

Guvenen (2009)

1 Variables

1. The set of 2 endogenous state variables has the following elements:

$$K_t, B_t^n$$

CODE: Kx, Bnx (*code variables: states end with 'x', controls with 'y'*)

2. The set of 1 exogenous state variables has the following elements:

$$Z_t$$

CODE: Zx

3. The set of 17 control variables has the following elements:

$$\text{Core: } c_{h,t}, c_{n,t}, I_t, B_{t+1}^n, \lambda_{h,t}, \lambda_{n,t}, P_t^s, P_t^f$$

CODE: chy, cny, Iy, Bny, lambdahy, lambdany, Psy, Pfy

$$\text{Additional: } q_t, W_t, \frac{\Lambda_{h,t+1}}{\Lambda_{h,t}}, \frac{\Lambda_{n,t+1}}{\Lambda_{n,t}}, U_{h,t}, U_{n,t}, D_t, b_{h,t+1}, b_{n,t+1}$$

CODE: qy, Wy, Lambdahy, Lambdany, Uhy, Uny, Dy, bhy, bny

2 Equations

legend: **states in green** (given), **controls in orange** (given by policy guess from current states), **next period's exogenous states in magenta** (to be integrated over), and **next period's controls in blue** (given by policy guess at next period's states), parameters are black

Core Equations

$$P_t^f = \beta(1 + \lambda_{h,t})\mathbf{E}_t \left(\frac{\Lambda_{h,t+1}}{\Lambda_{h,t}} \right) \quad (1)$$

$$P_t^f = \beta(1 + \lambda_{n,t})\mathbf{E}_t \left(\frac{\Lambda_{n,t+1}}{\Lambda_{n,t}} \right) \quad (2)$$

$$P_t^s = \beta\mathbf{E}_t \left[\frac{\Lambda_{h,t+1}}{\Lambda_{h,t}} (P_{t+1}^s + D_{t+1}) \right] \quad (3)$$

$$\lambda_{h,t}(b_{h,t+1} + \underline{B}) = 0 \quad (4)$$

$$\lambda_{n,t}(b_{n,t+1} + \underline{B}) = 0 \quad (5)$$

$$c_{h,t} + P_t^f b_{h,t+1} + \frac{P_t^s}{\mu} = P_t^s + D_t + \frac{\chi \bar{K} - B_t^n}{\mu} + W_t \quad (6)$$

$$c_{n,t} + P_t^f b_{h,t+1} = P_t^s + \frac{B_t^n}{1-\mu} + W_t \quad (7)$$

replace it with low of motion of K

~~$$q_t = \beta \mathbf{E}_t \left\{ \frac{\Lambda_{h,t+1}}{\Lambda_{h,t}} \left[\theta Z_t K_t^{\theta-1} - \frac{I_{t+1}}{K_{t+1}} + q_{t+1} \left(1 - \delta + \Phi \left(\frac{I_{t+1}}{K_{t+1}} \right) \right) \right] \right\} \right\} \quad (8)$$~~

Additional Equations

$$q_t \Phi' \left(\frac{I_{t+1}}{K_{t+1}} \right) = 1 \quad (9)$$

$$w_t = (1 - \theta) Z_t \left(\frac{K_t}{L_t} \right)^\theta, \quad L_t = 1 \quad (10)$$

~~$$w_t = (1 - \theta) Z_t \left(\frac{K_t}{L_t} \right)^\theta, \quad L_t = 1 \quad \text{redundant} \quad (11)$$~~

$$\beta \frac{\Lambda_{h,t+1}}{\Lambda_{h,t}} = \beta^{\frac{1-\alpha}{1-\rho^h}} \left(\frac{c_{h,t+1}}{c_{h,t}} \right)^{-\rho^h} \left[\frac{\frac{U_{h,t+1}}{c_{h,t}}}{\left[\left(\frac{U_{h,t}}{c_{h,t}} \right)^{1-\rho^h} - (1-\beta) \right]^{1/(1-\rho^h)}} \right] \quad (12)$$

$$\beta \frac{\Lambda_{n,t+1}}{\Lambda_{n,t}} = \beta^{\frac{1-\alpha}{1-\rho^n}} \left(\frac{c_{n,t+1}}{c_{n,t}} \right)^{-\rho^n} \left[\frac{\frac{U_{n,t+1}}{c_{n,t}}}{\left[\left(\frac{U_{n,t}}{c_{n,t}} \right)^{1-\rho^n} - (1-\beta) \right]^{1/(1-\rho^n)}} \right] \quad (13)$$

$$U_{h,t} = \left\{ (1-\beta) c_{h,t}^{1-\rho^h} + \beta \left[\mathbf{E}_t(\mathbf{U}_{h,t+1}^{1-\alpha}) \right]^{\frac{1-\rho^h}{1-\alpha}} \right\}^{1/(1-\rho^h)} \quad (14)$$

$$U_{n,t} = \left\{ (1-\beta) c_{n,t}^{1-\rho^n} + \beta \left[\mathbf{E}_t(\mathbf{U}_{n,t+1}^{1-\alpha}) \right]^{\frac{1-\rho^n}{1-\alpha}} \right\}^{1/(1-\rho^n)} \quad (15)$$

$$D_t = Z_t K_t^\theta L_t^{1-\theta} - W_t L_t - I_t - (1 - P_t^f) \chi \bar{K} \quad (16)$$

$$B_{t+1}^n = (1 - \mu) b_{n,t+1} \quad (17)$$

$$\Phi(I/K) = \frac{\phi_k}{2} \left(\frac{I}{K} \right)^2$$

Low of Motion

$$\ln Z_{t+1} = \phi \ln Z_t + \epsilon_{t+1}, \quad \epsilon \sim N(0, \sigma_\epsilon^2) \quad (18)$$

(Ky, Bny) of today become (Kx, Bnx) of tomorrow.

We approximte the normally distributed shocks (in the laws of motion for the exogenous states) by Gauss-Hermite quadrature with only three quadrature points. Accordingly, the three realizations of these five shocks are as follows:

1.

$$\varepsilon_{A,t}(i) = \sqrt{2} \cdot q_p(i) \cdot s_A, i = 1, \dots, 3$$

where

$$q_p(1) = -1.224744871, q_p(2) = 0.000000000, q_p(3) = 1.224744871$$

The probabilities/weights of these realizations are given by:

$$q_w(1) = 0.2954089751/\sqrt{\pi}, q_w(2) = 1.181635900/\sqrt{\pi}, q_w(3) = 0.2954089751/\sqrt{\pi}$$

3 Solving the model with DNN

Let us denote the state of the economy by x , the policy by y , and the policy function by θ , thus:

1.

$$x = [K_t, B_t^n, Z_t] \in \mathbb{R}^3$$

2.

$$y = \left[c_{h,t}, c_{n,t}, I_t, B_{t+1}^n, \lambda_{h,t}, \lambda_{n,t}, P_t^s, P_t^f, q_t, W_t, \frac{\Lambda_{h,t+1}}{\Lambda_{h,t}}, \frac{\Lambda_{n,t+1}}{\Lambda_{n,t}}, U_{h,t}, U_{n,t}, D_t, b_{h,t+1}, b_{n,t+1} \right] \in \mathbb{R}^{17}$$

3.

$$\theta : x \Rightarrow y, \quad \theta : \mathbb{R}^3 \rightarrow \mathbb{R}^{17}$$

We approximate the policy function θ using a DNN with trainable prameters ρ :

1.

$$\mathcal{N}(\rho, x) \cong \theta(x) = \left[c_{h,t}, c_{n,t}, I_t, B_{t+1}^n, \lambda_{h,t}, \lambda_{n,t}, P_t^s, P_t^f, q_t, W_t, \frac{\Lambda_{h,t+1}}{\Lambda_{h,t}}, \frac{\Lambda_{n,t+1}}{\Lambda_{n,t}}, U_{h,t}, U_{n,t}, D_t, b_{h,t+1}, b_{n,t+1} \right] (x)$$

2.

$$\mathcal{L}(\rho, \mathcal{D}) = \frac{1}{|\mathcal{D}|} \sum_{x \in \mathcal{D}} \omega_1(e_x^1(\rho))^2 + \dots + \omega_{17}(e_x^{17}(\rho))^2$$

where \mathcal{D} is the training set and ρ are the trainable prameters, and $e_x^i(\rho), i = 1, \dots, 17$ are the residuals from the equilibrium equations, that is the right hand side of equations 1 to 17 when evaluated at x using the policy function $\mathcal{N}(\rho, x)$. The weight vector ω can initially be set equal to the unit vector, yet for scaling reasons we might want to play around with it.

4 Parameters

Symbol (Code)	Parameter	Value
α (alpha)	Risk aversion	6
β (beta)	Discount rate	0.9966
$b\theta$ (theta)	Cap. share	0.30
ρ^h (rho_h)	inv IES stock	1/0.3
ρ^h (rho_n)	inv IES nonstock	1/0.1
δ (delta)	depreciation rate	0.0066
μ (mu)	participation rate	0.2
ϕ_k (phi_k)	adjustment cost	0.4
χ (chi)	leverage ratio	0.005
\bar{K} (Kbar)	steadysate capital	$((1/\beta - 1 + \delta)/\theta)(1/(\theta - 1))$
\bar{B} (Bbar)	borrowing constraint	$-0.1 * (1 - \theta) * K_{ss}^\theta$