

# Lecture 7

Numerical Analysis

6/2

# Goals

- We have two goals.
- First, given a set of parameters, we want to solve our model (often called the DMP model) using Octave/Matlab and Python.
- Second, we want to find a set of realistic parameters for the model.
- In your own research, you should start with the second task, followed by the first.

# Model Equations

- Remember that the steady-state equilibrium is given by the set of  $(\theta, w, u)$  satisfying

$$w = p - (r + \lambda) \frac{pc}{q(\theta)}$$

$$w = \beta p + (1 - \beta)z + \beta pc\theta$$

$$u = \frac{\lambda}{\lambda + \theta q(\theta)}$$

- Pissarides textbook assumes that the vacancy cost is proportional to labor productivity (for extensions in later Chapters).

# Note on Vacancy Cost

- In many existing models, the vacancy cost is simply  $c$  rather than  $pc$ . Then, the system becomes

$$w = p - \frac{(r + \lambda)c}{q(\theta)}$$

$$w = \beta p + (1 - \beta)z + \beta c\theta$$

$$u = \frac{\lambda}{\lambda + \theta q(\theta)}$$

- Today, we shall employ this version of the DMP model.

# Specifications

- The first step is to specify all the functional forms.
- Let us suppose that the matching function is Cobb-Douglas, so that  $q(\theta) = A\theta^{-\alpha}$ .
- We then obtain

$$w = p - \frac{(r + \lambda)c}{A} \theta^\alpha$$
$$w = \beta p + (1 - \beta)z + \beta c \theta$$
$$u = \frac{\lambda}{\lambda + A\theta^{1-\alpha}}$$

# Parameter Values

- Finally, we need to specify the set of parameters for the analysis. But, how?
- To make the model's prediction useful, we need to find a realistic set of parameters.
- For now, let us pretend that we know a reasonable set of parameters and jump into Maxima (and Python).

# Shimer (2005)

$r$	$A$	$\alpha$	$\lambda$	$c$	$\beta$	$z$	$p$
0.012	1.355	0.72	0.1	0.213	0.72	0.4	1

- Robert Shimer, “The Cyclical Behavior of Equilibrium Unemployment and Vacancies,” *American Economic Review*, 2005.
- This is an influential article on quantitative studies of the DMP model.
- One period = one quarter.
- We shall borrow parameters from his work.

# Solving the Model

In Maxima



# Parameter Values

## Model

- $r = 0.012$
- $A = 1.355$
- $\alpha = 0.72$
- $\lambda = 0.1$
- $c = 0.213$
- $\beta = 0.72$
- $z = 0.4$
- $p = 1$
- Colon ":" means assignment "=".

## Maxima

```
/* Parameter values */  
r:0.012;  
A:1.355;  
alpha:0.72;  
lambda:0.1;  
c:0.213;  
beta:0.72;  
z:0.4;  
p:1;
```

# Job Creation Condition

- The equation is

$$w = p - \frac{(r + \lambda)c}{A} \theta^\alpha$$

- It is convenient to define a function of  $\theta$ .
- “:=” defines a function.

```
/* Job-creation condition */  
JC(theta) := p - (r + lambda)*(c/A)*theta^alpha;
```

# Wage Equation

- The equation is

$$w = \beta p + (1 - \beta)z + \beta c\theta$$

- It is convenient to define a function of  $\theta$ .

```
/* Wage equation */  
Wage(theta) := beta*p + (1 - beta)*z + beta*c*theta;
```

# Beveridge Curve

- The equation is

$$u = \frac{\lambda}{\lambda + A\theta^{1-\alpha}}$$

- It is convenient to define a function of  $\theta$ .

```
/* Unemployment rate */  
Unemp(theta) := lambda/(lambda + A*(theta^(1 - alpha)));
```

# Job-Finding Rate

- The expression is

$$f_t = A\theta_t^{1-\alpha}$$

- We shall compute the equilibrium JFR in the model.

```
/* Job-Finding Rate */  
JFR(theta) := A*(theta^(1-alpha));
```

# Diagram

- As in the textbook, let us draw a diagram with two curves, the Job-creation condition and the wage equation.

```
/* Diagram */  
plot2d([JC(theta),Wage(theta)],[theta,0.5,1.5]);
```

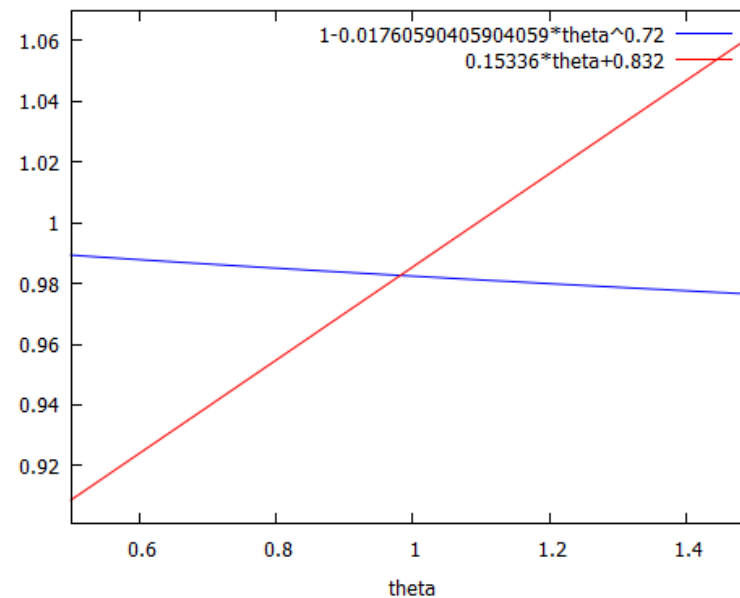
# Solving the System of Equations

- Unfortunately, the `solve()` command cannot solve our model.
  - It only solves equations that reduce to polynomials.
- We use the `mnewton()` command:

```
/* Solving the model */  
load("mnewton")$  
mnewton([w=JC(theta),w=Wage(theta)], [w,theta], [1,1])[1][1];  
mnewton([w=JC(theta),w=Wage(theta)], [w,theta], [1,1])[1][2];  
eqtheta: rhs(mnewton([w=JC(theta),w=Wage(theta)], [w,theta], [1,1])[1][2]);
```

- We need the initial condition for computation.

# Maxima output



$w=0.9826210585631064$   
 $\theta=0.9821404444647004$   
 $u=0.06905218641629844$



# Consistent with Shimer (2005)

## Maxima Output

- JFR in the model is 1.348180067711758.
- The maxima code is available from NUCT.

## Shimer (2005)

month, so the flow arrival rate of job offers  $\mu\theta^{1-\alpha}$  should average approximately 1.35 on a quarterly basis. I do not have a long time series

# Further Readings

- Robert Shimer, “The Cyclical Behavior of Equilibrium Unemployment and Vacancies,” *American Economic Review*, 2005.
- For more information about Maxima, visit <http://maxima.sourceforge.net/documentation.html>
  - Maxima can be your good friend throughout your graduate study.

# Solving the Model

In Python

# Python Code

- We need to start with taking care of libraries.

```
%matplotlib inline  
import numpy as np  
import matplotlib.pyplot as plt  
from scipy.optimize import root
```

# Python Code

- We then define parameter values.
- To input Greek letters, ¥alpha then press Tab.
- You must not use lambda as a parameter name because it is kept as a function in Python.

```
r = 0.012
A = 1.355
 $\alpha$  = 0.72
 $\lambda$  = 0.1
c = 0.213
 $\beta$  = 0.72
z = 0.4
p = 1
```

# Python Code

- We then define functions:

```
# Job-creation condition
def JC( $\theta$ ):
    return  $p - (r + \lambda) * (c/A) * (\theta^{**\alpha})$ 
# Wage equation
def Wage( $\theta$ ):
    return  $\beta * p + (1 - \beta) * z + \beta * c * \theta$ 
# Unemployment rate
def Unemp( $\theta$ ):
    return  $\lambda / (\lambda + A * (\theta^{**}(1 - \alpha)))$ 
# Job-Finding rate
def JFR( $\theta$ ):
    return  $A * (\theta^{**}(1 - \alpha))$ 
```

# Diagram

```
# Drawing a Diagram

plt.rcParams['figure.figsize'] = (10,7)

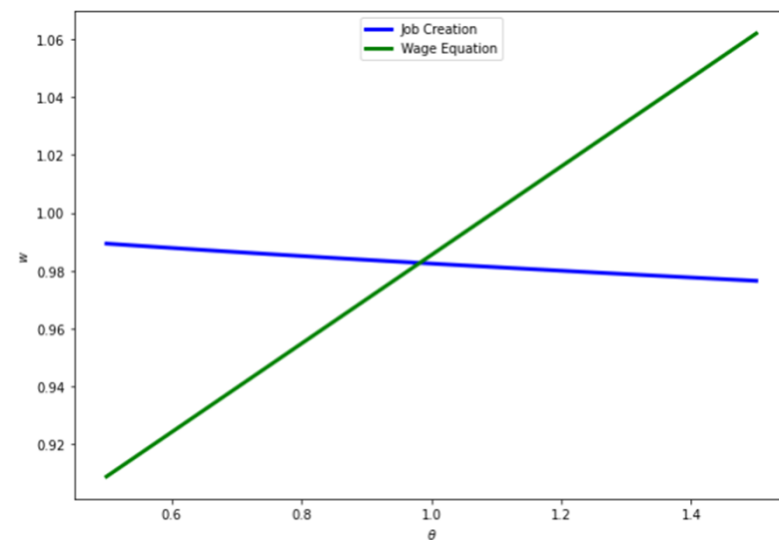
theta_grid = np.linspace(0.5, 1.5, 100)
fig, ax = plt.subplots()

JC_curve = JC(theta_grid)
Wage_curve = Wage(theta_grid)

ax.plot(theta_grid, JC_curve, 'b-', lw=3, label='Job Creation')
ax.plot(theta_grid, Wage_curve, 'g-', lw=3, label='Wage Equation')

ax.set_xlabel('$\theta$')
ax.set_ylabel('$w$')
ax.legend(loc='upper center')

plt.show()
```



# Solving the Model

- We then solve the system of nonlinear equations.
- The initial condition for numerical computation is set at (1,1).

```
def func(x):  
    f = [JC(x[0]) - x[1],  
         Wage(x[0]) - x[1]]  
    return f  
sol = root(func, [1,1])  
sol.x
```



# Output

- We finally print the output.

```
eq_theta = sol.x[0]
eq_wage = sol.x[1]
eq_u = Unemp(eq_theta)
eq_JFR = JFR(eq_theta)
```

```
print(f'Equilibrium tightness is
{eq_theta:.4f}')
print(f'Equilibrium wage rate is
{eq_wage:.4f}')
print(f'Equilibrium unemployment
rate is {eq_u:.4f}')
print(f'Equilibrium job-finding rate
is {eq_JFR:.4f}')
```

# Python Output

Equilibrium tightness is 0.9821

Equilibrium wage rate is 0.9826

Equilibrium unemployment rate is 0.0691

Equilibrium job-finding rate is 1.3482

# Calibration

# Calibration

- Now we have solved the steady state of the DMP model, given a set of parameters.
- How do we find the **right parameter values**?
- Modern macroeconomic research typically employs the method of **calibration**:
  - Borrow parameter values from empirical facts:
    - From the data.
    - From existing research.
  - Find parameter values using the model equations.

# The Set of Parameters

- $r$  : discount rate
- $A$  : matching efficiency
- $\alpha$  : matching elasticity
- $\lambda$  : separation rate
- $c$  : vacancy cost
- $\beta$  : bargaining power
- $z$  : UI benefit
- $p$  : labor productivity
- We have 8 parameters to determine.
- The basic idea is to introduce 8 equations to determine the parameter values.
- The facts for determining the parameter values are called **targets**.

# Target Economy

- In principle, you may choose any country for your quantitative analysis.
- The advantage of studying the US economy is that you can find many articles.
  - There are many learning opportunities.
  - You can borrow parameter values.
  - You can compare your results with others.
- Many non-US economists choose their own countries in addition to US.
  - I study Japan and US for my own research.

# Model Period

- Because our model has discount rate ( $r$ ), we need to choose the model period.
- Typical choices are:
  - Quarter
  - Month
- Today we choose one period to be a quarter.

# Calibration: $r$

- Collect the long run time series data on the real interest rate.
  - There is no such things as “the” real interest rate.
  - There are hopelessly many interest rates in the real world.
  - Many people, however, adopt  $r = 0.01$  by arguing that the annual real interest rate is 4~5%
- If the annual real interest rate is 4%, then the quarterly real interest rate is approximately 1%.



# Calibration: $r$

- Suppose that the quarterly interest rate is  $r_q$ . Then the annual interest rate  $r_a$  is given by

$$1 + r_a = (1 + r_q)^4$$

- Consider the nonlinear function  $f(x) = (1 + x)^4$ . Linearize it at  $x = 0$  to obtain

$$\begin{aligned} f(x) &\approx f(0) + f'(0)(x - 0) \\ &= (1 + 0)^4 + 4(1 + 0)^3(x - 0) \\ &= 1 + 4x \end{aligned}$$

- Thus, for  $r_q$  close to 0, we can use

$$r_a \approx 4r_q$$

# Calibration: $r$

- Suppose the target annual real interest rate is 5%. Then,

$$1.05 = (1 + r)^4$$

- Solving it to obtain  
 $r = 0.01227$

- The annual discount factor is

$$\frac{1}{1 + 0.05} = 0.95238$$

**Shimer (2005)**

I normalize a time period to be one quarter, and therefore set the discount rate to  $r = 0.012$ , equivalent to an annual discount factor of 0.953.

# Calibration: $\lambda$

- To pin down the separation rate, we need worker flow data.
  - US: Current Population Survey (CPS)
  - Japan: Labor Force Survey
- Construction of worker flow data requires too much work and time.
- We typically rely on someone else's work, such as <https://www.federalreserve.gov/pubs/feds/2004/200434/200434abs.html>

# Calibration: $\lambda$

- Shimer (2005) adopts the quarterly separation rate in the US to be 0.1, so

$$\lambda = 0.1$$

- Lin and Miyamoto (2012) find that the monthly separation rate in Japan is 0.0048, implying that the quarterly separation rate in Japan is

$$\lambda = 0.0048 \times 3 = 0.0144$$

# Calibration: $\alpha$

- Remember:

$$M_t = AU_t^\alpha V_t^{1-\alpha}$$

where  $A > 0$  and  $0 < \alpha < 1$ .

- $\alpha$  is the elasticity of matching with respect to unemployment.
- If you take log of the equation, we obtain
$$\log M_t = \log A + \alpha \log U_t + (1 - \alpha) \log V_t$$
- We can estimate  $\alpha$  by regression analysis.

# Calibration: $\alpha$

- There are many estimates.
- Shimer's (2005) estimate is 0.72.
- An often-cited summary of empirical results is:  
Petrongolo and Pissarides, "Looking into the Black Box: A Survey of the Matching Function" *Journal of Economic Literature*, 2001.
  - They conclude that the plausible range of  $\alpha$  is 0.5 to 0.7.
- Lin and Miyamoto (2014) estimate that  $\alpha = 0.6$  in Japan.

# Calibration: $\beta$

- This is the worker's bargaining power in Nash wage bargaining.
  - There is no direct estimate on this value.
- In the literature, there are 2 ways:
  - Set  $\beta = 0.5$  because we have no information.
  - Set  $\beta = \alpha$  to ensure the Hosios condition.
    - This means job creation is efficient.
    - Shimer (2005) adopts this method.

# Calibration: $p$

- We normalize the steady-state level to be  $p = 1$ .
- If we are interested in the effect of an increase in  $p$  on  $u$ , then we can compare the steady states with  $p = 1$  and  $p = 1.01$  (one percentage increase).



# Calibration: Targets

- Sometimes we need to use model equations to find model parameters
- A calibration **target** is:
  - An endogenous variable (not a parameter) of the model,
  - corresponding long-run data is available,
  - Not the main variables of interest.

# Calibration: $A$

- Job-finding rate in the model =  $A\theta^{1-\alpha}$ .
- Monthly job-finding rate = 0.142
- Average UV ratio in Japan = 0.78
- Matching elasticity =  $\alpha = 0.5$
- From these values,  
$$A(0.78)^{1-0.5} = 3 \times 0.142,$$

- Thus,

$$A = \frac{3 \times 0.142}{(0.78)^{0.5}} = 0.482$$

# Calibration: $z$ and $c$

- Steady-state equations:

$$w = p - \frac{(r + \lambda)c}{A} \theta^\alpha$$
$$w = \beta p + (1 - \beta)z + \beta c\theta$$

- Given a set of parameters, we can find the solution  $(w, \theta)$ .
- Basic idea: we set targets for  $w$  and  $\theta$  and use these equations to find  $(z, c)$ .
  - Reverse engineering!

# Calibration: $z$ and $c$

- One target is  $\theta = 0.78$  from the data.
- Another target is the **unemployment benefit replacement rate**.
- In Japan, replacement rate is 0.6.
- Thus,  $\frac{z}{w} = 0.6$

	<i>Direct Rigidities</i>		<i>Treatment of the Unemployed</i>		
	<i>1 Employment Protection</i>	<i>2 Labor Standards</i>	<i>3 Benefit Replacement Rate (%)</i>	<i>4 Benefit Duration (years)</i>	<i>5 Active Labor Market Policies</i>
Austria	16	5	50	2	8.3
Belgium	17	4	60	4	14.6
Denmark	5	2	90	2.5	10.3
Finland	10	5	63	2	16.4
France	14	6	57	3	8.8
Germany (W)	15	6	63	4	25.7
Ireland	12	4	37	4	9.1
Italy	20	7	20	0.5	10.3
Netherlands	9	5	70	2	6.9
Norway	11	5	65	1.5	14.7
Portugal	18	4	65	0.8	18.8
Spain	19	7	70	3.5	4.7
Sweden	13	7	80	1.2	59.3
Switzerland	6	3	70	1	8.2
U.K.	7	0	38	4	6.4
Canada	3	2	59	1	5.9
U.S.	1	0	50	0.5	3.0
Japan	8	1	60	0.5	4.3
Australia	4	3	36	4	3.2
New Zealand	2	3	30	4	6.8

Nickell, "Unemployment and Labor Market Rigidity: Europe versus North America," *Journal of Economic Perspectives*, 1997.

# Calibration: $z$ and $c$

- Original system  $(w, \theta)$ :

$$w = p - \frac{(r + \lambda)c}{A} \theta^\alpha$$
$$w = \beta p + (1 - \beta)z + \beta c \theta$$

- Calibration equations  $(z, c)$ :

$$\frac{z}{0.6} = p - \frac{(r + \lambda)c}{A} 0.78^\alpha$$
$$\frac{z}{0.6} = \beta p + (1 - \beta)z + 0.78\beta c$$

- $(b, c)$  is chosen so that “the solution = targets”.

# Japanese Labor Market

- Purely exogenous parameters
  - $r = 0.01$
  - $\lambda = 0.0048 \times 3$
  - $\alpha = 0.6$
  - $\beta = 0.5$
  - $p = 1$
- Endogenous (!) parameters
  - $A = 0.482$
  - $z = ?$
  - $c = ?$

# Further Readings

- Robert Shimer, “The Cyclical Behavior of Equilibrium Unemployment and Vacancies,” *American Economic Review*, 2005.
- Lin and Miyamoto, “Gross Worker Flows and Unemployment Dynamics in Japan,” *Journal of the Japanese and International Economies*, 2012.
- Lin and Miyamoto, “An Estimated Search and Matching Model of the Japanese Labor Market,” *Journal of the Japanese and International Economies*, 2014.