## Structural Unemployment

#### Benedikt ${\sf Herz}^1$ and ${\sf Thijs}\;{\sf van}\;{\sf Rens}^2$

UNIL Université de Lausanne

Faculté des Hautes Etudes Commerciales (HEC)

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<sup>1</sup>Universitat Pompeu Fabra

<sup>2</sup>CREI, Universitat Pompeu Fabra and Barcelona GSE

#### Mismatch and structural unemployment

Firms have jobs, but can't find appropriate workers. The workers want to work, but can't find appropriate jobs. There are many possible sources of **mismatch**—geography, skills, demography—and they are probably all at work.

Whatever the source, though, it is hard to see how the Fed can do much to cure this problem. Monetary stimulus has provided conditions so that manufacturing plants want to hire new workers. But the Fed does not have a means to transform construction workers into manufacturing workers.

*Given the structural problems in the labor market, I do* **not expect unemployment to decline rapidly**.

Kocherlakota. Inside the FOMC, Speech in Marquette, MI, Aug 17, 2010

## This paper

- Model of segmented labor market
  - Search frictions within segments  $\Rightarrow$  frictional unemployment
  - Mismatch across segments  $\Rightarrow$  mismatch unemployment
- A structural increase in mismatch unemployment?
  - Less cyclical? More persistent?
  - More important in the Great Recession?
- Occompose mismatch unemployment into its sources
  - Worker and job mobility costs
  - Wage adjustment costs

# Model

## Model of mismatch unemployment

- Labor market consists of segments (submarkets)
  - Each worker searches in one submarket
  - Each vacancy searches in one submarket
- Search frictions within submarkets
  - Matching technology with diminishing returns
  - Submarket *i* characterized by  $p_i$ ,  $S_i^W$ ,  $q_i$ ,  $S_i^J$
- Adjustment costs between submarkets
  - $\Rightarrow$  Dispersion labor market conditions
  - $\Rightarrow$  Mismatch unemployment

## Model of mismatch unemployment



## Worker mobility

• Value of searching in segment i

$$\mathsf{z}^W_i = \mathsf{b}_i + \mathsf{p}_i S^W_i$$

• Arbitrage through worker mobility

$$z_i^W = \bar{z}^W \Rightarrow p_i S_i^W = \bar{z}^W - b_i$$

- Unemployed workers move to more attractive segment
- Decreases  $p_i$ , decreases  $S_i^W$  (wage)
- Worker mobility curve: attractive jobs are hard to find

$$\hat{p}_i + \hat{S}^W_i = \alpha^{WM}_i$$

- Deviations from WM curve  $\Rightarrow$  mismatch
  - Differences in unemployment benefits:  $\alpha_i^{WM} = -\frac{\bar{b}}{\bar{z}^W \bar{b}} \hat{b}_i$
  - Worker mobility costs

## Sources of mismatch unemployment

Worker mobility

$$\hat{p}_i + \hat{S}^W_i = \alpha^{WM}_i$$

• Differences in unemployment benefits:  $\alpha_i^{WM} = -\frac{\bar{b}}{\bar{z}^W - \bar{b}} \hat{b}_i$ 

- Worker mobility costs
- Ø Job mobility

$$\hat{q}_i + \hat{S}_i^J = \alpha_i^{JM}$$

• Differences in vacancy posting costs:  $\alpha_i^{JM} = rac{ar{k}}{ar{z}^J + ar{k}} \hat{k}_i$ 

Job mobility costs (costs of moving vacancy)

Wage setting

$$\hat{S}_i^W = \hat{S}_i^J + \alpha_i^{WB}$$

- Differences in effective bargaining power:  $\alpha_i^{WB} = \frac{\phi_i}{1-\phi_i}$
- Wage adjustment costs (rebargaining costs, wage rigidities)
- Matching technology

$$\hat{q}_i = -\mu \hat{ heta}_i = -rac{\mu}{1-\mu} \hat{p}_i + lpha_i^{MF}$$

• Differences in matching technology:  $\alpha_i^{MF} = \frac{\hat{B}_i}{1-\mu} - \frac{\mu}{1-\mu} \left(\bar{p} - \bar{q}\right) \hat{\mu}_i$ 

#### Sources of mismatch unemployment

Worker mobility

$$\hat{p}_i + \hat{S}_i^W = \alpha_i^{WM}$$

Iob mobility

$$\hat{q}_i + \hat{S}_i^J = \alpha_i^{JM}$$

Wage setting

$$\hat{S}_i^W = \hat{S}_i^J + \alpha_i^{WB}$$

Matching technology

# Data and Measurement

### Measuring match surplus

• BE for match surplus

$$(1+r) S_{it} = y_{it} + (1-\tau_{it}) E_t S_{it+1}$$

- Match payoff:  $y_{it}^W = w_{it} b_{it}$ ,  $y_{it}^J = \pi_{it} + k_{it}$  (CPS, NIPA)
- Turnover:  $\tau_{it}^W = \lambda_{it} + p_{it}$ ,  $\tau_{it}^J = \lambda_{it} + q_{it}$  (CPS)

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• Turnover: 
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,  $au_{it}^J = \lambda_{it} + q_{it}$  (CPS)

Solving forward

$$S_{it} = \frac{y_{it}}{r + \tilde{\tau}_{it}}$$

- Turnover constant over duration match
- State-level wages and profits random walk (Blanchard and Katz 1992; Haefke, Sonntag and van Rens 2008)
- Unemployment benefit  $b_i/w_i = 0.73$  (Mortensen and Nagypal 2007) Vacancy posting costs  $k_i/\pi_i = 0.03$  (Silva and Toledo 2009) Discount rate r = 5% per annum

# Measuring match surplus: Bellman equations

Workers

$$W_{it} = \frac{1}{1+r} \left[ w_{it} + \lambda_{it} E_t U_{it+1}^W + (1 - \lambda_{it}) E_t W_{it+1} \right]$$
$$U_{it}^W = \frac{1}{1+r} \left[ b_{it} + p_{it} E_t W_{it+1} + (1 - p_{it}) E_t U_{it+1}^W \right]$$
$$S_{it}^W = W_{it} - U_{it}^W = \frac{1}{1+r} \left[ w_{it} - b_{it} + (1 - \lambda_{it} - p_{it}) E_t S_{it+1}^W \right]$$

Jobs

$$J_{it} = \frac{1}{1+r} \left[ \pi_{it} + \lambda_{it} E_t U_{it+1}^J + (1 - \lambda_{it}) E_t J_{it+1} \right]$$
$$U_{it}^J = \frac{1}{1+r} \left[ -k_{it} + q_{it} E_t J_{it+1} + (1 - q_{it}) E_t U_{it+1}^J \right]$$
$$S_{it}^J = J_{it} - U_{it}^J = \frac{1}{1+r} \left[ \pi_{it} + k_{it} + (1 - \lambda_{it} - q_{it}) E_t S_{it+1}^J \right]$$

#### Measuring match surplus: persistence

• BE for match surplus

$$(1+r) S_{it} = y_{it} + (1-\tau_{it}) E_t S_{it+1}$$

- Match payoff:  $y_{it}^W = w_{it} b_{it}$ ,  $y_{it}^J = \pi_{it} + k_{it}$  (CPS, NIPA)
- Turnover:  $au_{it}^W = \lambda_{it} + p_{it}$ ,  $au_{it}^J = \lambda_{it} + q_{it}$  (CPS)
- Assumptions that matter
  - Level and persistence of payoffs (wages and profits)

$$y_{it+1} = (1-\delta) y_{it} + \delta \bar{y}_t$$

- Level of turnover:  $au_{it+s} = au_{it}$  or  $au_{it+s} = ar{ au}_t$
- Solving forward

$$S_{it} = \frac{\bar{y}_t}{r + \tilde{\tau}_{it}} + \frac{y_{it} - \bar{y}_t}{r + \tilde{\tau}_{it} + \delta}$$

- Worker heterogeneity
  - 40 homogeneous groups of workers based on observables (2 gender x 5 education x 4 potential experience)
  - Calculate  $\hat{S}_i^W$  separately for 40 groups, then average
  - Same for  $\hat{S}_{i}^{J}$ , assuming  $\log \pi_{it}^{*} = \log \pi_{it}^{\mathsf{NIPA}} \log w_{it}^{\mathsf{CPS}} + \log w_{it}^{*\mathsf{CPS}}$
- Compensating differentials
  - Job characteristics not observable
  - Assume constant over time  $\Rightarrow$  state-specific FE
  - $\Rightarrow$  Do not interpret *level* of mismatch unemployment

## Measuring job and worker finding rates

- Job finding rate p<sub>i</sub>
  - Observe by state, 1967-2009 (CPS)
- Worker finding rate q<sub>i</sub>
  - Observe directly, 2000-2009 (JOLTS, confidential)
  - Assume matching technology constant across states

$$\hat{q}_i = -\mu \hat{ heta}_i = -rac{\mu}{1-\mu} \hat{p}_i$$

- Elasticity matching function  $\mu=$  0.6 (Mortensen and Nagypal 2007)
- Heterogeneity: control same as for surplus

# Results

#### Sources of mismatch unemployment

Worker mobility

$$\hat{p}_i + \hat{S}_i^W = \alpha_i^{WM}$$

Iob mobility

$$\hat{q}_i + \hat{S}_i^J = \alpha_i^{JM}$$

Wage setting

$$\hat{S}_i^W = \hat{S}_i^J + \alpha_i^{WB}$$

Matching technology

$$\begin{aligned} \hat{q}_{i} &= -\mu \hat{\theta}_{i} = -\frac{\mu}{1-\mu} \hat{p}_{i} + \alpha_{i}^{MF} \\ &\downarrow \\ \hat{p}_{i} &= (1-\mu) \left( \alpha_{i}^{WM} - \alpha_{i}^{JM} - \alpha_{i}^{WB} + \alpha_{i}^{MF} \right) \end{aligned}$$

### Worker mobility



B. Herz (UPF) & T. van Rens (CREI, UPF)

# Job mobility





#### Worker mobility



# Job mobility



B. Herz (UPF) & T. van Rens (CREI, UPF)

## Wage setting



## Deviations from worker mobility curve

| States with larg | States with largest difference |      | distance $(miles)$ |  |
|------------------|--------------------------------|------|--------------------|--|
| Wyoming          | Alaska                         | 0.86 | 2297               |  |
| Wyoming          | Massachusetts                  | 0.61 | 1798               |  |
| Wyoming          | New York                       | 0.61 | 1565               |  |
| Alaska           | Florida                        | 0.59 | 3840               |  |
| Wyoming          | Kansas                         | 0.57 | 552                |  |
| Average distan   | ce                             |      | 2010               |  |

| $States \ with \ smallest \ difference$ |          | $\alpha_i^{WM} - \alpha_j^{WM}$ | distance (miles) |
|---|----------|---------------------------------|------------------|
| South Dakota                            | DC       | 0.0001                          | 1239             |
| North Dakota                            | Ohio     | 0.0005                          | 994              |
| Louisiana                               | Kentucky | 0.0005                          | 589              |
| New Mexico                              | Indiana  | 0.0011                          | 1138             |
| North Dakota                            | Utah     | 0.0011                          | 797              |
| Average distance                        | l.       |                                 | 952              |

- Level of disaggregation
  - 37 industries, SIC, 1979-2002
  - 35 industries, NAICS, 1997-2009
- Job finding rate by industries Where do unemployed workers search?
  - Industry where they last held a job (BLS)
  - In industry where they find a job (robustness)
- Everything else same as for states

## Worker mobility



# Job mobility



# Wage setting



| Industries with largest difference |                                      | $\alpha_i^{WM} - \alpha_j^{WM}$ |
|------------------------------------|--------------------------------------|---------------------------------|
| Broadcasting and telecom           | Machinery manufacturing              | 1.07                            |
| Broadcasting and telecom           | Chemical manufacturing               | 1.03                            |
| Broadcasting and telecom           | Publishing (except internet)         | 0.99                            |
| Broadcasting and telecom           | Furniture and fixtures manufacturing | 0.97                            |
| Broadcasting and telecom           | Textile, apparel, and leather manuf. | 0.90                            |

| Industries with smallest difference |  | $\alpha_i^{WM} - \alpha_j^{WM}$ |
|-------------------------------------|--|---------------------------------|
| Transportation and warehousing      | Motion picture and sound recording     | 0.00017                         |
| Wholesale trade                     | Nonmetallic mineral product manuf.     | 0.0008                          |
| Accommodation                       | Computer and electronic product manuf. | 0.00116                         |
| Retail trade                        | Food services and drinking places      | 0.00133                         |
| Miscellaneous manufacturing         | Arts, entertainment, and recreation    | 0.00138                         |

|                                    | across states |          | across industries     |          |          |                           |
|------------------------------------|---------------|----------|-----------------------|----------|----------|---------------------------|
|                                    | WM costs      | JM costs | ${ m WB}\ { m costs}$ | WM costs | JM costs | $\operatorname{WB}$ costs |
| baseline                           | 0.15          | 0.18     | 0.33                  | 0.31     | 0.29     | 0.61                      |
| no comp diff                       | 0.35          | 0.29     | 0.53                  | 0.98     | 0.92     | 1.41                      |
| $\mu = 0.5$                        | 0.15          | 0.16     | 0.26                  | 0.30     | 0.26     | 0.55                      |
| $\mu = 0.7$                        | 0.17          | 0.21     | 0.46                  | 0.33     | 0.36     | 0.70                      |
| $b_{it}/w_{it} = 0.4$              | 0.10          | 0.18     | 0.34                  | 0.21     | 0.29     | 0.55                      |
| $b_{it}/w_{it} = 0.95$             | 0.69          | 0.18     | 0.60                  | 1.83     | 0.29     | 1.85                      |
| $\tilde{\tau}_{it} = \bar{\tau}_t$ | 0.21          | 0.28     | 0.14                  | 0.17     | 0.30     | 0.27                      |

#### Mismatch unemployment

 $\bullet$  Adjustment costs  $\Rightarrow$  dispersion in job finding rates

$$\hat{p}_i = (1 - \mu) \left( \alpha_i^{WM} - \alpha_i^{JM} - \alpha_i^{WB} 
ight)$$

• Dispersion  $\Rightarrow$  lower average job finding rate

$$\frac{\bar{p}'}{\bar{p}} = \left(\frac{E\left[\left(1+\hat{p}_i\right)^{\frac{1}{1-\mu}}\right]}{E\left[\left(1+\hat{p}'_i\right)^{\frac{1}{1-\mu}}\right]}\right)^{1-\mu}$$

- $\bar{p}' < \bar{p} \Leftrightarrow \theta_i$  mean-preserving spread of  $\theta_i'$
- Concavity job finding rate in  $\theta_i$  determines size effect
- Counterfactual unemployment rate:  $ar{u}=rac{ar{\lambda}}{ar{\lambda}+ar{p}}$

#### Contribution mismatch to unemployment: across states



#### Results

- Mismatch is large, but contributes little to unemployment
  - Defining submarkets (level disaggregation) is crucial
  - 50 states, 40 industries
- A structural increase in mismatch unemployment?

#### Unemployment due to mismatch across states



#### Unemployment due to mismatch across states



#### Unemployment due to mismatch across states



#### Unemployment due to mismatch across industries



#### Unemployment due to mismatch across industries



#### Unemployment due to mismatch across industries



#### Results

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  - Equally cyclical, no more persistent
  - Great Recession similar to previous recessions
- Sources of mismatch unemployment

#### Sources of mismatch across states



## Sources of mismatch across industries



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- Sources of mismatch unemployment
  - Wage adjustment costs most important source
  - Encouraging worker mobility likely to have small effects

#### Adjustment costs may offset each other

• Total effect depends on correlation

$$\hat{p}_i = (1 - \mu) \left( \alpha_i^{WM} - \alpha_i^{JM} - \alpha_i^{WB} \right)$$

- High  $\alpha_i^{WB}$  states (industries) have high wages, all else equal
  - $\bullet\,$  Attractive to workers  $\Rightarrow$  want to move in
  - Unattractive to firms  $\Rightarrow$  want to move vacancies out
  - Worker and job mobility costs prevent this from happening
- Removing mobility costs may increase unemployment
  - High  $\alpha_i^{WM}$  states (industries) have relatively many unemployed workers
  - High  $\alpha_i^{JM}$  states (industries) have relatively many vacancies

#### Adjustment costs across states



## Adjustment costs across industries



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