
**EFFICIENCY AND
LABOR MARKET DYNAMICS IN A MODEL OF
LABOR SELECTION**

APRIL 18, 2017

MOTIVATION

- ❑ **Goals of project**

- ❑ **Explain business-cycle volatility in unemployment and job-finding**

- ❑ **Using efficient allocations**
 - ❑ (Avoid wage formation altogether...)

- ❑ **Using costs of hiring distinct from vacancy posting costs**

- ❑ (Note: Efficient allocations in “baseline” search and matching framework will not get us there)
 - ❑ “Shimer puzzle”

METHODOLOGY

- ❑ Methodology of project
- ❑ **Exploit cross-sectional heterogeneity amongst (potential) new hires' characteristics**
- ❑ **Discipline with micro-economic data**
- ❑ **Micro-data about cross-sectional heterogeneity?**
 - ❑ **Person i -specific productivity difficult (impossible?) to measure (“How much can person i produce?”)**

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- ❑ **Micro-data about cross-sectional heterogeneity?**
 - ❑ Person i -specific productivity difficult (impossible?) to measure (“How much can person i produce?”)
 - ❑ Wage data easily available
 - ❑ **BUT** our model **intentionally avoids** how wages are determined
 - ❑ (return to this point soon...)
- ❑ **Our framework uses micro-level “match quality” data**
 - ❑ **Costs of “integrating” / “training” potential new hires**

DATA

- ❑ **Empirics**
- ❑ **Cost of training / hiring**
 - ❑ **Apply only in the first period of employment**
 - ❑ **As new workers learn the methods of their new firm**
- ❑ **Incumbent workers incur zero training costs**
- ❑ **Real life examples of training costs**
 - ❑ **Shadowing other workers to observe how job is performed**
 - ❑ **Understanding the culture of the firm**
 - ❑ **Computer setup and configurations**
 - ❑ **Etc...**
- ❑ **Barron, Black, and Loewenstein (1989 *JLE*)**
 - ❑ **Firm-level costs of interviewing/hiring/training/integrating new workers**
 - ❑ **Based on 1982 EOPP (Employment Opportunities Pilot Project)**
 - ❑ **"...workers of varying abilities are matched to positions with different training requirements."**

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- ❑ **Reports first moments and cross-sectional second moments**

- ❑ **(Any other evidence on cross-sectional second moments?...)**

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- ❑ **(Any other evidence on cross-sectional second moments?...)**

- ❑ **1982 EOPP data continues to be used in various applications**
 - ❑ **Different investment in match-specific capital for different education groups (Cairo and Cajner, 2013 WP)**

 - ❑ **Size of labor turnover costs (relevant for search and matching models) – e.g., Silva and Toledo (2009 *MD*)**

 - ❑ **Effects of training costs of firm-specific labor turnover (Dolfin 2006 *Applied Economics*)**

MODEL

- **Main components**
 - Fixed cost γ^h of “training” each new hire (systematic component)
 - **Idiosyncratic training cost for each new hire i**

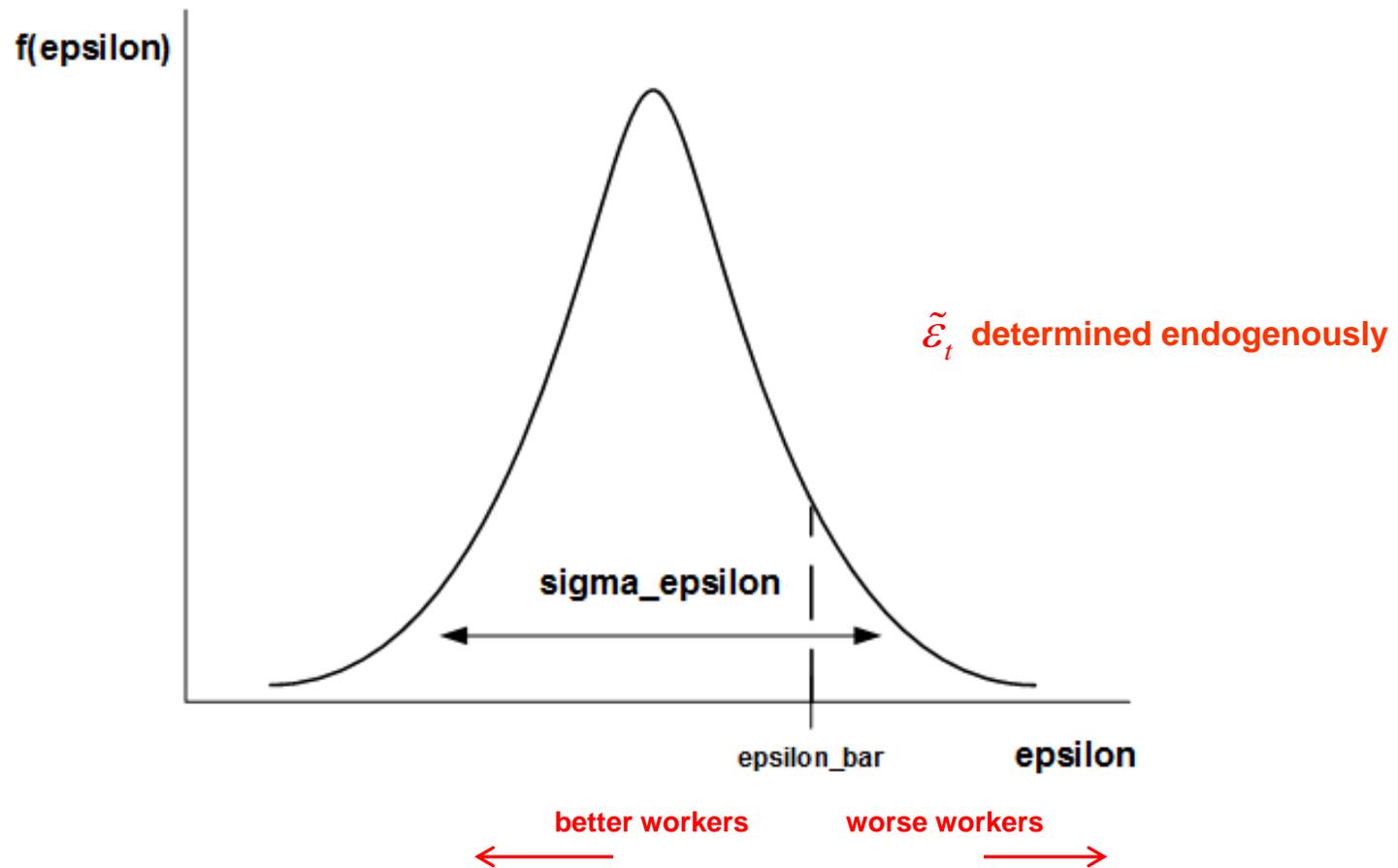
- **Total training cost for new worker i in period $t = \gamma^h + \varepsilon^i$**

↑
Idiosyncratic training/residual cost for new hire i

$$\varepsilon^i \sim \text{iid } N(0, \sigma_\varepsilon^2)$$

DISTRIBUTION

- Cross-sectional distribution of training costs in period t



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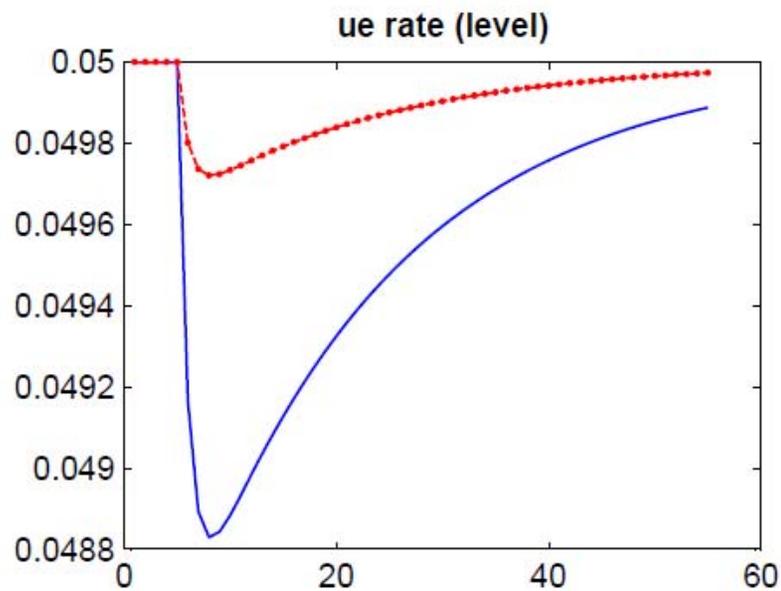
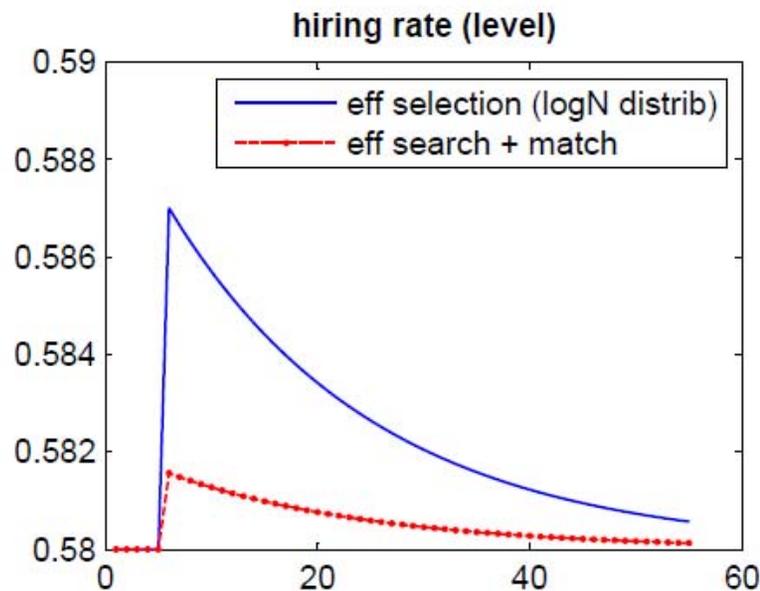
- ❑ **Cross-sectional SD σ_ε informed by Barron et al**

- ❑ **Dispersion of training costs considered a primitive**
 - ❑ (Similar to matching function taken as primitive in DMP-based models)

- ❑ **No search and matching component**
 - ❑ To focus on the **endogenous selection** component
 - ❑ Davis, Faberman, and Haltiwanger (2013 *QJE*): Evidence of heavily reliance on other margins for hiring in addition to vacancy postings (JOLTS)

MAIN RESULTS AND CONTRIBUTIONS

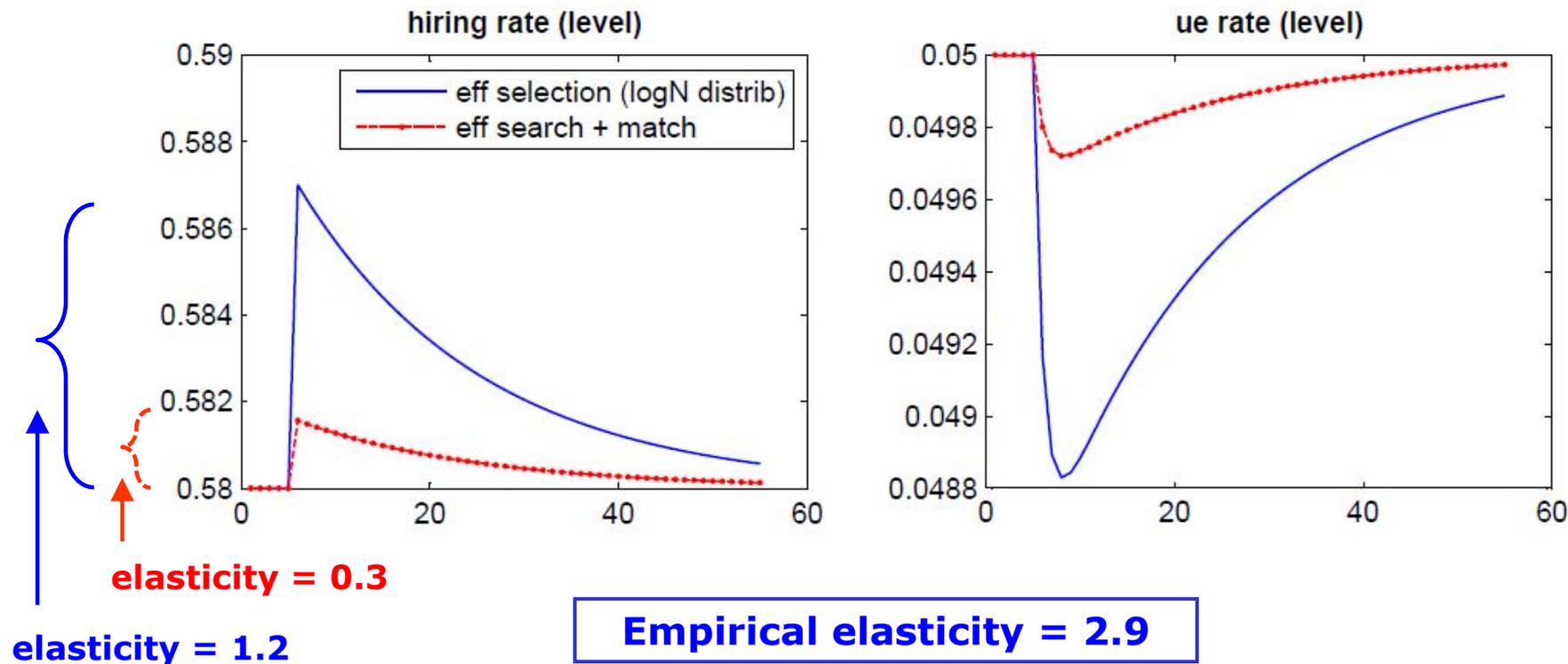
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 - **No wage decentralization in model**
 - **Conditional on TFP shocks**



MAIN RESULTS AND CONTRIBUTIONS

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EX-ANTE VS. EX-POST HIRING COSTS

- ❑ **Think about selection model as hiring candidates who have the “best skills”**
- ❑ **Interpret “matching process” as a costly “contact process” or “meeting process”**

firm pays cost to
advertise it is hiring



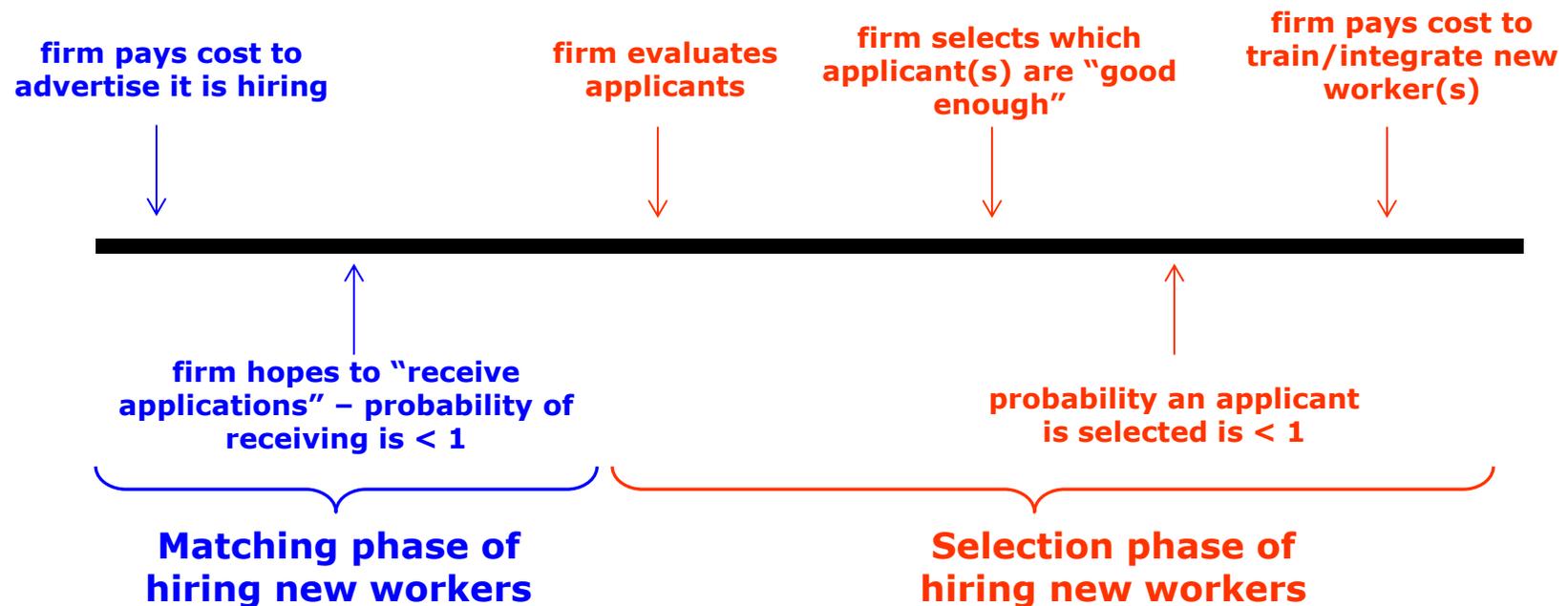
firm hopes to “receive
applications” – probability of
receiving is < 1



**Matching phase of
hiring new workers**

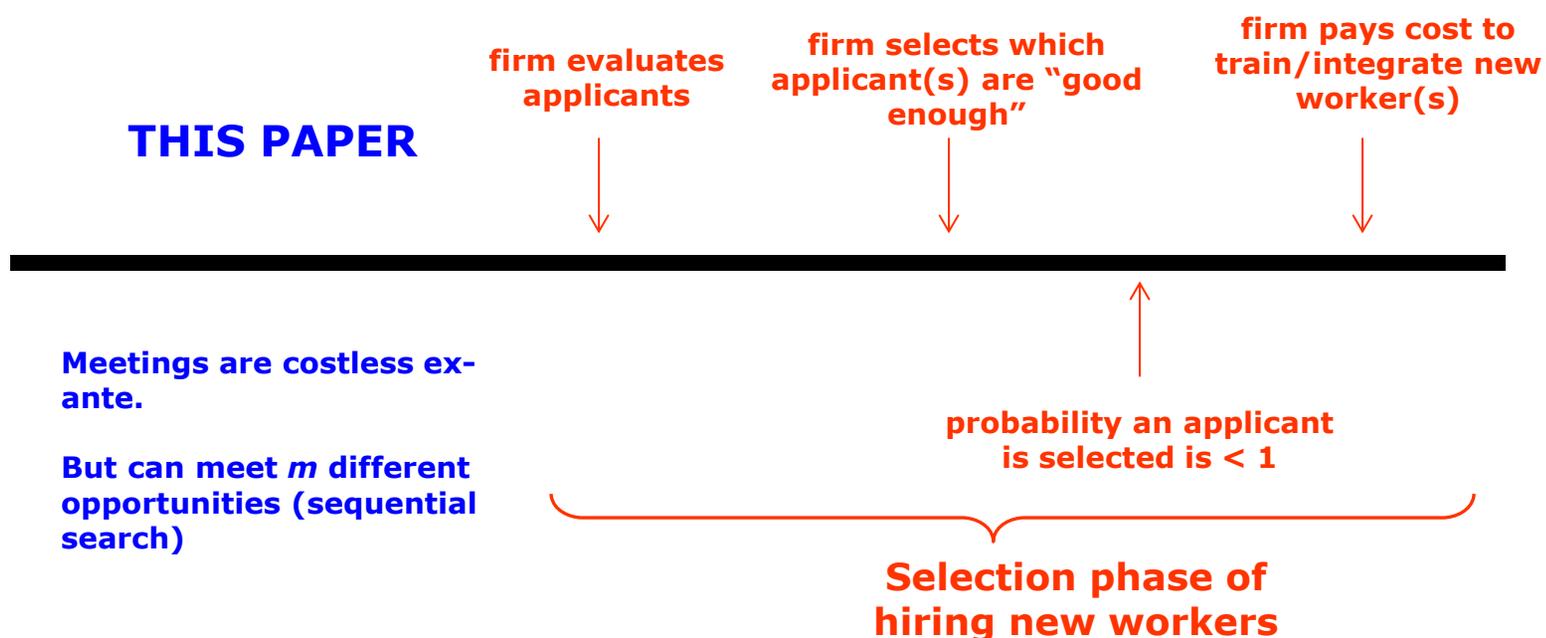
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- ❑ But also allow other costs in the hiring of workers



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- ❑ But also allow other costs in the hiring of workers



- ❑ Baseline: Each unemployed individual meets only one firm in any period
- ❑ Can generalize to allow N Poisson meetings per period (i.e., $N=2$, $N=3$, ...)

MAIN RESULTS AND CONTRIBUTIONS

- ❑ **Efficient volatility occurs and is meaningful**
 - ❑ No wage decentralization in model
 - ❑ Conditional on TFP shocks

- ❑ **Elasticity of hiring rate wrt TFP: Empirical value = 2.9**
 - ❑ Has not appeared in literature (as far as we know...)
 - ❑ Constructed using data from Shimer (2005) and Michailat (2012)
 - ❑ (Potentially?) another contribution

- ❑ **Endogenous value from previous example: 1.2**

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- ❑ **Caveat / Question**
 - ❑ Depends on the data we employ to calibrate SD σ_ε ...
 - ❑ ... we use training cost dispersion
 - ❑ **What if we use new hires' wage dispersion as "upper bound" on SD σ_ε ?**

σ_ε from wage dispersion = 1.5 σ_ε from hiring cost dispersion

→ Volatility results dampen a tiny bit...

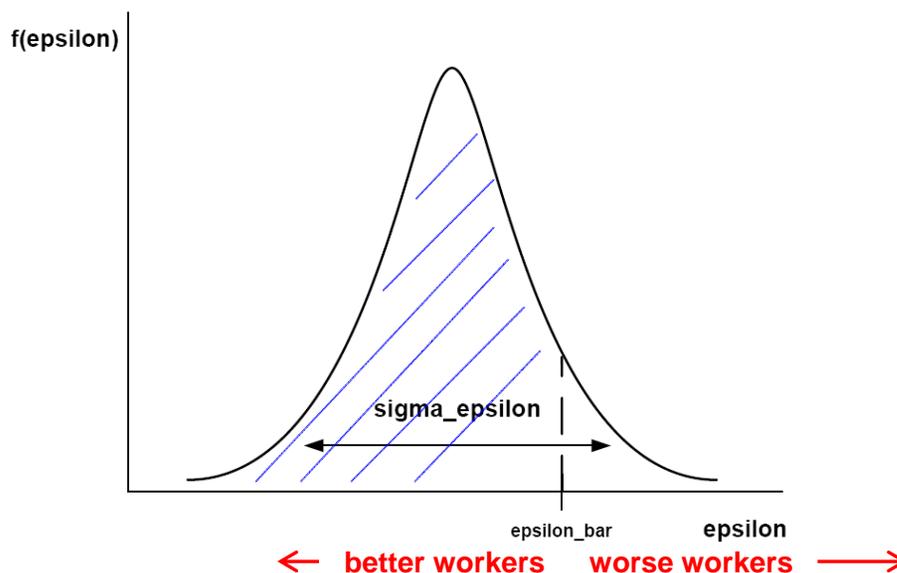


Definitions

SELECTION MARGIN

- **Optimal decision characterized by cutoff rule**
 - **Choose endogenous threshold $\tilde{\varepsilon}_t$ below which everybody is selected to work**
- **CDF (hiring rate, aka selection rate, aka job-finding rate)**

$$\eta(\tilde{\varepsilon}_t) = \int_{\varepsilon_t^i \leq \tilde{\varepsilon}_t} f(\varepsilon_t^i) \cdot d\varepsilon_t^i$$



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- **Training cost for threshold new worker = $\gamma^h + \tilde{\varepsilon}_t$**
- **Average idiosyncratic training costs for those individuals who are hired**

$$H(\tilde{\varepsilon}_t) = \int_{\varepsilon_t^i \leq \tilde{\varepsilon}_t} \varepsilon_t^i f(\varepsilon_t^i) \cdot d\varepsilon_t^i$$



Social Planner Model (Partial Equilibrium)

EFFICIENT SELECTION

□ Dynamic surplus maximization problem

$$\max_{\{n_t, \tilde{\varepsilon}_t\}} E_0 \sum_{t=0}^{\infty} \left(\frac{1}{1+r} \right)^t \left[z_t n_t + s_t (1 - \eta(\tilde{\varepsilon}_t)) \overset{\text{non-market payoff}}{\downarrow} b - s_t \eta(\tilde{\varepsilon}_t) \left(\gamma^h + \frac{H(\tilde{\varepsilon}_t)}{\eta(\tilde{\varepsilon}_t)} \right) \right]$$

$$n_t = (1 - \rho)n_{t-1} + s_t \eta(\tilde{\varepsilon}_t)$$

$$s_t = lfp - (1 - \rho)n_{t-1}$$

lfp fixed in partial equilibrium

EFFICIENT ALLOCATION

□ **Definition:** efficient allocations are endogenous processes $\{\tilde{\varepsilon}_t, n_t\}_{t=0}^{\infty}$ that satisfy

□ **Selection condition**

$$\underbrace{\gamma^h + \tilde{\varepsilon}_t}_{\text{Asset value of a new worker}} = z_t - b + \left(\frac{1-\rho}{1+r} \right) E_t \left\{ \underbrace{H(\tilde{\varepsilon}_{t+1}) - \tilde{\varepsilon}_{t+1}\eta(\tilde{\varepsilon}_{t+1})}_{\text{Expected social cost of a replacement new worker hired in } t+1} + \underbrace{\gamma^h + \tilde{\varepsilon}_{t+1}}_{\text{Asset value of a replacement new worker}} \right\}$$

□ **Law of motion for aggregate labor**

$$n_t = (1-\rho)n_{t-1} + s_t\eta(\tilde{\varepsilon}_t)$$

taking as given initial labor n_{-1} and exogenous stochastic process $\{z_t\}_{t=0}^{\infty}$



Shape of Distribution

Slope of Distribution at Threshold

How to Calibrate σ_ε

ELASTICITIES

- **Steady-state elasticities**

- **Elasticity of selection threshold wrt TFP**

$$\frac{\partial \ln \tilde{\varepsilon}}{\partial \ln z} = \frac{z}{\tilde{\varepsilon}} \cdot \frac{1+r}{r+\rho+(1-\rho)\eta(\tilde{\varepsilon})}$$

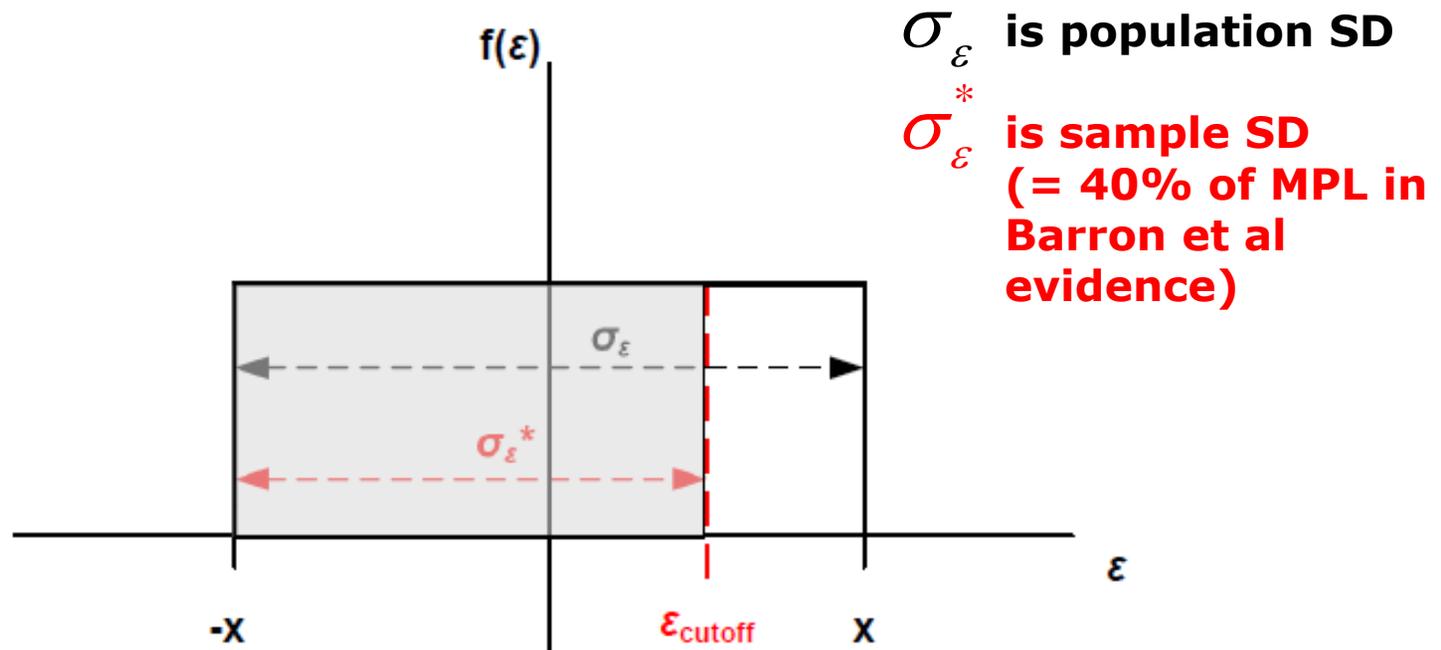
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- **Empirical data to measure slope at endogenous cutoff point $\tilde{\varepsilon}$?**
- **Depends on shape of distribution...**

UNIFORM DISTRIBUTION

- Warm-up example
- $\eta'(\tilde{\varepsilon})$ independent of ε_i



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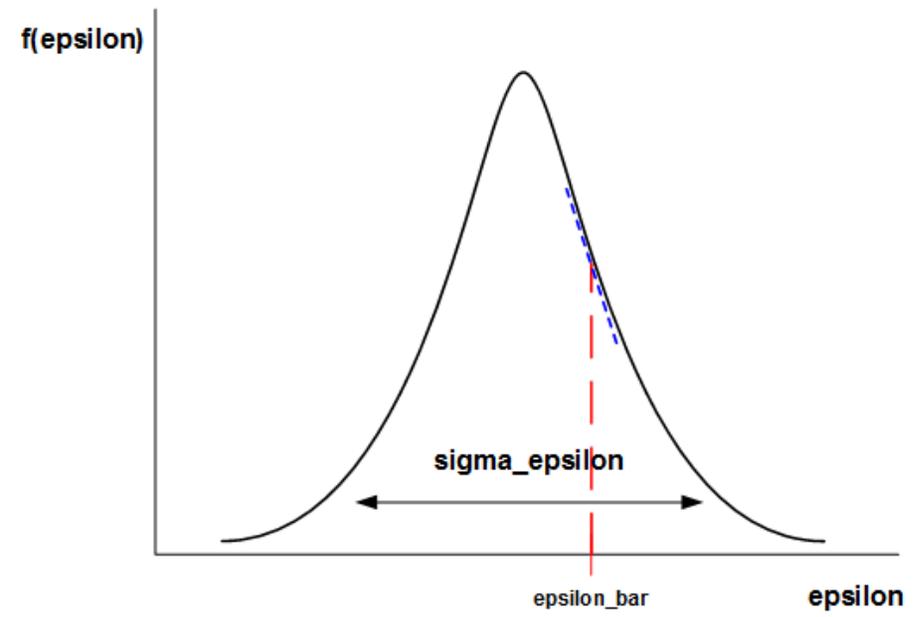
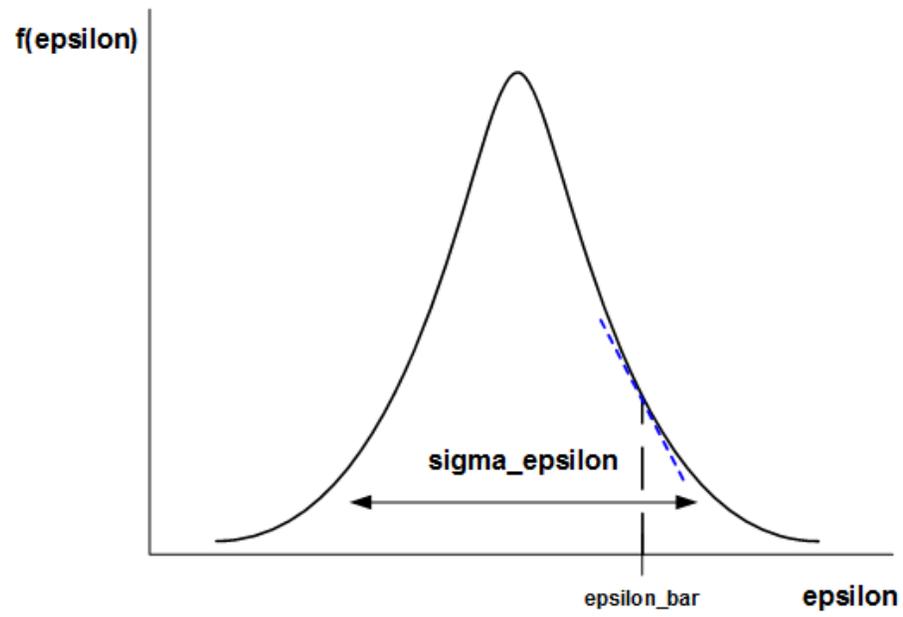
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Warm-up example

U[-1.2, 1.2] $\rho = 0.1$
 $r = 0.01$
 $\eta(\varepsilon) = 0.58$

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 (= 1.15)

Compared to 2.9 empirical elasticity



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Compared to 2.9 empirical elasticity

- ❑ **Two micro data sources to measure σ_{ε}^***
 - ❑ **Short-term training cost dispersion (EOPP: Employment Opportunity Pilot Project)**
 - ❑ **Wage dispersion for new hires**

CALIBRATION

- ❑ **Quantitative DSPE example**

- ❑ **Distribution of training costs assumed log-normal**
 - ❑ σ_ε chosen to hit cross-sectional SD of training costs of 40 percent of MPN

 - ❑ **Barron, Black, and Loewenstein (1989, p. 5): SD across new hires of training costs during first three months of employment = 207 hours (= 40% of MPL)**

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- ❑ **Calibrate γ^h to hit average hiring rate $\approx 58\%$ (a macro calibration approach)**
 - ❑ **Average hiring cost turns out $>$ Barron et al's measure (= 150 hours)**
 - ❑ **Nobody has negative training costs \rightarrow skewed distribution**

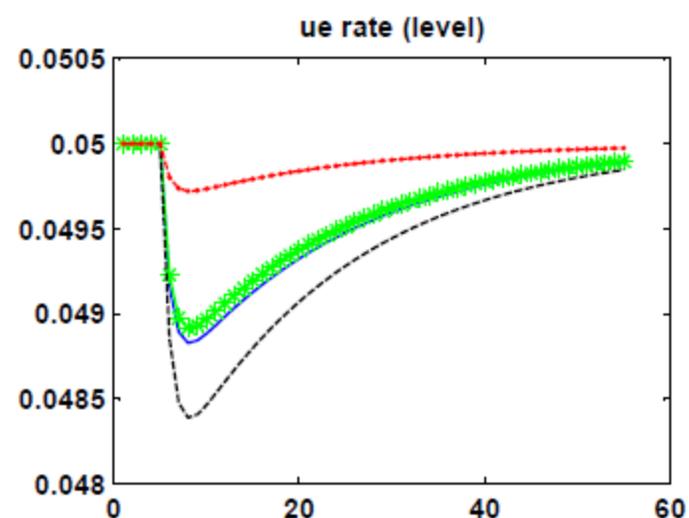
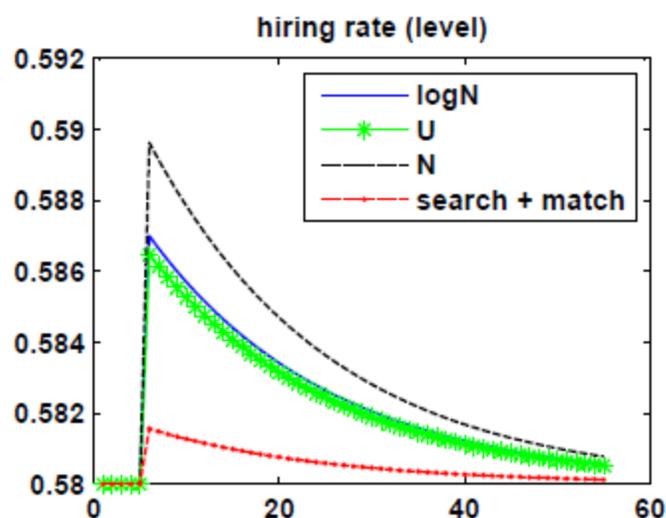
CALIBRATION

- Quantitative DSPE example
- Conventional parameters
 - $r = 0.01$
 - Standard quarterly TFP process
($\rho_z = 0.95, \sigma_z = 0.007$)
- Outside option b
- $b = 0$
- Doesn't matter at all for efficient allocations!

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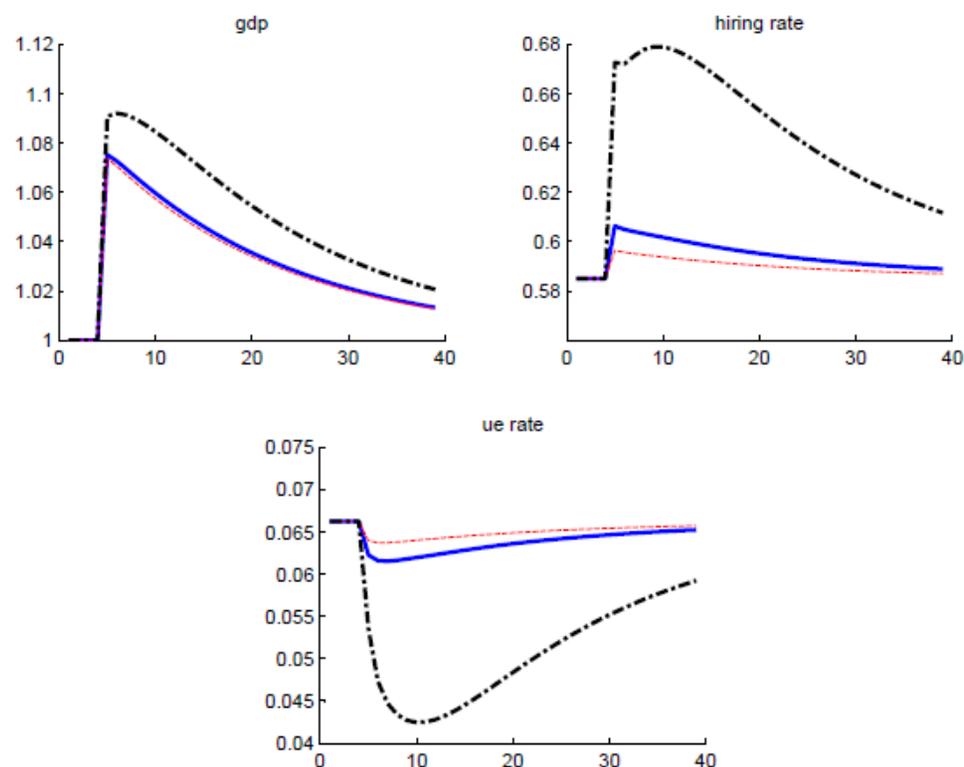
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- Outside option $b = 0$

- Various σ_ε values

- $= 0.2$
- $= 0.4$ (Barron et al)
- $= 0.6$ (wage dispersion)





Sequential Search

SELECTION AND SEQUENTIAL SEARCH

- ❑ The model readily admits **sequential search** (e.g., McCall (1970), Mortensen (1970))
- ❑ Suppose Poisson meetings **N** occur during a quarter
- ❑ Baseline considered: **$N = 1$**
- ❑ For **$N \geq 1$** meetings during period, job-acceptance _{j} condition modifies to

$$\eta(\tilde{\varepsilon}_t) = m(\tilde{\varepsilon}_t) \cdot \sum_{j=1}^N (1 - m(\tilde{\varepsilon}_t))^{j-1}$$

- ❑ $\eta(\tilde{\varepsilon}_t)$ is probability that a searching worker accepts a job within a quarter
- ❑ $m(\tilde{\varepsilon}_t)$ is chance that a searching worker accepts a particular contact m during a quarter

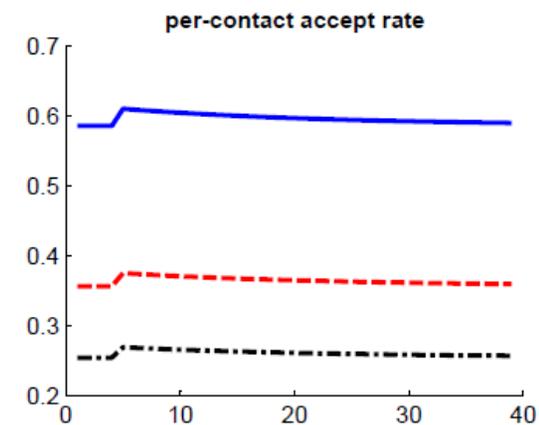
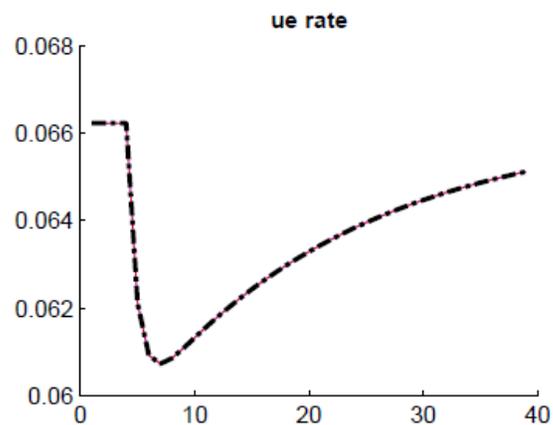
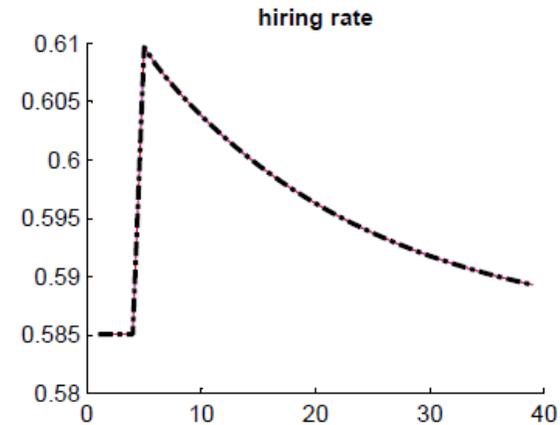
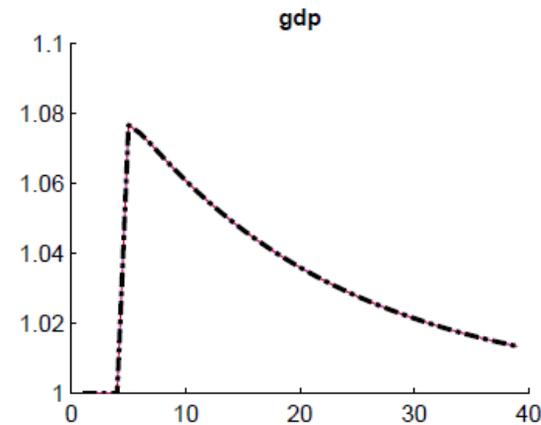
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 - ❑ $m = 1 \rightarrow \eta(\tilde{\varepsilon}_t) = m(\tilde{\varepsilon}_t)$
 - ❑ $m = 2 \rightarrow \eta(\tilde{\varepsilon}_t) = m(\tilde{\varepsilon}_t) \{1 + (1 - m(\tilde{\varepsilon}_t))\}$
 - ❑ $m = 3 \rightarrow \eta(\tilde{\varepsilon}_t) = m(\tilde{\varepsilon}_t) \{1 + (1 - m(\tilde{\varepsilon}_t)) + (1 - m(\tilde{\varepsilon}_t)) \cdot (1 - m(\tilde{\varepsilon}_t))\}$

SELECTION AND SEQUENTIAL SEARCH

- $m = 1$
- $m = 2$
- $m = 3$

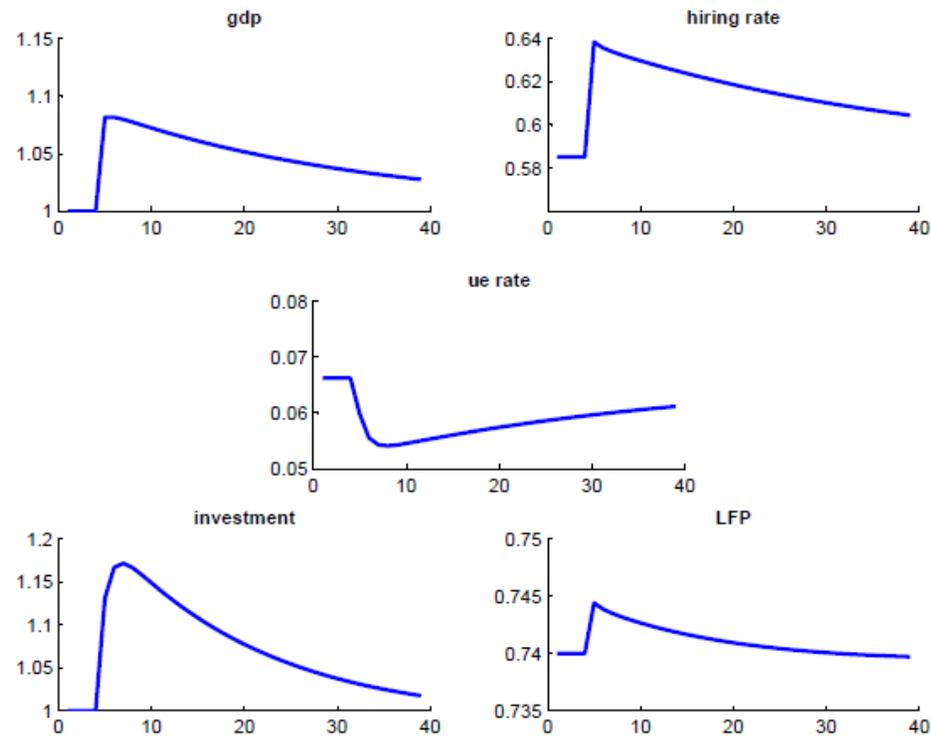




General Equilibrium

GENERAL EQUILIBRUM

- ❑ Endogenous labor supply (endogenous LFP)
- ❑ Physical capital investment
- ❑ (hence MRSs and MRTs nest textbook RBC model)



SUMMARY

- ❑ **Large fluctuations in labor markets induced by aggregate TFP**
 - ❑ Does **NOT** require any particular wage decentralization scheme
- ❑ **(General equilibrium model works the same way – see Table 2)**
- ❑ **Tractable to model in DSPE**

- ❑ **Seems a clever way of calibrating σ_ε**
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- ❑ **Labor selection and labor matching complementary mechanisms**
 - ❑ **Selection stresses cross-sectional issues**
("I hope this new worker integrates into the job easily")

 - ❑ **Sequential search and DMP search + matching stress intertemporal issues**
("I hope we find any suitable candidates at all")