarrow$  UE flows are *procyclical*
- if  $\frac{d\delta/\delta}{df/f} < -\frac{u}{1-u}$ , then  $\frac{dUE/UE}{du/u} > 0 \rightarrow$  UE flows are *countercyclical*

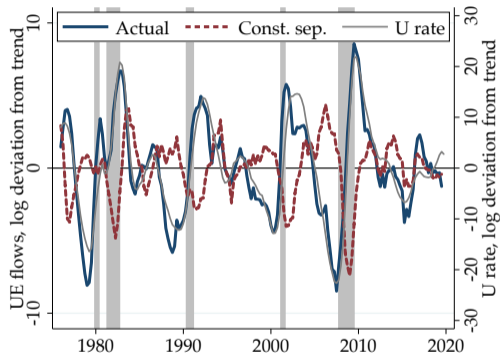
## Cyclicality of job finding and separation probabilities



# Cyclicality of job finding and separation probabilities



more



OECD

## 2. Limited capacity to absorb increases in unemployment by hiring

**Congestion** (in hiring): Firms' limited capacity to absorb "pure" unemployment increases

- i.e. increases in unemployment that leave fundamentals (productivity) unchanged

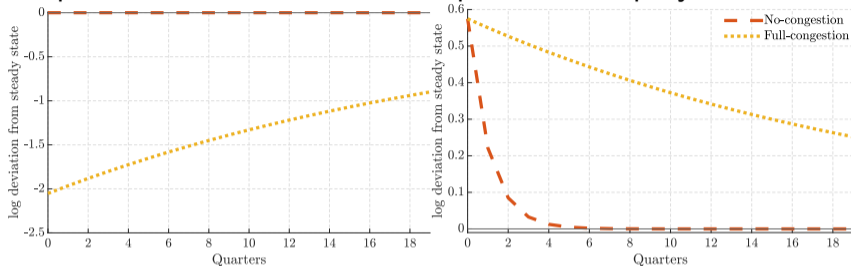
Intuition from standard search model

- hiring depends on fundamentals (e.g. productivity), but market size is irrelevant
- rise in unemployed, with unchanged productivity, (immediately) hired away

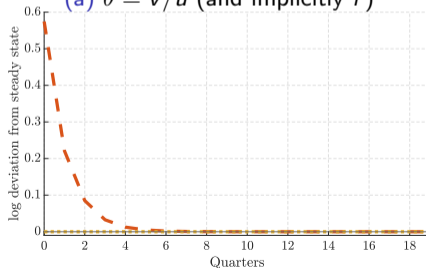
Dynamics after "pure" increase in unemployment (consider one-time exogenous  $\uparrow EU$ ):

- $u' = u - f(\bar{\theta})u + \uparrow EU$ : **no congestion** ( $v$  moves 1-to-1 with  $u$ , i.e.  $\theta = v/u$  fixed)
- $u' = u - \overline{UE} + \uparrow EU$ : **full congestion** (economy absorbs fixed  $\overline{UE}$  at a time)

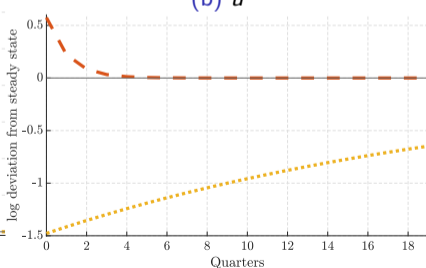
## 2. Response of labor market to “pure” unemployment increase



(a)  $\theta = v/u$  (and implicitly  $f$ )



(b)  $u$



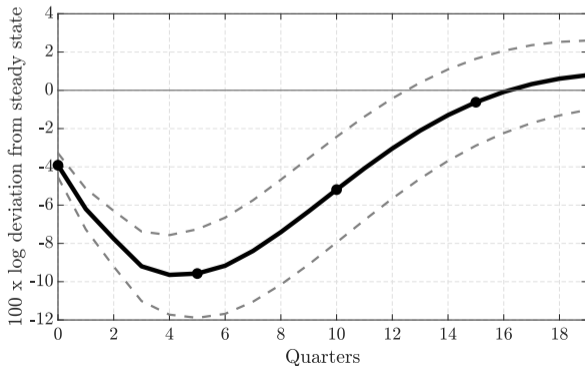
(c)  $UE = f \cdot u$

(d)  $v$

## 2. Response of labor market to “pure” unemployment increase

Estimate VAR:  $\mathbf{y}_t = [\ln ALP_t, \ln \delta_t, \ln \theta_t]$

- $ALP$  = average labor productivity,  $\delta$  = EU probability,  $\theta$  = labor market tightness
- Cholesky identification: response to  $\delta$  keeps  $ALP$  fixed upon impact **ALP?**
- IRF of labor market tightness  $\theta = v/u$  to separation rate shock:



## 2. Model

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*Main idea*

## Main idea

Use standard Diamond-Mortensen-Pissarides (DMP) model, but

- incorporate diminishing returns to cohorts of new hires

Within this framework

- in recessions, UE flows increase (as separations increase) (**empirical fact 1**)
- diminishing marginal product of new hires—our congestion mechanism!
- discourages job creation (**empirical fact 2**)

Modeling choice for the mechanism:

**Imperfect substitution b/w workers** with different labor market experiences

- skill accumulation, career steps, internal labor markets, etc...



## Main idea

Imperfect substitution between workers with different types of labor market experience

$$Y = z \left( \sum_{k=1}^K \alpha_k n_k^\sigma \right)^{1/\sigma}$$

- $z$ : aggregate (total factor) productivity
- $k$ : particular type of labor market experience (e.g. job ladder, skill, ...)
- $n_k$ : # of workers of type  $k$
- $\alpha_k$ : relative productivity
- $\sigma$ : guides diminishing returns (elasticity of substitution:  $\frac{1}{1-\sigma}$ )
  - $\sigma = 1$ : “no congestion” (standard) model
  - $\sigma < 1$ : model featuring congestion in hiring

## Main idea

Relative supply of worker types matters for productivity:

$$p_k = \frac{\partial Y}{\partial n_k} = \alpha_k n_k^{\sigma-1} \frac{Y}{\sum_{l=1}^K \alpha_l n_l^{\sigma}}$$

In recessions, when UE flows rise

- recently unemployed become relatively abundant in employment (**empirical fact 1**)
- depressed marginal product of new hires
- discouraged job creation (**empirical fact 2**)

**Countercyclical congestion** reinterprets recessions: why does unemployment rise?

- Standard question: Why do firms hire so little?
- Our answer:

Firms are actually (gross) hiring more—the jobs to be filled by the unemployed are already crowded.

## 2. Model

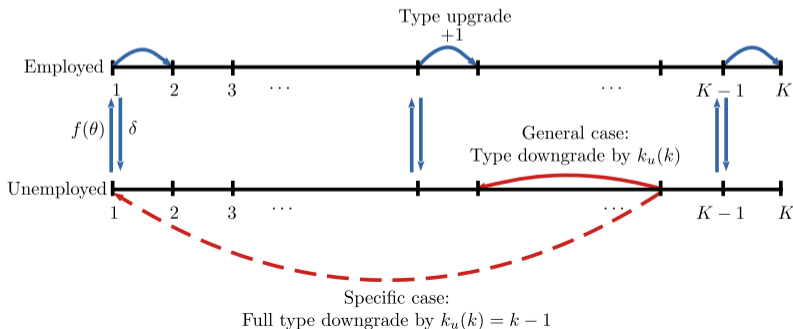
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### *Details*

# Worker heterogeneity and congestion

## Worker “types”

- $k \in \mathcal{K} = \{1, \dots, K\}$ : particular worker types
- employed workers move up one level each period:  $k_{t+1} = k_t + 1$
- unemployed workers move down  $k_u(k)$  levels each period:  $k_{t+1} = k_t - k_u(k)$ 
  - $k_u(k) \in \{0, 1, \dots, k - 1\}$  nests no-, full- and partial-downgrading [more](#)



# Worker heterogeneity and congestion

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## Congestion

- final good produced combining intermediate goods ( $n_k$ ):  $Y = z \left( \sum_{k=1}^K \alpha_k n_k^\sigma \right)^{1/\sigma}$
- intermediate goods produced by “firms” using linear technology
- competitive market prices of intermediate goods:  $p_k = \frac{\partial Y}{\partial n_k} = \alpha_k n_k^{\sigma-1} \frac{Y}{\sum_{l=1}^K \alpha_l n_l^\sigma}$

Everything that follows mirrors “standard” search model

# Environment and timing

## Environment

- workers hired by intermediate-goods firms in frictional labor market
  - random search, matches occur according to  $M(u, v)$ ,  $f(\theta) = M/u$  and  $q(\theta) = M/v$
  - worker-firm matches separate with time-varying probability  $\delta$
- final goods firm buys intermediate-inputs in perfectly competitive market
- wages are determined by period-by-period Nash bargaining, no wage rigidity
- free entry of intermediate-goods firms

## Timing

- aggregate shocks: productivity and separation rate,  $(z, \delta)$ , materialize
- separated workers join unemployment pool, active matches produce
- employed upgrade, unemployed downgrade types and thereafter matching occurs

## Worker and firm value functions

Value of employment and of unemployment for type-k worker

$$W_{k,t} = w_{k,t} + \beta \mathbb{E}_t [(1 - \delta_{t+1})W_{k+1,t+1} + \delta_{t+1}U_{k+1,t+1}],$$

$$U_{k,t} = b + \beta \mathbb{E}_t [f(\theta_t)(1 - \delta_{t+1})W_{k-k_u(k),t+1} + (1 - f(\theta_t))(1 - \delta_{t+1})U_{k-k_u(k),t+1}]$$

$w_{k,t}$ : wage of type-k worker,  $b$ : flow value of unemployment

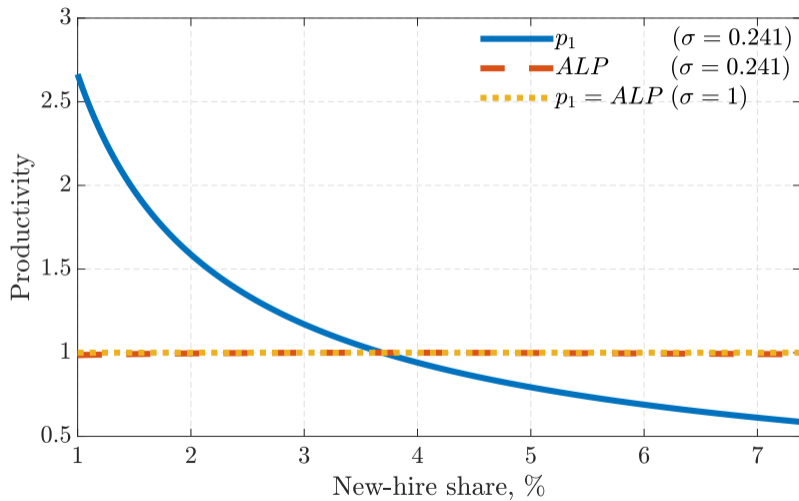
Value of a job filled with type-k worker and of unfilled job

$$J_{k,t} = p_{k,t} - w_{k,t} + \beta \mathbb{E}_t [(1 - \delta_{t+1})J_{k+1,t+1} + \delta_{t+1}V_{t+1}],$$

$$V_t = -\kappa + q(\theta_t)\beta \mathbb{E}_t \left[ (1 - \delta_{t+1}) \frac{u_{k,t}}{u_t} J_{k-k_u(k),t+1} + \delta_{t+1} V_{t+1} \right]$$

$\kappa$ : flow cost of having an open vacancy

# Productivity and size of hiring cohort





## Model mechanism and alternatives

**Congestion** occurs because recently unemployed become abundant in recessions

- fall in marginal product slows further hiring (despite free entry)
- depends on distribution of types in (un-)employment (and  $\sigma$ )

Alternative: what if not all new hires cause congestion?

- Extension in paper: only  $1 - x$  cause congestion:

$$Y = z \left[ (1 - x) \left( \sum_{k=1}^K \alpha_k^c (n_k^c)^\sigma \right)^{1/\sigma} + x \left( \sum_{k=1}^K \alpha_k^{nc} n_k^{nc} \right) \right]$$

→ isomorphic to our baseline, subject to “iso-congestion” reparameterization of  $\sigma$

visualiation

Alternative: what if hiring is slowed by increased costs?

- $\kappa(UE_t)' > 0$ : gives similar amplification, but misses a range of other results
- Related: Hall (2005), Petrosky-Nadeau (2014), Coles and Kelishomi (2018), Engbom (2020)

### 3. Business cycle performance

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*Sources of amplification and congestion unemployment*

## Parameterization

“Standard” features parametrized in “standard” fashion (Shimer, 2005) [more](#)

- in particular,  $b$  s.t. replacement rate of 40 percent (i.e. high fundamental surplus)

Worker heterogeneity

- $K = 160$ : absorbing “max type” until separation
- $k_u(k) = k - 1$ : full downgrading (but recall robustness w.r.t. no-congestion hires)
- $\alpha_k$ : s.t.  $p_k = p = 1$  for all  $k$  (all types have same surplus in steady state)

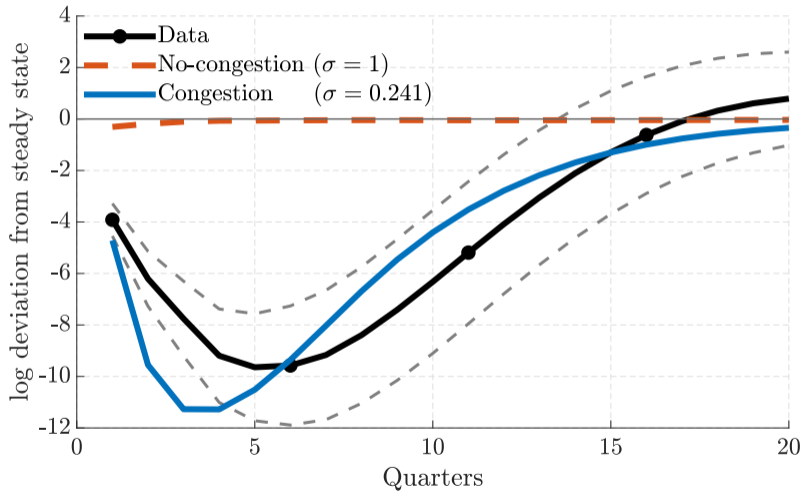
Aggregate shocks

- $z$  and  $\delta$ : target volatility and persistence of  $ALP$  and  $UE/E$  in data
  - $UE/E$  crucial for congestion mechanism
  - Robustness in paper: match  $\delta$  directly

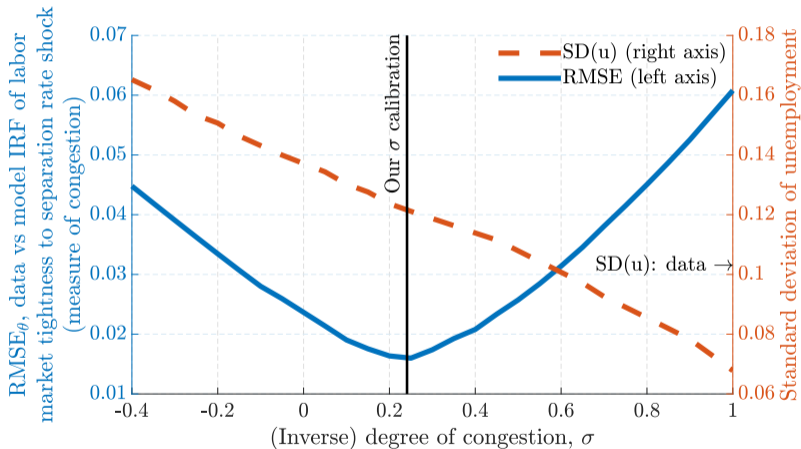
Congestion parameter  $\sigma$

- match limited capacity to absorb unemployed (IRF of  $\theta$  w.r.t.  $\delta$ )
- Robustness/validation: new hires' wage cyclical

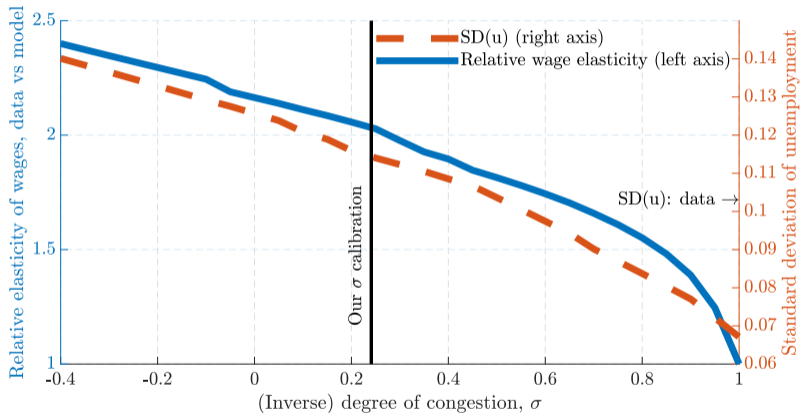
## Parameterizing $\sigma$ : IRF of $\theta$ w.r.t $\delta$



# Parametrizing $\sigma$ : Congestion and amplification



# Parameterizing $\sigma$ : Excess cyclical-ity of new-hire wages



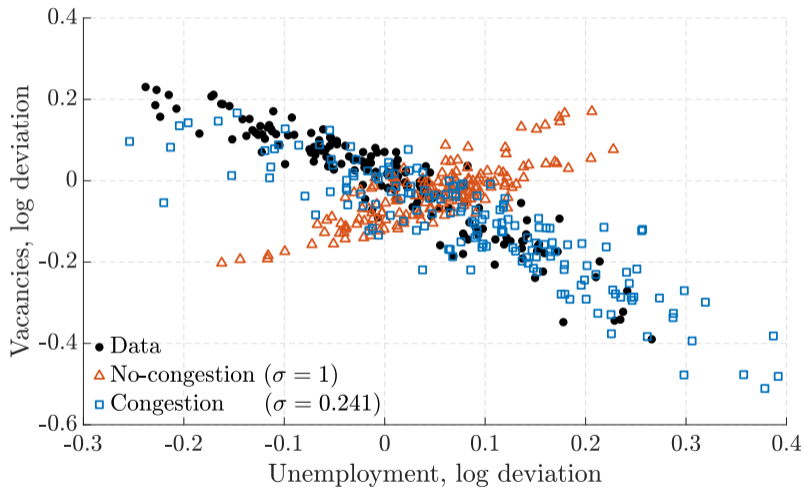
## Model performance: Business cycle statistics

	<i>ALP</i>	<i>f</i>	$\delta$	<i>u</i>	<i>v</i>	$\theta$	<i>UE/E</i>	<i>MPL</i>
	<i>data</i>							
<i>s.d.(x)</i>	0.010	0.053	0.067	0.104	0.127	0.229	0.067	NA
<i>corr(u, x)</i>	-0.11	-0.93	0.85	1	-0.94	-0.98	0.83	NA
	<i>baseline model (<math>\sigma = 0.241</math>)</i>							
<i>s.d.(x)</i>	0.010	0.059	0.122	0.121	0.102	0.207	0.067	0.055
<i>corr(u, x)</i>	-0.46	-0.92	0.74	1	-0.72	-0.94	0.87	-0.86
	<i>standard model (<math>\sigma = 1</math>) without separation shocks</i>							
<i>s.d.(x)</i>	0.010	0.005	0.000	0.004	0.014	0.016	0.003	0.010
<i>corr(u, x)</i>	-0.65	-0.65	0.00	1	-0.49	-0.65	-0.27	-0.65
	<i>standard model (<math>\sigma = 1</math>) with separation shocks</i>							
<i>s.d.(x)</i>	0.010	0.005	0.073	0.055	0.046	0.017	0.054	0.010
<i>corr(u, x)</i>	-0.50	-0.62	0.91	1	0.96	-0.62	0.70	-0.50

full table

more

## Model performance: Beveridge curve





## Sources of amplification

Surplus relevant for hiring in “standard” model ( $S$ ) and in congestion model ( $S_1$ )

$$S_t = z_t - b + \beta \mathbb{E}_t(1 - \delta_{t+1})(1 - f(\theta_t)\phi)S_{t+1}$$

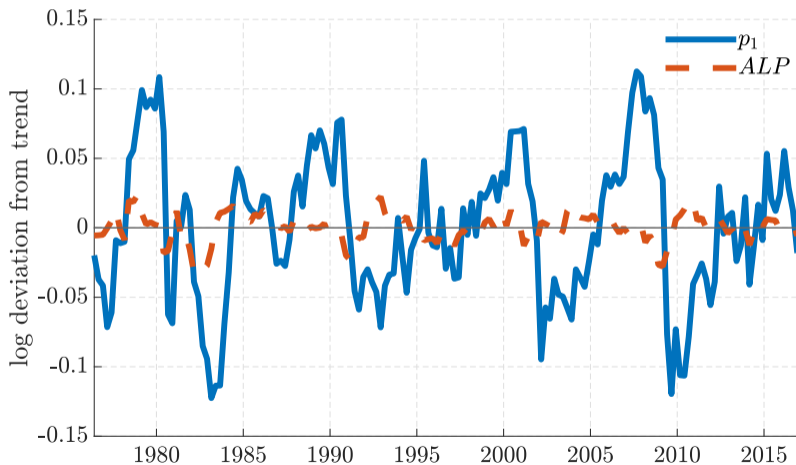
$$S_{1,t} = p_{1,t} - b + \beta \mathbb{E}_t[(1 - \delta_{t+1})(S_{2,t+1} - f(\theta_t)\phi S_{1,t+1})]$$

Differences between standard and our congestion model

- **flow productivity channel:**  $sd(p_1) > sd(z)$
- **cohort channels:** continuation values have different dynamics
  - continuing in employment entails upgrading to  $S_{2,t+1}$
  - falling into unemployment entails downgrading to  $S_{1,t+1}$

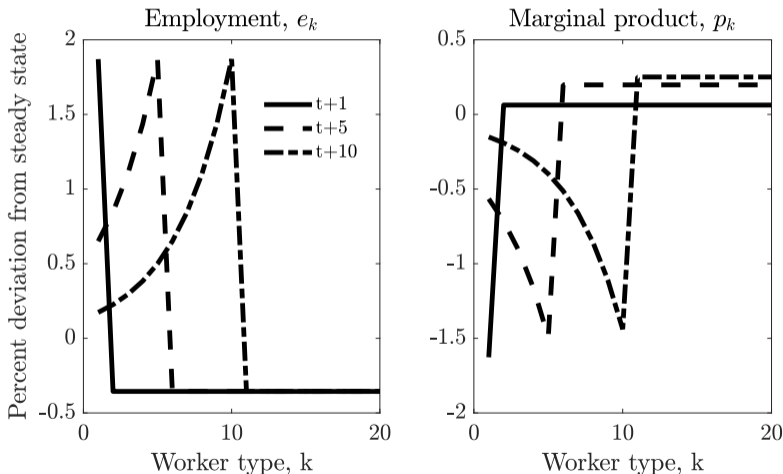
## Sources of amplification: Flow productivity channel

Use data on ALP and UE/E to construct  $p_1$



## Sources of amplification: Cohort channels

IRFs of employment distributions to a one-time positive  $\delta$  shock



## Sources of amplification: Quantification

$$S_{1,t} = p_{1,t} - b + \beta \mathbb{E}_t [(1 - \delta_{t+1})S_{2,t+1} - f(\theta_t)(1 - \delta_{t+1})\phi S_{1,t+1}]$$

$$S_{1,t} = \underbrace{z_t - b + \beta \mathbb{E}_t [(1 - \delta_{t+1})(1 - f(\theta_t^{st})\phi)S_{t+1}^{st}]}_{\text{(i) No-congestion model surplus}} + \underbrace{S_t^* - S_t^{st}}_{\text{(ii) Flow productivity channel}}$$

$$+ \underbrace{\beta \mathbb{E}_t [(1 - \delta_{t+1})(1 - f(\theta_t)\phi)(S_{2,t+1} - S_{t+1}^*)]}_{\text{(iii) Present value channel (cohort effect of "upgrading")}}$$

$$+ \underbrace{\beta \mathbb{E}_t [(1 - \delta_{t+1})f(\theta_t)\phi(S_{2,t+1} - S_{1,t+1})]}_{\text{(iv) Outside option channel (cohort effect of "downgrading")}}$$

$$S_t^{st} = z_t - b + \beta \mathbb{E}_t [(1 - \delta_{t+1})(1 - f(\theta_t^{st})\phi)S_{t+1}^{st}] \quad \& \quad \theta_t^{st}: \text{no-congestion surplus} \quad \& \quad \theta$$

$$S_t^* = p_{1,t} - b + \beta \mathbb{E}_t [(1 - \delta_{t+1})(1 - f(\theta_t)\phi)S_{t+1}^*]: \text{match surplus with } p_{1,t}$$

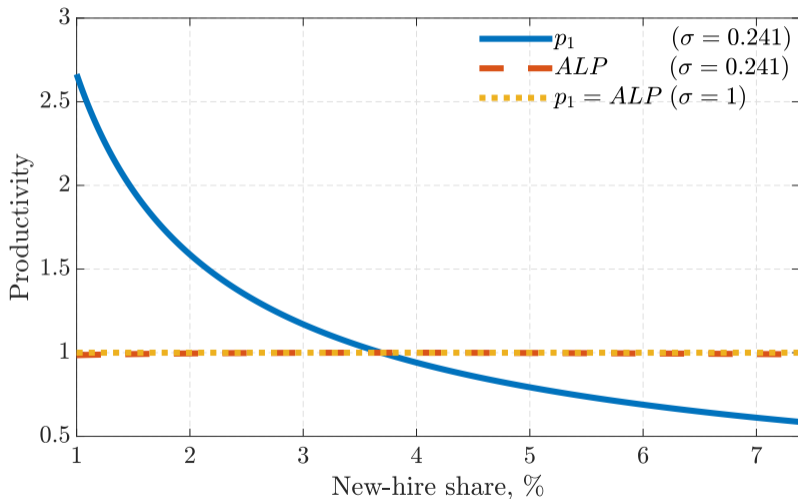
## Sources of amplification: Quantification

Variation in counterfactual labor market tightness ( $\tilde{\theta}$ ) driven by

- (i) no-congestion model surplus
- (ii) flow productivity channel
- (iii) present value channel
- (iv) outside option channel

		Standard deviation	Contribution to total
No-congestion model	(i)	0.019	0.049
+ Flow productivity channel	(i)+(ii)	0.052	0.162
+ Present value channel	(i)+(ii)+(iii)	0.178	0.851
+ Outside option channel	(i)+(ii)+(iii)+(iv)	0.207	1

## Congestion unemployment: Historical decomposition



## Congestion unemployment: Historical decomposition

Unemployment fluctuations driven solely by congestion: **congestion unemployment**

$$S_{k,t}^c = p_{k,t} \cdot \frac{\bar{z}}{z_t} - b + \beta \mathbb{E}_t(1 - \bar{\delta}) S_{k+1,t+1}^c - \beta \mathbb{E}_t(1 - \bar{\delta}) f(\theta_t^c) \phi S_{1,t+1}^c \quad \forall k.$$

- $S_k^c$ : surplus variation only due to congestion

$$\kappa = q(\theta_t^c) \beta \mathbb{E}_t(1 - \bar{\delta}) S_{1,t}^c$$

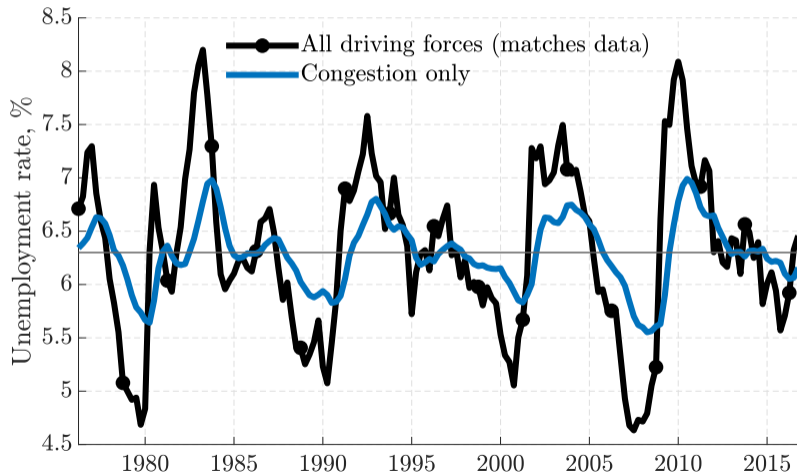
- $\theta^c$ : variation in labor market tightness only due to congestion

$$u_{t+1}^c = (1 - f(\theta_t^c)) u_t^c + \bar{\delta}(1 - u_t^c)$$

- $u^c$ : congestion unemployment

# Congestion unemployment: Historical decomposition

Use data on ALP and UE/E to estimate time-path of entire model (Kalman filter)





## 4. Additional applications

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*Congestion and three macroeconomic regularities*

# Macroeconomic implications of congestion

1. Business cycle accounting: the labor wedge
  - countercyclical in the data and attributed to “household side”
2. Costs of entering labor market and of displacement
  - large and countercyclical in the data
3. Sensitivity to labor market policies
  - relatively low, hard to square with high labor market volatility

# 1. The labor wedge

The **labor wedge** is defined as  $MPL_t(1 - \tau_t) = MRS_t$

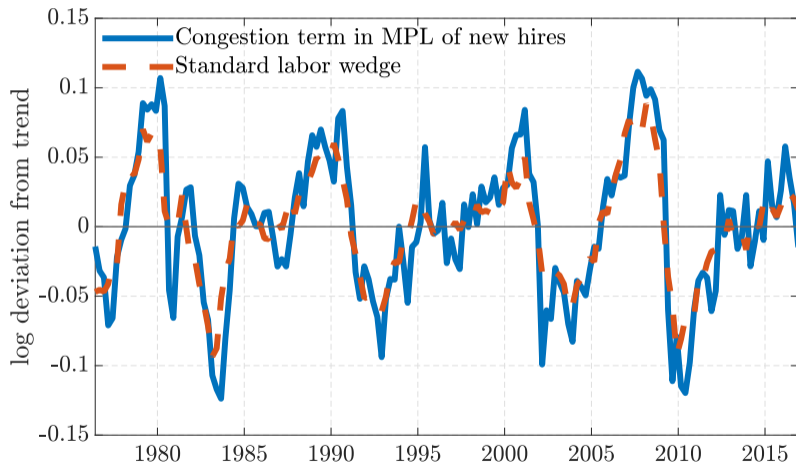
- o estimates in data show a *cyclical* labor wedge
  - o see e.g. Hall (1997), Chari, Kehoe, McGrattan (2007), Shimer (2009)
- o moreover, fluctuations in labor wedge assigned mainly to “household” (MRS) side
  - o should focus on how MRS deviates from real wage (e.g. Karabarbounis, 2014)

Extend our baseline model to include capital ( $\tilde{K}$ ):  $Y = z\tilde{K}^a \left( \left[ \sum_{k=1}^K \alpha_k n_k^\sigma \right]^{\frac{1}{\sigma}} \right)^{1-a}$

- o considering the (spot) productivity of new hires  $p_1$  only

$$\underbrace{p_1}_{\text{allocative/new hires' MPL}} = \underbrace{(1-a)\frac{Y}{N}}_{\text{standard MPL}} \underbrace{\frac{\alpha_1 s_1^{\sigma-1}}{\sum_{k=1}^K \alpha_k s_k^\sigma}}_{\text{labor wedge (congestion term)}} = MRS$$

# 1. The labor wedge: Congestion as a resolution?



## 2. Countercyclical costs of job displacement and labor market entry

### Large and persistent

- ... earnings losses from graduating in recessions (e.g. Kahn, 2010, Schwandt and von Wachter, 2019)
  - graduation costs: skill mismatch, employer quality (e.g. Oreopoulos et al., 2012)
- ... and countercyclical displacement costs (e.g. Davis and von Wachter, 2011)
  - driven by wage drops, employer “quality” (e.g. Schmieder et al., 2019)

### *Level* of costs explained through various theories

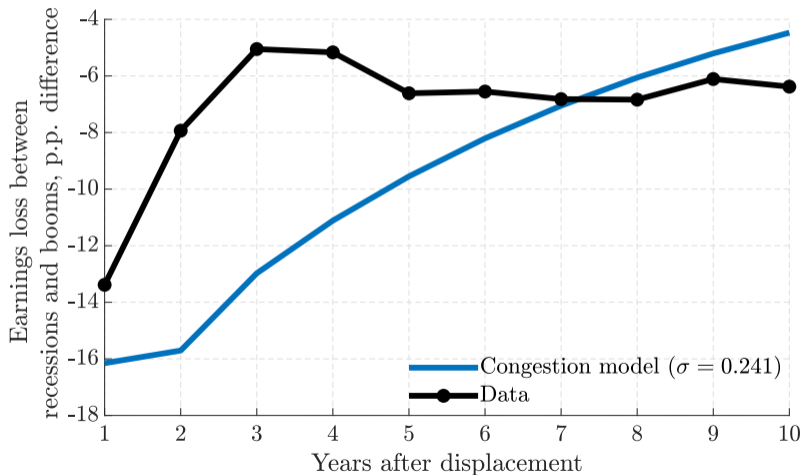
- displacement costs: fall off a job ladder (e.g. Jarosch, 2015, Jung and Kuhn, 2018)

### Our model speaks to the cyclical nature of these costs

- not well understood in existing literature

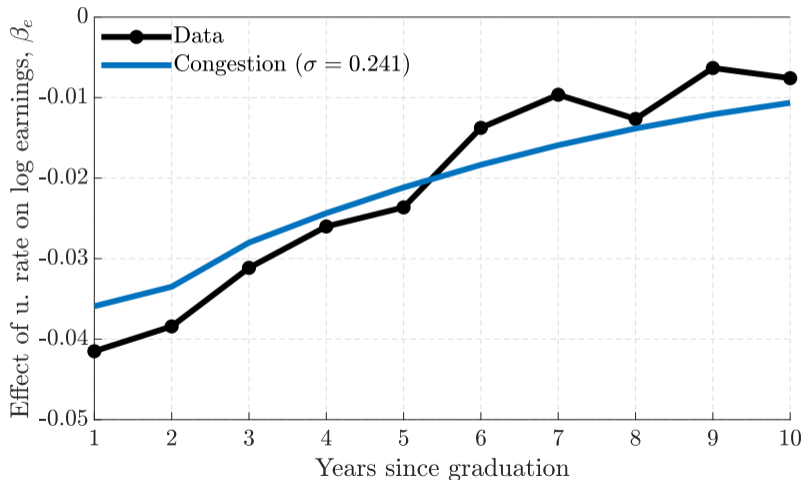
## 2. Costs of displacement

Davis and von Wachter (2011): earnings losses in recessions relative to booms



## 2. Costs of labor market entry

Schwandt and von Wachter (2019): earnings losses of new hires and unemployment



## 2. Costs of displacement and labor market entry

Our model offers a basic explanation

- relatively large cohorts of new hires are abundant in employment
- pushes down their wages (reflecting marginal products)
- cohort effects make these initial conditions long-lasting

The above mechanism is broadly consistent with the available evidence

- earnings losses linked to persistent wage declines
- driven primarily by a shift towards jobs of “lower quality”
  - Schmieder et al (2019) and Schwandt and von Wachter (2019)



### 3. Sensitivity to labor market policies

Costain and Reiter (2008): search models have a hard time

- simultaneously matching labor market volatility
- ... and sensitivity of the labor market to changes in policies
- estimate long-run elasticity of  $u$  w.r.t.  $b$  of  $\epsilon_{u,b} \in (2, 3.5)$

In our model, labor market volatility is not generated by low fundamental surplus

- instead, countercyclical congestion makes labor market variables volatile
- implied long-run elasticity of  $u$  w.r.t.  $b$  of  $\epsilon_{u,b} \approx 2.6$

## Conclusion

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# Conclusion: A congestion theory of unemployment fluctuations

Two key empirical facts

1. employment shifts towards recently unemployed in downturns
2. unemployment increases not absorbed quickly, even with unchanged fundamentals

We propose a model consistent with the above facts

- worker types are imperfect substitutes
- abundant types see their marginal productivity fall, discouraging their hiring
- congestion is a strong amplification mechanism

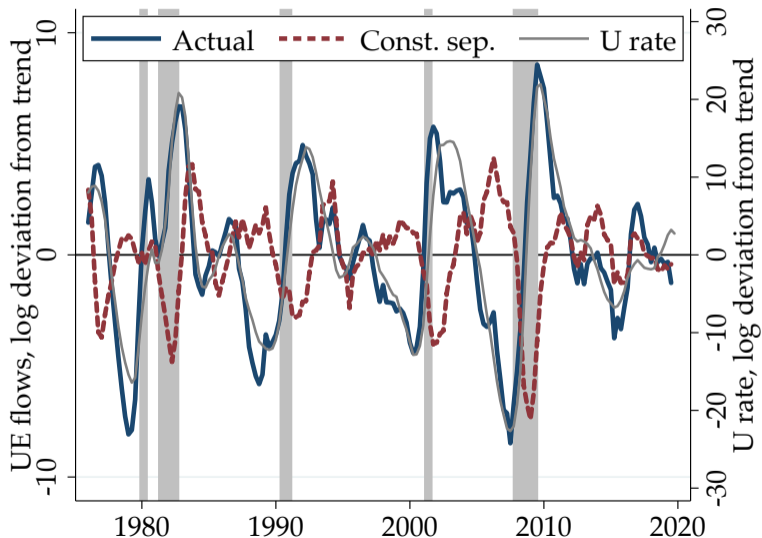
Our baseline model sheds new light on a range of macroeconomic patterns

- labor market variables over the business cycle
- relative wage cyclicality of new hires
- countercyclical labor wedge
- countercyclical costs of displacement and labor market entry
- low sensitivity of labor market variables to labor market policies

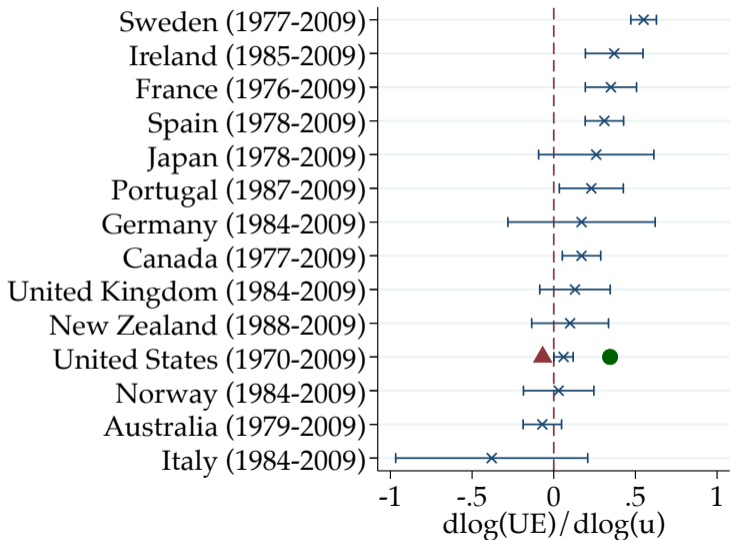
Thanks

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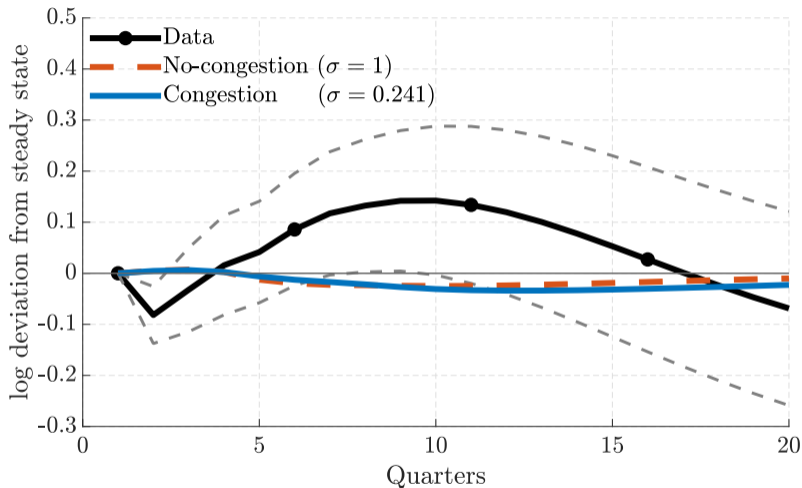
# UE flows: observed and counterfactual (constant separations)



## UE Flows in the Data: OECD

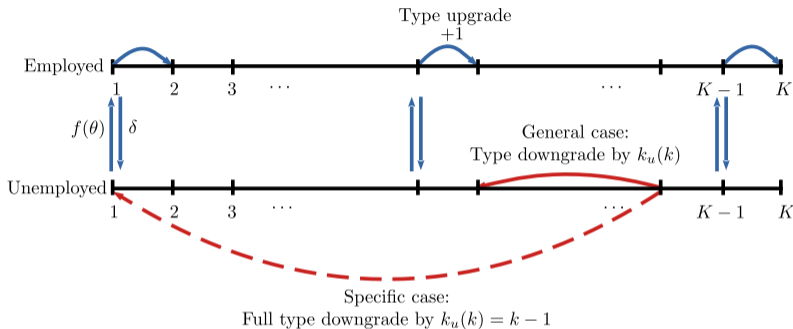


# Response of labor markets to “pure” unemployment increase



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# Worker type and (un-)employment evolution



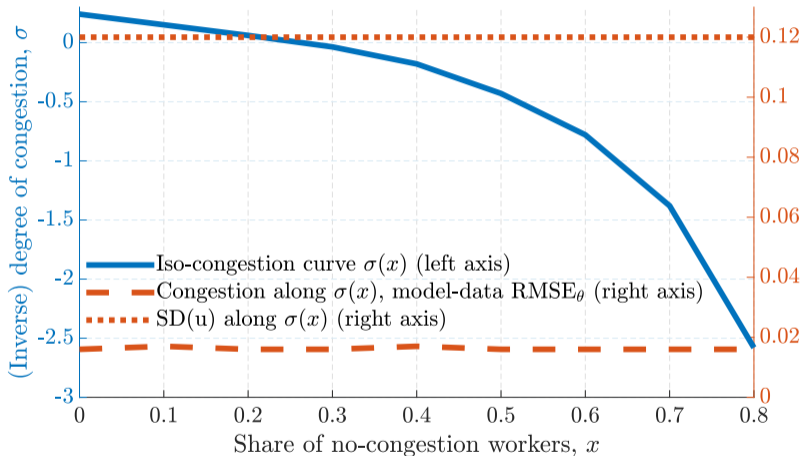
## Laws of motion for (un-)employment

$$u_{k-k_u(k),t} = (1 - f(\theta_{t-1}))u_{k,t-1} + \delta_t e_{k-k_u(k),t} \text{ for } k \in \mathcal{K}$$

$$e_{k-k_u(k),t} = (1 - \delta_{t-1})e_{k-k_u(k)-1,t-1} + f(\theta_{t-1})u_{k,t-1} \text{ for } k \in \mathcal{K}$$



# Iso-congestion model



# Full set of parameter values

Parameter	Value		Target/Source	Data	Model	
	<i>Congestion</i>	<i>No congestion</i>				
$\beta$	Discount factor	0.99	Annual interest rate	0.04	0.04	
$\mu$	Matching elasticity	0.72	Shimer (2005)	0.72	0.72	
$\bar{m}$	Matching efficiency	0.57	Job finding probability	0.57	0.57	
$\eta$	Bargaining power	0.72	Hosios condition	0.72	0.72	
$b$	Unemployment flow value	0.39	Avg. replacement rate	0.40	0.40	
$\kappa$	Vacancy posting cost	0.21	Normalization $\theta = 1$	—	1.00	
$\bar{z}$	Productivity shock, mean	1	Normalization	—	1.00	
$\sigma_z$	Productivity shock, st. dev.	0.008	St. dev. of ALP	0.010	0.010	
$\rho_z$	Productivity shock, persistence	0.956	Persistence of ALP	0.74	0.69	
$\bar{\delta}$	Separation shock, mean	0.037	Unemployment rate	0.063	0.063	
$\sigma_\delta$	Separation shock, st. dev.	0.107	0.080	St. dev. of UE/E	0.067	0.067
$\rho_\delta$	Separation shock, persistence	0.709	0.670	Persistence of UE/E	0.84	0.74
$\rho_{\delta,z}$	Correlation of shocks to $z$ and $\delta$	-0.505	-0.560	$\text{corr}(ALP, \delta)$	-0.41	-0.41
$\sigma$	Elasticity of substitution b/w workers	0.241	1	Impulse response of $\theta$ to $\delta$ , see IRF Figure		
$\alpha_k$	Relative productivities of worker types	see Appendix		$p_k = 1$ for all $k$		

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## Model Performance: Full business cycle statistics

	$ALP$	$f$	$\delta$	$u$	$v$	$\theta$	$UE/E$	$p_1$
<i>Panel A: Data</i>								
Standard deviation	0.010	0.053	0.067	0.104	0.127	0.229	0.067	NA
Autocorrelation	0.75	0.87	0.77	0.93	0.93	0.94	0.84	NA
Correlation matrix								
$f$	0.04	1						
$\delta$	-0.41	-0.71	1					
$u$	-0.11	-0.93	0.85	1				
$v$	0.30	0.87	-0.87	-0.94	1			
$\theta$	0.22	0.92	-0.87	-0.98	0.99	1		
$UE/E$	-0.17	-0.72	0.57	0.83	-0.72	-0.78	1	
<i>Panel B: Congestion Model</i>								
Standard deviation	0.010	0.059	0.122	0.121	0.102	0.207	0.067	0.055
Autocorrelation	0.69	0.90	0.53	0.84	0.86	0.90	0.74	0.75
Correlation matrix								
$f$	0.44	1						
$\delta$	-0.41	-0.51	1					
$u$	-0.46	-0.92	0.74	1				
$v$	0.35	0.92	-0.16	-0.72	1			
$\theta$	0.34	1.00	-0.51	-0.94	0.91	1		
$UE/E$	-0.44	-0.93	0.39	0.87	-0.88	-0.94	1	
$p_1$	0.49	0.95	-0.43	-0.86	0.90	0.95	-0.97	1

# Model Performance: Full business cycle statistics

	$ALP$	$f$	$\delta^{imp}$	$u$	$v$	$\theta$	$UE/E$	$p_1$
<i>Panel A: Congestion Model - Calibrating to <math>UE/E</math></i>								
Standard deviation	0.010	0.059	0.122	0.121	0.102	0.207	0.067	0.055
Autocorrelation	0.69	0.90	0.53	0.84	0.86	0.90	0.74	0.75
Correlation matrix								
$f$	0.44	1						
$\delta$	-0.41	-0.51	1					
$u$	-0.46	-0.92	0.74	1				
$v$	0.35	0.92	-0.16	-0.72	1			
$\theta$	0.34	1.00	-0.51	-0.94	0.91	1		
$UE/E$	-0.44	-0.93	0.39	0.87	-0.88	-0.94	1	
$p_1$	0.49	0.95	-0.43	-0.86	0.90	0.95	-0.97	1
<i>Panel B: Congestion Model - Calibrating to <math>\delta</math></i>								
Standard deviation	0.010	0.041	0.084	0.086	0.077	0.144	0.052	0.054
Autocorrelation	0.69	0.92	0.62	0.87	0.82	0.92	0.76	0.75
Correlation matrix								
$f$	0.36	1						
$\delta$	-0.41	-0.43	1					
$u$	-0.42	-0.88	0.77	1				
$v$	0.20	0.88	0.04	-0.56	1			
$\theta$	0.36	1.00	-0.43	-0.90	0.87	1		
$UE/E$	-0.31	-0.91	0.52	0.89	-0.72	-0.91	1	
$p_1$	0.47	0.92	-0.55	-0.88	0.73	0.91	-0.97	1