The Cyclical Behavior of Equilibrium Unemployment and Vacancies
Shimer (2005)

Presented by: XING Mingjie

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Bellman Equation

Worker

- Unempolyment value

$$U_p = z + \delta\{f(\theta_p)\mathbb{E}_p W_{p'} + (1 - f(\theta_p))\mathbb{E}_p U_{p'}\} \tag{1}$$

- Employment value

$$W_p = w_p + \delta\{(1 - s)\mathbb{E}_p W_{p'} + s\mathbb{E}_p U_{p'}\} \tag{2}$$

Bellman Equation

Firm

- Hiring value

$$J_p = p - w_p + \delta(1-s)\mathbb{E}_p J_{p'} \tag{3}$$

- Vacancy value

$$V_p = -c + \delta q(\theta_p)\mathbb{E}_p J_{p'} \equiv 0 \tag{4}$$

Productivity

The log of productivity follows AR(1) process

$$\log(p) = \rho \log(p) + \varepsilon \tag{5}$$

where

$$\log(p) \sim N(\mu_\lambda, \sigma_\lambda^2),\ \varepsilon \sim N(\mu_\varepsilon, \sigma_\varepsilon^2)$$

Optimal Control

Market tightness

- Control in this problem consists of w_p, θ_p, u_p and the state is p
- Market tightness θ_p is given by solving the following equation of hire rate from free entry condition

$$q(\theta_p) = \frac{c}{\delta \mathbb{E}_p J_{p'}} \tag{6}$$

- And market tightness

$$\theta_p = (\frac{q(\theta_p)}{\mu})^{-\frac{1}{\eta}} \tag{7}$$

- Employ Rate is given by

$$f(\theta_p) = \mu^{\frac{1}{\eta}} q^{\frac{\eta-1}{\eta}} \tag{8}$$

Optimal Control
Continued

- Optimal wage at each productivity level is given by the Nash Bargaining:
$$W_p - U_p = \beta(W_p - U_p + J_p) \tag{9}$$
- Note Bellman Equation of W_p given by 2, U_p given by 1, J_p given by 3
- Following the algebra given in slide 16, optimal wage for each p is
$$w_p = \beta p + (1-\beta)z + \beta c\theta_p \tag{10}$$
- And unemployment rate
$$u_p = \frac{\delta}{\delta + f(\theta_p)} \tag{11}$$

Calibration

Parameter	Symbol	Value
Productivity std.	σ_{logp}	0.05
Productivity mean	μ_{logp}	1
Stochastic std.	σ_{ε}	0.03
Stochastic mean	μ_{ε}	0
Separation rate	s	0.1
Discount rate	r	0.012
Value of leisure	z	0.4
Matching function	μ	1.355
Matching function	α	0.72
Bargaining Power	β	0.72
Cost of vacancy	c	0.213

Table 1: Parameter Calibration

Question a I

Discretization Algorithm

Inspired by Karen A. Kopecky 2006 Lecture Note

1. Choose a relateive error tolerance level `tol`;
2. Discretize the state space by constructing a grid for productivity

$$p = \exp\{logp\} \text{ where } logp = \{logp_1, logp_2, \ldots, logp_n\}$$

   given by the Tauchen-Hussey (1991) method. The n is chosen at 100;
3. Start with an initial guess of the value function $V^{(0)}(p)$ is a vector of length n, i.e., $V^{(0)}(p) = \{V_i^{(0)}\}_{i=1}^n$, where $V_i^{(0)} = V^{(0)}(p_i)$. V here represents U, W, J. The initial guess is ones.

Question a II

Discretization Algorithm

4. Update the value function using eqautions 1 to 10, specifically
   - 4.1 Fix the current productivity level at one of the grid points, p_i from $i = 1$
   - 4.2 For each possible choice of productivity next period, calculate optimal control in the following order:

$$q(\theta_{p_i}) = \frac{c}{\delta \sum_{j=1}^{n} p_{i,j} J^{(0)}(p_j)}$$

$$f(\theta_{p_i}) = \mu^{\frac{1}{\eta}} q^{\frac{\eta-1}{\eta}}$$

$$\theta_{p_i} = (\frac{q(\theta_{p_i})}{\mu})^{-\frac{1}{\eta}}$$

$$w_{p_i} = \beta p_i + (1-\beta) z + \beta c \theta_{p_i}$$

   - 4.3 and update the value function system with

Question a III

Discretization Algorithm

$$U_{p_i}^{(1)} = z + \delta\{f(\theta_{p_i})\sum_{j=1}^{n} p_{i,j} W^{(0)}(p_j) + (1 - f(\theta_{p_i}))\sum_{j=1}^{n} p_{i,j} U^{(0)}(p_j)\}$$

$$W_{p_i}^{(1)} = w_{p_i} + \delta\{(1-s)\sum_{j=1}^{n} p_{i,j} W^{(0)}(p_j) + s\sum_{j=1}^{n} p_{i,j} U^{(0)}(p_j)\}$$

$$J_{p_i}^{(1)} = p_i - w_{p_i} + \delta(1-s)\sum_{j=1}^{n} p_{i,j} J^{(0)}(p_j)$$

4.4 Choose a new grid point for productivity, go through 4.1 to 4.3. Once we have done the update for all productivity grid, we have new system of value function $V_p^{(1)}$

4.5 Compute distance between the two systems of value functions following the sup norm

$$d = \max_{i \in \{1,\dots,n\}} |V_i^{(0)} - V_i^{(1)}|$$

Question a IV

Discretization Algorithm

4.6 If distance is within the error tolerance level, $d \leq tol * ||V_1^{(1)}||$, the functions have converged and go to step 5, or else go back to step 4.

5. Calculate the optimal control for each productivity level:

$$q(\theta^*_{p_i}) = \frac{c}{\delta \sum_{j=1}^{n} p_{i,j} J^*(p_j)}$$

$$f(\theta^*_{p_i}) = \mu^{\frac{1}{\eta}} q^{\frac{\eta-1}{\eta}}$$

$$\theta^*_{p_i} = (\frac{q(\theta^*_{p_i})}{\mu})^{-\frac{1}{\eta}}$$

$$w^*_{p_i} = \beta p_i + (1-\beta)z + \beta c \theta^*_{p_i}$$

$$u^*_p = \frac{\delta}{\delta + f(\theta^*_p)}$$

where J^* is the converged value function.

Tauchen-Hussey 1991

Use `discretizeAR1_Tauchen` function from the Matlab Toolbox of Kirkby (2023) .

Question a

Parametric Approximation

0. Choose Hermite interpolation polynomials to approximate $\hat{V}(p; \mathbf{coefs})$ in the form of $f(x) = a(x - x_1)^3 + b(x - x_1)^2 + c(x - x_1) + d$ with Matlab code `pchip`. Report initial paramters with `pp.coefs` for each value function, save as `old`
1. Maximize control and calculate value function at each productivity level as done in Discretization method
2. Fit for new value function system and report parameters and save as `new`
3. If $||\hat{V}(p; \mathbf{coef\_sold}) - \hat{V}(p; \mathbf{coefs\_new})|| < tol$, stop; else go to step 1.

Two Results

Question b

To use the polynomial interpolation function from class directly

log(p)	w	u	f
0.400			
0.700			
1.000			
1.300			
1.600			

Table 2: Wage, Unemployment rate and Job Finding Rate

Two Results

Question c

	u	f	p
Data Std.	0.190	0.118	0.020
Approximation Model Std.			
Discretization Model Std.			

Table 3: Model fit on Unemployment, Job finding rate and productivity

Appendix A

Optimal wage

$$
\begin{aligned}
W_p - U_p &= \beta(W_p - U_p + J_p) \\
\Leftrightarrow w_p - z + \delta(1 - s - f(\theta_p))(\mathbb{E}_p W_{p'} - \mathbb{E}_p U_{p'}) &= \\
\beta(p - z + \delta(1 - s - f(\theta_p))(\mathbb{E}_p W_{p'} - \mathbb{E}_p U_{p'}) &+ \delta(1-s)\mathbb{E}_p J_{p'}) \\
\Leftrightarrow w_p = \beta p + (1-\beta)z + (\beta - 1)\delta(1 - s - f(\theta_p))(\mathbb{E}_p W_{p'} - \mathbb{E}_p U_{p'}) \\
+ \frac{\beta c(1-s)}{q(\theta_p)} \\
\Leftrightarrow w_p = \beta p + (1-\beta)z - \frac{\beta c \delta(1 - s - f(\theta_p))}{q(\theta_p)} + \frac{\beta c(1-s)}{q(\theta_p)} \\
\Leftrightarrow w_p = \beta p + (1-\beta)z + \beta c \theta_p
\end{aligned}
$$

where we use the fact that $\mathbb{E}_p W_{p'} - \mathbb{E}_p U_{p'} = \frac{\beta}{1-\beta}\mathbb{E}_p J_{p'}$ and $f(\theta_p)/q(\theta_p) = \theta_p$

Reference I

Kirkby, R. (2023), 'Value function iteration (vfi) toolkit for matlab', `https://github.com/vfitoolkit/VFIToolkit-matlab`. Github.

Shimer, R. (2005), 'The cyclical behavior of equilibrium unemployment and vacancies', *American Economic Review* **95**(1), 25–49.
URL: *https://www.aeaweb.org/articles?id=10.1257/0002828053828572*