The State Dependent Effectiveness of Hiring Subsidies

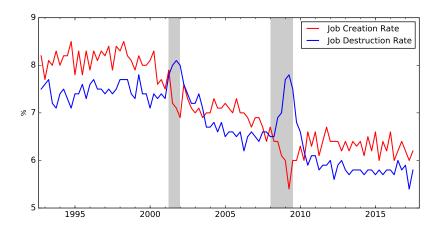
Sebastian Graves
New York University

Job Creation and Destruction Rates

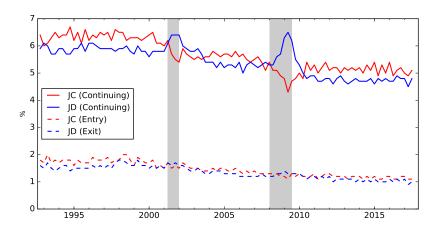
 $\label{eq:matter} \textbf{Employment Growth} = \textbf{Job Creation Rate} - \textbf{Job Destruction Rate}$

- ▶ Job Creation: Increase in employment from continuing and entering establishments
- ▶ Job Destruction: Decrease in employment from continuing and exiting establishments

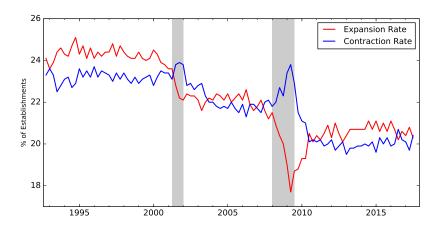
Job Creation and Destruction Rates



Job Creation and Destruction: Continuing vs Entry/Exit



Employment Adjustment is Lumpy



Motivation

- What impact does lumpy employment adjustment have on the responsiveness of job creation and destruction to aggregate shocks?
- ▶ Does lumpy adjustment have implications for the effectiveness of different labor market policies over the business cycle?

Outline

- Significant evidence of time-varying responsiveness of job creation and destruction rates:
 - Job creation rate significantly more responsive when employment growth is high.
 - Job destruction rate significantly less responsive when employment growth is high.
- 2. Heterogeneous-firm model with lumpy employment adjustment can replicate these facts.
- 3. Policy implications: hiring subsidy is less effective in downturns, while firing tax is more effective.

Empirical Evidence: Bartik Approach

Construct predicted employment growth following Bartik (1991):

$$B_{i,t} = \sum_{k=1}^{K} \varphi_{i,k,\tau} g_{-i,k,t}^{N}$$

- $\varphi_{i,k,\tau}$ = employment share of industry k in state i in base year $\tau = 1990$
- $g_{-i,k,t}$ = national employment growth in industry k excluding state i in year t

Empirical Evidence: Bartik Approach

- Does response to predicted employment growth vary over time?
- Panel Regressions:

$$\Delta Y_{i,t} = \alpha_i + \gamma_t + \beta_0 B_{i,t} + \beta_1 B_{i,t} \cdot g_{i,t-1}^N + \Gamma' Z_{i,t-1} + \epsilon_{i,t}$$

- Annual state-level JC/JD data: 1977-2014
- Annual state-industry 3-digit employment data: 1990-2016

Empirical Evidence: Bartik Approach

	$\Delta g_{i,t}^N$	$\Delta JC_{i,t}$	$\Delta JD_{i,t}$
B _{i,t}	0.99***	0.40***	-0.59***
,	(0.20)	(0.13)	(0.14)
$B_{i,t} \cdot g_{i,t-1}^N$	-0.003	0.025**	0.027***
	(0.017)	(0.010)	(0.010)
$\hat{eta}_0 + \hat{eta}_1 ar{ar{g}}^N$	0.99	0.45	-0.532
$\hat{eta}_0 + \hat{eta}_1(ar{g}^N + 1SD)$	0.98	0.54	-0.44
$\log(\frac{\sigma_{95}}{\sigma_5})$	-0.03	0.60	-0.55
Observations	1173	1173	1173
R^2	0.415	0.305	0.369

Notes: Standard errors clustered by state. Asterisks denote significance levels (***= 1%,** = 5%,* = 10%). The mean and standard deviation of state-level employment growth are 2.1% and 3.3%. The 5th and 95th percentiles of the state-level employment growth distribution are -3.4% and 7.1%.

Empirical Evidence: Volatility Approach

- ► Do job creation, job destruction, or employment growth exhibit time-varying volatility?
- ► Panel Regressions:

$$|\Delta Y_{i,t}| = \alpha_i + \gamma_t + \beta_0 |\Delta Y_{i,t-1}| + \beta_1 g_{i,t-1}^N + \epsilon_{i,t}$$

Annual state-level JC/JD data: 1977-2014

Empirical Evidence: Volatility Approach

	$ \Delta g_{i,t}^N $	$ \Delta JC_{i,t} $	$ \Delta JD_{i,t} $
$g_{i,t-1}^N$	0.015 (0.028)	0.080*** (0.013)	-0.101*** (0.024)
$ \widehat{\Delta Y_{i,t}}(\bar{g}^N) $	2.46	1.41	1.70
$ \widehat{\Delta Y_{i,t}}(ar{g}^N+1SD) $	2.50	1.68	1.37
$\log(\frac{\sigma_{95}}{\sigma_5})$	0.06	0.63	-0.64
Observations	1836	1836	1836
R ²	0.17	0.06	0.16

Notes: Standard errors clustered by state. Asterisks denote significance levels (***= 1%,**= 5%,* = 10%). The 5th and 95th percentiles of the annual employment growth distribution at the state level are -3.4% and 7.1%.

Model: Firms

 Continuum of firms, mass normalized to 1, operating production function:

$$y = Az_rz_in^{\alpha}$$

▶ Idiosyncratic (z_i) , regional (z_r) , and aggregate (A) productivity follow independent AR(1) processes:

$$\begin{split} \log A' &= \rho_A \log A + \sigma_A \epsilon_A', & \ \epsilon_A' \sim N(0,1) \\ \log z_r' &= \rho_r \log z_r + \sigma_r \epsilon_r', & \ \epsilon_r' \sim N(0,1) \\ \log z_i' &= \rho_i \log z_i + \sigma_i \epsilon_i', & \ \epsilon_i' \sim N(0,1) \end{split}$$

▶ Each period, firms choose employment level for the following period, subject to a linear hiring cost, κ .

Model: Representative Household

Representative household with utility function:

$$U(\mathsf{C},\mathsf{N}) = rac{1}{1-\gamma} \left(\mathsf{C} - \psi rac{\mathsf{N}^{1+\phi}}{1+\phi}
ight)^{1-\gamma}$$

► SDF:

$$\Lambda(\mathbf{S}, \mathbf{S'}) = \beta \left(\frac{C(\mathbf{S'}) - \psi \frac{N(\mathbf{S'})^{1+\phi}}{1+\phi}}{C(\mathbf{S}) - \psi \frac{N(\mathbf{S})^{1+\phi}}{1+\phi}} \right)^{-\gamma}$$

► FOCs of intra-temporal problem provide aggregate wage:

$$w(\mathbf{S}) = -\frac{U_N(C, N)}{U_C(C, N)} = \psi N(\mathbf{S})^{\phi}$$

Model: Firm's Recursive Problem

$$\begin{split} V(z_r, z_i, n; \textbf{S}) &= \max_{n'} A z_r z_i n^{\alpha} - w(\textbf{S}) n - \kappa \Delta n' \mathbb{1}\{n' > n\} \\ &+ \mathbb{E}\left[\Lambda(\textbf{S}, \textbf{S}') V(z_r', z_i', n'; \textbf{S}')\right] \end{split}$$

Aggregate state: $\mathbf{S} = (\mathsf{A}, \mu)$

Model: Firm's Recursive Problem

$$V(z_r, z_i, n; \mathbf{S}) = \max_{n'} A z_r z_i n^{\alpha} - w(\mathbf{S}) n - \kappa \Delta n' \mathbb{1}\{n' > n\}$$

+ $\mathbb{E}\left[\Lambda(\mathbf{S}, \mathbf{S}') V(z_r', z_i', n'; \mathbf{S}')\right]$

Aggregate state: $S = (A, \mu)$

Hiring \rightarrow region of inactivity. The FOCs conditional on hiring/firing:

$$\mathbb{E}[\Lambda(\mathbf{S}, \mathbf{S}') V_n(z'_r, z'_i, n'; \mathbf{S}')] = \kappa \text{ if } n' > n$$

$$\mathbb{E}[\Lambda(\mathbf{S}, \mathbf{S}') V_n(z'_r, z'_i, n'; \mathbf{S}')] = 0 \text{ if } n' < n$$

Model: Firm's Recursive Problem

$$V(z_r, z_i, n; \mathbf{S}) = \max_{n'} Az_r z_i n^{\alpha} - w(\mathbf{S}) n - \kappa \Delta n' \mathbb{1}\{n' > n\}$$

+ $\mathbb{E}\left[\Lambda(\mathbf{S}, \mathbf{S}')V(z_r', z_i', n'; \mathbf{S}')\right]$

Aggregate state: $S = (A, \mu)$

Hiring \rightarrow region of inactivity. The FOCs conditional on hiring/firing:

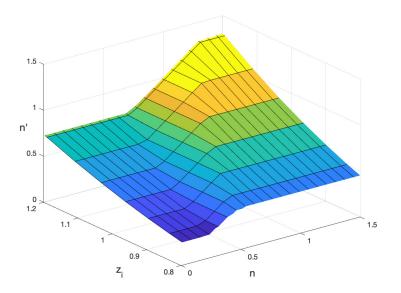
$$\mathbb{E}[\Lambda(\mathbf{S}, \mathbf{S}') V_n(z'_r, z'_i, n'; \mathbf{S}')] = \kappa \text{ if } n' > n$$

$$\mathbb{E}[\Lambda(\mathbf{S}, \mathbf{S}') V_n(z'_r, z'_i, n'; \mathbf{S}')] = 0 \text{ if } n' < n$$

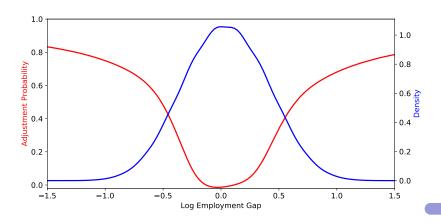
Inaction if:

$$0 \leq \underbrace{\mathbb{E}[\Lambda(\boldsymbol{S}, \boldsymbol{S'})V_n(\boldsymbol{z'_r}, \boldsymbol{z'_i}, n; \boldsymbol{S'})]}_{MB} \leq \kappa$$

Policy Function: Lumpy Employment Adjustment



Employment Gaps and Adjustment Probabilities



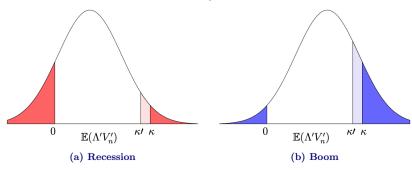
Calibration and Computation

Parameter		Baseline	Frictionless
Hiring cost	κ	0.47	0
Regional shock volatility	σ_{r}	0.0045	0.0025
Idiosyncratic shock volatility	σ_{i}	0.10	0.079
Regional productivity persistence	ρ_{r}	0.97	0.97
Idiosyncratic productivity persistence	ρ_{i}	0.97	0.97
Aggregate shock volatility	σ_{A}	0.0049	0.0039
Aggregate productivity persistence	ρ_{A}	0.974	0.984
Decreasing returns to scale	α	0.65	0.65
Discount factor	β	0.99	0.99
Risk Aversion	γ	1	1
Elasticity of labor supply	$\frac{1}{\phi}$	2	2
Disutility of labor supply	$\overset{\sigma}{\psi}$	0.78	0.73

Computational Detail

Lumpy Adjustment and Time-Varying Responsiveness

The distribution of $\mathbb{E}[\Lambda(S, S')V_n(z'_r, z'_i, n; S')]$ varies over time:



Model Validation

I use simulated data from both versions of the model to repeat the experiments conducted using state-level data:

- Baseline model: time-varying responsiveness quantitatively close to that estimated in the data
- Frictionless model: no time-varying responsiveness

Model Validation: Bartik Regressions

	Baseline			Frictionless		
	$\Delta g_{i,t}^N$	$\Delta JC_{i,t}$	$\Delta JD_{i,t}$	$\Delta g_{i,t}^N$	$\Delta JC_{i,t}$	$\Delta JD_{i,t}$
$B_{i,t}$	1.05	0.59	-0.46	1.00	0.54	-0.45
	(0.97,1.14)	(0.54,0.63)	(-0.50,-0.43)	(0.93,1.07)	(0.50,0.58)	(-0.49,-0.42)
$B_{i,t} \cdot g_{i,t-1}^N$	0.006	0.028	0.022	0.000	0.002	0.001
	(-0.02,0.03)	(0.01,0.04)	(0.01,0.04)	(-0.02,0.02)	(-0.01,0.01)	(-0.01,0.01)
$\begin{array}{c} \hat{\beta}_0 + \hat{\beta}_1 \overline{g}^N \\ \hat{\beta}_0 + \hat{\beta}_1 (\overline{g}^N + 1 \text{SD}) \\ \log(\frac{\sigma_{95}}{\sigma_5}) \end{array}$	1.05	0.59	-0.46	1.00	0.54	-0.45
	1.07	0.68	-0.39	1.00	0.54	-0.45
	0.06	0.50	-0.50	0.01	0.03	-0.03

Notes: 90% confidence intervals in parenthesis, calculated using 500 simulations of the model for 1200 periods. In the model mean employment growth is zero, use the standard deviation of state employment growth of 3.3% estimated in the data. For the 95th and 5th percentiles of the employment growth distribution luse -5.2% and 5.2% (re-centering the values of -3.4% and 7.1% from the data).

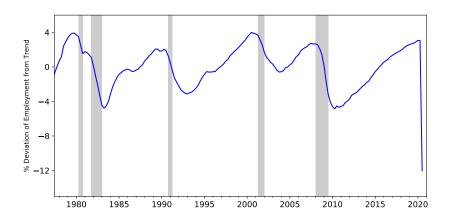
Model Validation: Volatility Regressions

	Baseline			Frictionless		
	$ \Delta g_{i,t}^N $	$ \Delta JC_{i,t} $	$ \Delta JD_{i,t} $	$ \Delta g_{i,t}^N $	$ \Delta JC_{i,t} $	$ \Delta JD_{i,t} $
$g_{i,t-1}^N$	0.02	0.069	-0.049	0.006	0.012	-0.005
	(-0.01,0.05)	(0.05,0.09)	(-0.06,-0.03)	(-0.04,0.05)	(-0.01,0.04)	(-0.02,0.01)
$\frac{ \widehat{\Delta Y_{i,t}} (\bar{g}^N)}{ \widehat{\Delta Y_{i,t}} (\bar{g}^N+1SD)} \\ \log(\frac{\sigma_{95}}{\sigma_5})$	1.99	1.11	0.88	3.78	2.05	1.72
	2.06	1.35	0.72	3.80	2.10	1.71
	0.10	0.67	-0.59	0.02	0.06	-0.03

Notes: 90% confidence intervals in parenthesis, calculated using 500 simulations of the model for 1200 periods. In the model mean employment growth is zero. I use the standard deviation of state employment growth of 3.3% estimated in the data. For the 95th and 5th percentiles of the employment growth distribution I use -5.2% and 5.2% (re-centering the values of -3.4% and 7.1% from the data).

Aggregate Implications

To understand the aggregate implications, I find sequence $\{A_t\}$ such that log deviation of employment from SS value is equal to detrended log US employment:



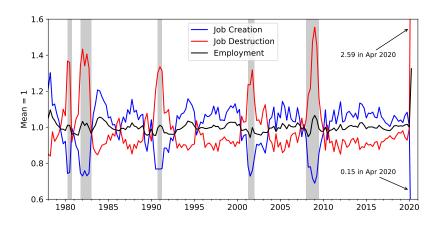


Responsiveness Indices

At each point in time, the "responsiveness" of job creation, job destruction and employment is:

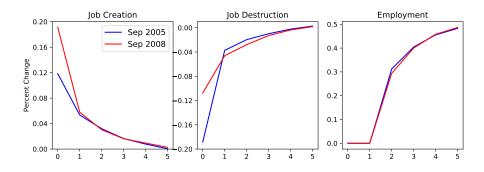
$$\begin{split} R_t^{JC} &\equiv JC(exp(log(A_t) + \sigma_A), \mu_t) - JC(A_t, \mu_t) \\ R_t^{JD} &\equiv JD(exp(log(A_t) + \sigma_A), \mu_t) - JD(A_t, \mu_t) \\ R_t^N &\equiv N(exp(log(A_t) + \sigma_A), \mu_t) - N(A_t, \mu_t) \end{split}$$

Responsiveness Indices: Baseline Model



► Robustness

State Dependent Impulse Response Functions



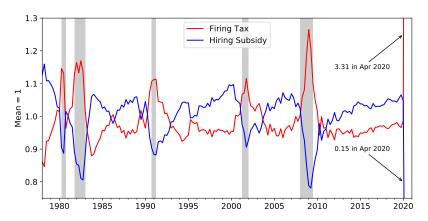
Policy Implications

Employment stabilization policies in Great Recession fell into two categories:

- 1. Policies promoting job creation
 - e.g. Social security contribution relief on new hires: France, Portugal,
 Spain
- 2. Policies discouraging job destruction
 - e.g. Short-Time Work schemes: Germany, Italy, Belgium

Time-Varying Policy Effectiveness

Impact on employment of hiring subsidy/firing tax equal to 25% of average quarterly wage:



Relation to the Literature

Lumpy employment adjustment:

- Bentolila & Bertola (1990)
- Hopenhayn & Rogerson (1993)
- Caballero, Engel, & Haltiwanger (1997)
- Campbell & Fisher (2000)
- Elsby & Michaels (2013)
- Fujita & Nakajima (2016)
- Cooper, Meyer, & Schott (2017)

Lumpy adjustment and time-varying responsiveness:

- Caballero & Engel (2013) (Investment)
- Berger & Vavra (2015) (Durable Consumption)

Conclusion

- Job creation and destruction rates exhibit time-varying responsiveness:
 - ▶ Job creation is more responsive when employment growth is high.
 - Job destruction is more responsive when employment growth is low.
- Heterogeneous-firm model with linear hiring costs can explain these facts.
- Targeting the job destruction margin is likely to be the most effective policy during recessions.

Computational Details

- Model solved using Krusell-Smith algorithm.
- Firm's problem:

$$\begin{split} V(z_r,z_i,n;A,N) &= \max_{n'} p\{Az_rz_in^\alpha - w(N)n - \kappa \Delta n'\mathbb{1}\{n'>n\}\} \\ &+ \beta \mathbb{E}[V(z_r',z_i',n';A',N')] \\ &\text{subject to} \\ &\log N' = \delta_N^0 + \delta_N^1 \log N + \delta_N^2 \log A \\ &\log p = \delta_N^0 + \delta_N^1 \log N + \delta_N^2 \log A \end{split}$$

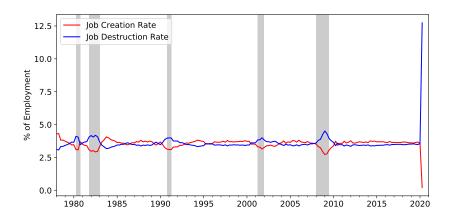
▶ *N* is pre-determined, and wage only depends on *N*. Therefore, each period only need to do market clearing on *p*.

Krusell-Smith Accuracy Tests

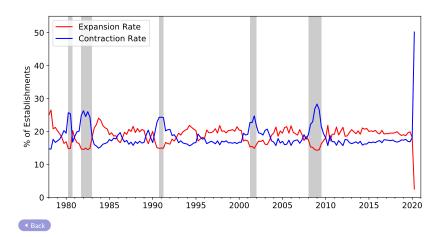
	Baseline	Frictionless
a_N	0.001	-0.004
b_N	0.515	0.000
c_N	0.555	1.170
a_p	0.365	N/A
b_p	-0.184	N/A
c_p	-1.569	N/A
R_N^2	0.99982	0.99999
R_N^2 R_p^2	0.99997	N/A
Max Error N(%)	0.17	0.11
Mean Error N (%)	0.04	0.10
Max Error p (%)	0.11	N/A
Mean Error p (%)	0.04	N/A

Notes: Mean/maximum errors constructed by simulating the model for 5000 periods and comparing *p* and *N* series from the model with those from the forecasting rules.

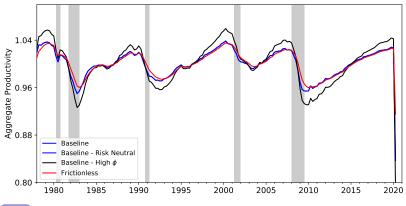
Model-Implied JC/JD Rates



Model-Implied Expansion/Contraction Rates

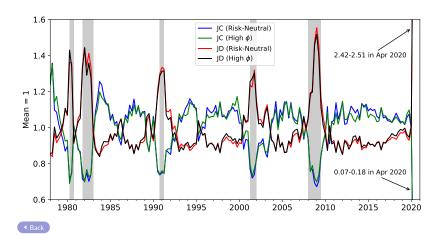


Model-Implied Aggregate Productivity Series



◆ Back

Robustness



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Caballero, Engel, & Haltiwanger (1997) Figure 1

