Numerical Methods for New Keynesian ZLB Models: Part I

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What is covered

- Equilibrium with a simple Taylor rule
- Tauchen's (1986) method for approximating AR(1) process
- Equilibrium under the optimal discretionary policy (Adam and Billi, 2007)

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Two-state shock process

• Exogenous shocks take only $N_s=2$ values, $s_t \in \{s_H, s_L\}$. The stochastic process follows a Markov chain with the transition matrix:

$$\left[\begin{array}{cc} 1-p_H & p_H \\ 1-p_L & p_L \end{array}\right].$$

ullet p_H is the frequency of crisis and p_L is the duration of crisis.

Equilibrium with Taylor rule

• Equilibrium conditions are

$$y_{t} = E_{t}y_{t+1} - (i_{t} - E_{t}\pi_{t+1} - s_{t}),$$

$$\pi_{t} = \kappa y_{t} + \beta E_{t}\pi_{t+1},$$

$$i_{t}^{*} = r^{*} + \phi_{\pi} E_{t}\pi_{t+1},$$

and the zero lower bound

$$i_t = \max\left\{0, i_t^*\right\}.$$

Solving the model: Two routes

• The solution has a form of (we omit time subscripts for the policy function)

$$y = y(s), \quad \pi = \pi(s), \quad i = i(s).$$

- We are solving the same model with two different methods.
 - Analytical solution.
 - Numerical solution, using the policy function iteration (time iteration).

Analytical solution

• We know that the functions have only two values, i.e.,

$$y = \begin{cases} y_H, & \pi = \begin{cases} \pi_H, \\ \pi_L, \end{cases} \quad i = \begin{cases} i_H, \\ i_L. \end{cases}$$

Analytical solution, cont'd

• We assume that $i_H > 0$ and $i_L = 0$. Then we have

$$\begin{split} y_H &= (1-p_H)y_H + p_H y_L - (i_H - [(1-p_H)\pi_H + p_H \pi_L] - s_H) \,, \\ \pi_H &= \kappa y_H + \beta \left[(1-p_H)\pi_H + p_H \pi_L \right] \,, \\ i_H &= r^* + \phi_\pi \left[(1-p_H)\pi_H + p_H \pi_L \right] \,, \\ y_L &= (1-p_L)y_H + p_L y_L - (0 - [(1-p_L)\pi_H + p_L \pi_L] - s_L) \,, \\ \pi_L &= \kappa y_L + \beta \left[(1-p_L)\pi_H + p_L \pi_L \right] \,, \\ i_L &= 0 . \end{split}$$

• There are 6 equations and 6 unknowns, so we can solve for the unknowns.

Numerical solution: Policy function iteration

- We can solve the same model with the policy function iteration method.
- The method takes four steps:
 - Initial guess for the policy function
 - 2 Solving for endogenous variables at each grid
 - Updating the policy function
 - Repeat 2-3 until convergence

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Policy function iteration: Initial guess

A guess of the policy functions

$$y = y^{(0)}(s), \quad \pi = \pi^{(0)}(s), \quad i = i^{(0)}(s).$$

• Consider the general case of $N_s \geq 2$. We know the values of the functions only at each *grid point*, i.e.,

$$\begin{split} y^{(0)}(s) &= [y_1, y_2, ..., y_{N_s}]', \\ \pi^{(0)}(s) &= [\pi_1, \pi_2, ..., \pi_{N_s}]', \\ i^{(0)}(s) &= [i_1, i_2, ..., i_{N_s}]'. \end{split}$$

Policy function iteration: Solving at each grid

• At each grid point $k = 1, ..., N_s$, we solve

$$y_k = y^e - (i_k - \pi^e - s_k),$$

 $\pi_k = \kappa y_k + \beta \pi^e,$
 $i_k = \max\{0, r^* + \phi_\pi \pi^e\},$

for (y_k, π_k, i_k) , where

$$y^{e} = \sum_{l=1}^{N_{s}} p(k, l) y^{(0)}(s_{l}),$$
$$\pi^{e} = \sum_{l=1}^{N_{s}} p(k, l) \pi^{(0)}(s_{l}).$$

and p(k, l) is the (k, l) element of the transition matrix.

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Policy function iteration: Updating and convergence

Once this is done for all the grid points, we update

$$y^{(1)}(s) = [y_1, y_2, ..., y_N]',$$

$$\pi^{(1)}(s) = [\pi_1, \pi_2, ..., \pi_N]',$$

$$i^{(1)}(s) = [i_1, i_2, ..., i_N]'.$$

• We repeat the procedure until the policy functions converge, i.e., $\left\|x^{(j)}(s)-x^{(j-1)}(s)\right\|<\epsilon$ for $x\in\{y,\pi,i\}.$

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Tauchen's method

- Tauchen (1986) developed a method for approximating AR(1) stochastic process by using Markov chain.
- We have the following AR(1) stochastic process

$$x' = c + \rho x + \varepsilon', \varepsilon' \sim N(0, \sigma_\varepsilon^2).$$

• We want to approximate the stochastic process by a Markov chain $x_k \in \{x_1, x_2, ..., x_N\}$.

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Tauchen's method: Grid points

• We set the grid points for x:

$$x_k \in \mathcal{I} = \{x_1, x_2, ..., x_N\},\$$

where k is an index for the set of grid points \mathcal{I} .

• For example, we set $x_1=\frac{-m\sigma_\varepsilon}{\sqrt{1-\rho^2}},\ x_N=\frac{m\sigma_\varepsilon}{\sqrt{1-\rho^2}}$ and $x_k=x_{k-1}+w$ for k=2,...,N-1, where $w=\frac{x_N-x_1}{N-1}$.

Tauchen's method: Transition matrix

- Given the grid points for x. What is the probability of moving from one point x_k to another x_l ?
- We know

$$\varepsilon' = x' - c - \rho x_k \sim N(0, \sigma_{\varepsilon}^2).$$

Tauchen's method: Transition matrix, cont'd

• Then, the probability of $x' \in \left[x_l - \frac{w}{2}, x_l + \frac{w}{2}\right]$ can be used as approximation. That is,

$$p_{kl} = \Phi\left(x_l + \frac{w}{2} - c - \rho x_k\right) - \Phi\left(x_l - \frac{w}{2} - c - \rho x_k\right),$$

where $\Phi(\cdot)$ is the cdf of $N(0, \sigma_{\varepsilon}^2)$. Be careful at the boundary points.

• Once this is done for all k, l, we have the transition matrix

$$P = \left[\begin{array}{cccc} p_{11} & \cdots & \cdots & p_{1N} \\ p_{21} & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ p_{N1} & \cdots & \cdots & p_{NN} \end{array} \right].$$

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Optimal Discretionary Policy

• The policymaker chooses $\{\pi_t, y_t, i_t\}$ so as to maximize

$$V_0 \equiv -E_0 \sum_{t=0}^{\infty} \beta^t \left(\pi_t^2 + \lambda y_t^2 \right)$$

subject to

$$y_{t} = E_{t}y_{t+1} - (i_{t} - E_{t}\pi_{t+1}) + g_{t},$$

$$\pi_{t} = \kappa y_{t} + \beta E_{t}\pi_{t+1} + u_{t},$$

$$i_{t} \ge 0,$$

taking $E_t y_{t+1}$ and $E_t \pi_{t+1}$ as given.

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Shock processes

Exogenous shocks are given by

$$g_t = (1 - \rho_g)g + \rho_g g_{t-1} + \varepsilon_{g,t},$$

$$u_t = \rho_u u_{t-1} + \varepsilon_{u,t},$$

where $\varepsilon_{g,t} \sim N(0,\sigma_g^2)$ and $\varepsilon_{u,t} \sim N(0,\sigma_u^2)$.

 $\bullet \ \ {\rm Note \ that} \ g=r^*.$

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Optimal Discretionary Policy: Lagrangean

- We know that Markov-perfect equilibrium has only natural state variables.
- Lagrangean is

$$\mathcal{L} \equiv E_0 \sum_{t} \beta^t \left(\pi_t^2 + \lambda y_t^2 \right) + 2\phi_{PC,t} \left(-\pi_t + \kappa y_t + \beta E_t \pi_{t+1} + u_t \right) + 2\phi_{EE,t} \left(-y_t - i_t + E_t y_{t+1} + E_t \pi_{t+1} + g_t \right) + 2\phi_{ZLB,t} i_t.$$

• First-order necessary conditions are

$$\begin{split} \partial \pi_t : & \pi_t - \phi_{PC,t} = 0, \\ \partial y_t : & \lambda y_t + \kappa \phi_{PC,t} - \phi_{EE,t} = 0, \\ \partial i_t : & - \phi_{EE,t} + \phi_{ZLB,t} = 0. \end{split}$$

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Optimal Discretionary Policy: Complementary slackness

Complementary slackness condition:

$$\phi_{ZLB,t} > 0 \perp i_t > 0.$$

• When $i_t > 0$, $\phi_{ZLB,t} = 0$. Equilibrium conitions are

$$\begin{split} i_t &= -y_t + E_t y_{t+1} + E_t \pi_{t+1} + g_t, \\ \pi_t &= \kappa y_t + \beta E_t \pi_{t+1} + u_t, \\ 0 &= \lambda y_t + \kappa \pi_t. \end{split}$$

• When $i_t = 0$, $\phi_{ZLB,t} > 0$. Equilibrium conitions are

$$0 = -y_t + E_t y_{t+1} + E_t \pi_{t+1} + g_t,$$

$$\pi_t = \kappa y_t + \beta E_t \pi_{t+1} + u_t,$$

$$\phi_{ZLB,t} = \lambda y_t + \kappa \pi_t.$$

Solving the model

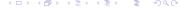
The solution has a form of

$$y = y(g, u), \quad \pi = \pi(g, u), \quad i = i(g, u).$$

 Now consider the case in which there are only two-state g shocks. We know that the functions have only two values, i.e.,

$$y = \begin{cases} y_H, & \pi = \begin{cases} \pi_H, \\ \pi_L, \end{cases} \quad i = \begin{cases} i_H, \\ i_L. \end{cases}$$

Again, we will look at analytical solution and numerical solution.



Analytical solution

• We assume that $i_H > 0$ and $i_L = 0$. Then we have

$$\begin{split} y_{H} &= (1 - p_{H})y_{H} + p_{H}y_{L} - (i_{H} - [(1 - p_{H})\pi_{H} + p_{H}\pi_{L}]) + g_{H}, \\ \pi_{H} &= \kappa y_{H} + \beta \left[(1 - p_{H})\pi_{H} + p_{H}\pi_{L} \right], \\ 0 &= \lambda y_{H} + \kappa \pi_{H}, \\ y_{L} &= (1 - p_{L})y_{H} + p_{L}y_{L} - (0 - [(1 - p_{L})\pi_{H} + p_{L}\pi_{L}]) + g_{L}, \\ \pi_{L} &= \kappa y_{L} + \beta \left[(1 - p_{L})\pi_{H} + p_{L}\pi_{L} \right], \\ \phi_{L} &= \lambda y_{L} + \kappa \pi_{L}. \end{split}$$

• There are 6 equations and 6 unknowns, so we can solve for the unknowns.

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Joint shock process

 Let's get back to the general case. The shock processes are appoximated by Markov chains. That is,

$$g_m \in \{g_1, g_2, ..., g_{N_g}\},\ u_n \in \{u_1, u_2, ..., u_{N_u}\},\$$

and

$$P^g = \left[\begin{array}{cccc} p_{11}^g & \cdots & \cdots & p_{1N_g}^g \\ p_{21}^g & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ p_{N_g1}^g & \cdots & \cdots & p_{N_gN_g}^g \end{array} \right], \quad P^u = \left[\begin{array}{cccc} p_{11}^u & \cdots & \cdots & p_{1N_g}^u \\ p_{21}^u & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ p_{N_g1}^u & \cdots & \cdots & p_{N_gN_g}^u \end{array} \right].$$

Joint shock process, cont'd

- 2D grid coordinate with q and u represents the joint shock process.
 - For example, when $N_a = N_u = 2$

$$s_1 = (g_1, u_1),$$

 $s_2 = (g_1, u_2),$
 $s_3 = (g_2, u_1),$
 $s_4 = (g_2, u_2).$

- Note that each index points to a pair of shocks, $s_k = (g_{m(k)}, u_{n(k)})$.
- A kronecker product of the transition matrices $P^s = P^g \otimes P^u$ is the transition matrix of the joint shock process.

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Policy function iteration: Initial guess

A guess of the policy functions

$$y = y^{(0)}(s), \quad \pi = \pi^{(0)}(s), \quad i = i^{(0)}(s).$$

• We know the values of the functions only at each grid point, e.g.,

$$y^{(0)}(s) = [y_1, y_2, ..., y_N]',$$

$$\pi^{(0)}(s) = [\pi_1, \pi_2, ..., \pi_N]',$$

$$i^{(0)}(s) = [i_1, i_2, ..., i_N]'.$$

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Policy function iteration: Solving at each grid

• At each grid point k = 1, ..., N, we solve

$$i_k = -y_k + y^e + \pi^e + g_{m(k)},$$

$$\pi_k = \kappa y_k + \beta \pi^e + u_{n(k)},$$

$$0 = \lambda y_k + \kappa \pi_k,$$

for (y_k, π_k, i_k) , where

$$y^{e} = \sum_{l=1}^{N_{s}} P^{s}(k, l) y^{(0)}(s_{l}),$$
$$\pi^{e} = \sum_{l=1}^{N_{s}} P^{s}(k, l) \pi^{(0)}(s_{l}).$$

Policy function iteration: Solving at each grid, cont'd

• Check $i_k \geq 0$. If not, we solve instead

$$0 = -y_k + y^e + \pi^e + g_{m(k)},$$

$$\pi_k = \kappa y_k + \beta \pi^e + u_{n(k)},$$

for (y_k, π_k) , and set $i_k = 0$.

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Policy function iteration: Updating and convergence

Once this is done for all the grid points, we update

$$y^{(1)}(s) = [y_1, y_2, ..., y_N]',$$

$$\pi^{(1)}(s) = [\pi_1, \pi_2, ..., \pi_N]',$$

$$i^{(1)}(s) = [i_1, i_2, ..., i_N]'.$$

• We repeat the procedure until the policy functions converge, i.e., $\left\|x^{(j)}(s)-x^{(j-1)}(s)\right\|<\epsilon$ for $x\in\{y,\pi,i\}.$

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Figure 4 in Adam and Billi (2007)

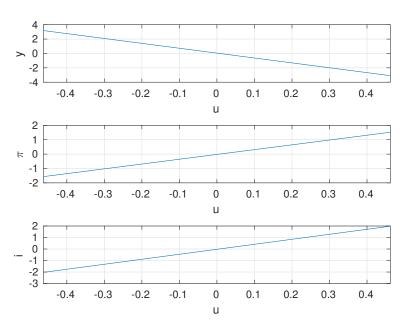


Figure 5 in Adam and Billi (2007)

