

# FNCE 924 — Problem Set 2

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## Part I: Asset Prices with Complete Markets

Consider the Lucas (1978) asset pricing model and assume consumers have CRRA preferences  $\left(u(c_t) = \frac{c_t^{1-\gamma}}{1-\gamma}\right)$  with risk aversion  $\gamma$  and discount rate  $\beta$ . Suppose that income/dividend/endowment growth obeys

$$\log(y_{t+1}/y_t) = g + \epsilon_{t+1}$$

where  $\epsilon_{t+1}$  is i.i.d.  $N(0, \sigma^2)$

a) Compute the market price of a risk free bond that is purchased in period  $t$  and pays exactly one unit of consumption in period  $t + 1$ .

The Lucas asset pricing model implies identical agents that can trade a stock and a bond. Shares in the aggregate endowment are given by  $\theta^s(S^t) \in \Theta^s$  and shares of the bond are given by  $\theta^b(S^t) \in \Theta^b$ . Each household solves the following problem

$$V(\theta^s, \theta^b, S) = \max_{\{c, \theta'^s, \theta'^b\}} \left\{ u(c) + \beta \int_S V(\theta'^s, \theta'^b, S') F(S, dS') \right\}$$

subject to the constraint  $c + \theta'^s Q^s + \theta'^b Q^b = [y + Q^s] \theta^s + \theta^b$ . Letting  $\lambda(S)$  denote the Lagrange multiplier in state  $S$ , we get the following first order conditions

$$\begin{aligned} u_c(c(\theta^s, \theta^b, S)) - \lambda(S) &= 0 \\ \beta \int_S V_{\theta^b}(\theta'^s, \theta'^b, S') F(S, dS') - \lambda(S) Q^b(S) &= 0 \end{aligned}$$

The bond price is given by combining the Envelope condition  $\left(V_{\theta^b} = \lambda(S) = u_c(c)\right)$  and the FOC

$$\begin{aligned} Q^b(S) &= \beta E_t \left[ \frac{u_c(c')}{u_c(c)} \right] \\ &= \beta E_t \left[ \left( \frac{c'}{c} \right)^{-\gamma} \right] \\ &= \beta E_t \left[ \exp \left\{ -\gamma \log \left( \frac{c'}{c} \right) \right\} \right] \\ &= \beta \exp \left\{ -\gamma g + \frac{1}{2} \gamma^2 \sigma^2 \right\} \end{aligned}$$

where the final equality holds because in equilibrium  $c_t = y_t$  and by the properties of log-normally distributed variables.

b) Next, compute the market price,  $Q_j^b$ , of a risk free bond that is purchased in period  $t$  and pays exactly one unit of consumption in period  $t + j$  for any maturity  $j$ .

The pricing kernel from time  $t$  to time  $j$  is given by

$$M_{t,t+j} = \beta^j \frac{u_c(c_{t+j})}{u_c(c_t)}$$

so the price of the bond is given by

$$\begin{aligned} Q_{t+j}^b &= E_t [M_{t+j}] \\ &= \beta^j E_t \left[ \left( \frac{c_{t+j}}{c_t} \right)^{-\gamma} \right] \\ &= \beta^j E_t \left[ \left( \frac{c_{t+j}}{c_{t+j-1}} \right)^{-\gamma} \dots \left( \frac{c_{t+1}}{c_t} \right)^{-\gamma} \right] \\ &= \beta^j \left( \exp \left( j(-\gamma g + \frac{1}{2} \gamma^2 \sigma^2) \right) \right) \end{aligned}$$

where the final equality holds again by  $c_t = y_t$  in equilibrium and properties of log-normal variables.

c) We can make the income/ dividend/ endowment growth process have a time-varying mean.

$$\log(y_{t+1}/y_t) \sim N(g_{t,t+1}, \sigma^2)$$

where  $g_{t,t+1}$  stands for the growth from  $t$  to  $t + 1$  and could follow an AR(1) process:

$$g_{t,t+1} = (1 - \varphi)\bar{g} + \varphi g_t + \sigma_g^2 \epsilon_{t+1} \quad (1)$$

with  $\epsilon \stackrel{iid}{\sim} N(0, 1)$  and  $\varphi \in (-1, 1)$ .

Then the time  $t$  price of a bond paying 1 unit of consumption for sure at time  $t + j$  would be:

$$\begin{aligned} Q_j^B &= \beta^j E_t \left[ \frac{u_c(c_{t+j})}{u_c(c_t)} \right] \\ &= \beta^j E_t \left[ \left( \frac{c_{t+j}}{c_t} \right)^{-\gamma} \right] \\ &= \beta^j E_t \left[ \exp \left\{ -\gamma \log \left( \frac{c_{t+j}}{c_t} \right) \right\} \right] \\ &= \beta^j \exp \left\{ E_t \left[ -\gamma \log \left( \frac{c_{t+j}}{c_t} \right) \right] + \frac{1}{2} \text{Var}_t \left[ -\gamma \log \left( \frac{c_{t+j}}{c_t} \right) \right] \right\} \quad \text{due to log normality} \\ &= \beta^j \exp \left\{ E_t \left[ -\gamma \log \left( \frac{y_{t+j}}{y_t} \right) \right] + \frac{1}{2} \text{Var}_t \left[ -\gamma \log \left( \frac{y_{t+j}}{y_t} \right) \right] \right\} \quad \text{due to Lucas (1978) resource constraint} \end{aligned}$$

Since we defined the process of log endowment growth above as the AR(1) process in equation 1 we can then

recursively solve for the two terms:

$$\begin{aligned}
E_t \left[ \log \left( \frac{y_{t+j}}{y_t} \right) \right] &= E_t [g_{t,t+j}] \\
&= \sum_{i=0}^{j-1} \varphi^i (1 - \varphi) \bar{g} + \varphi^j g_t \\
\text{Var}_t \left[ \log \left( \frac{y_{t+j}}{y_t} \right) \right] &= \text{Var}_t [g_{t,t+j}] \\
&= \text{Var}_t \left[ \sum_{i=0}^{j-1} \varphi^i \sigma_g \epsilon_{t+1+i} \right] && \text{already omitting constant terms} \\
&= \sigma_g^2 \sum_{i=0}^{j-1} \varphi^{2i} && \text{due to iid'ness of } \epsilon
\end{aligned}$$

Then the time  $t$  price of a bond paying 1 unit of consumption for sure at time  $t + 1$  is

$$Q_j^B = \beta^j \exp \left\{ -\gamma \left[ \sum_{i=0}^{j-1} \varphi^i (1 - \varphi) \bar{g} + \varphi^j g_t \right] + \frac{1}{2} \left[ \gamma^2 \sigma_g^2 \sum_{i=0}^{j-1} \varphi^{2i} \right] \right\} \quad (2)$$

Consequently, the yield of maturity  $j$  is:

$$R_j^b = (Q_j^B)^{-1/j} = \beta^{-1} \exp \left\{ -\gamma \left[ \sum_{i=0}^{j-1} \varphi^i (1 - \varphi) \bar{g} + \varphi^j g_t \right] + \frac{1}{2} \left[ \gamma^2 \sigma_g^2 \sum_{i=0}^{j-1} \varphi^{2i} \right] \right\}^{-1/j} \quad (3)$$

We see that due to the stationarity of the mean process  $g_t$  the further we look into the future ( $j \rightarrow \infty$ ), the less the most current observation  $g_t$  impacts our outlook and the closer we get to the long-term mean. Thereby the rate of convergence (i.e. the slope of the yield curve) depends on the persistence of the process which is in turn determined by how close  $\varphi$  is to  $|1|$ . For instance with the following parameters  $\gamma = 2$ ,  $\bar{g} = 0.1$ ,  $\sigma_g = 0.3$ ,  $g_t = 0.03$  we get the following yield curves:

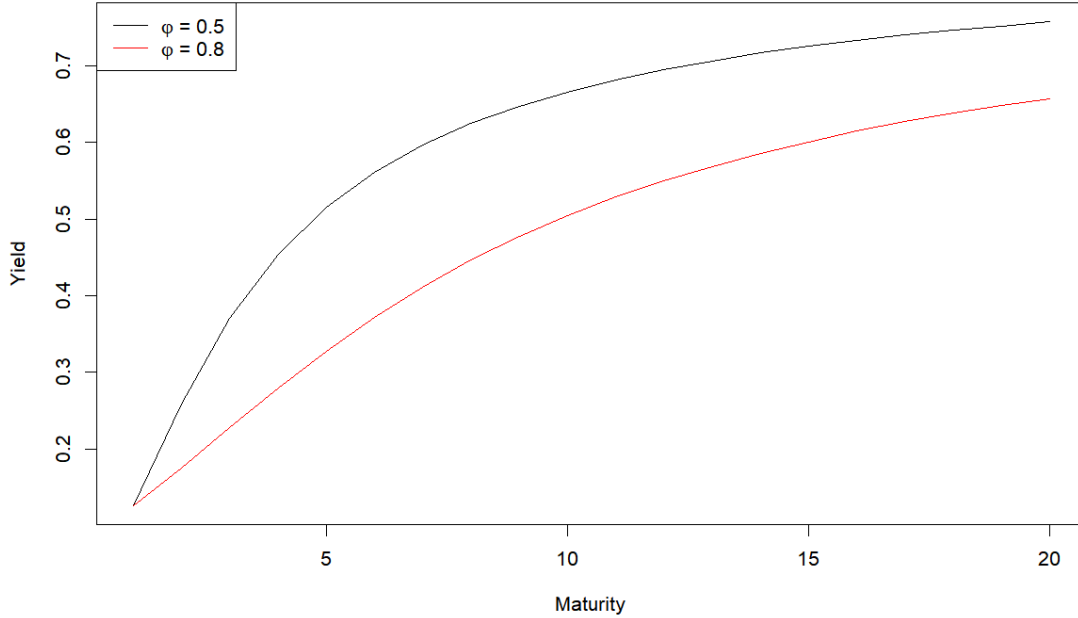


Figure 1: Implied yield curve with time-varying mean of the endowment growth process

## Part II: Computing Asset Prices in Incomplete Markets

a) The parameters of the economy in which each of the 5,000 households optimizes are:

$$\begin{aligned}\gamma &= 2 \\ \beta &= 0.97 \\ k_{t+1} &\geq -1 \\ \log y_t &= \rho \log y_{t-1} + \sigma(1 - \rho^2)^{1/2} \epsilon_t \text{ with } \rho = 0.6, \sigma = 0.35, \epsilon \sim^{iid} WN(0,1)\end{aligned}$$

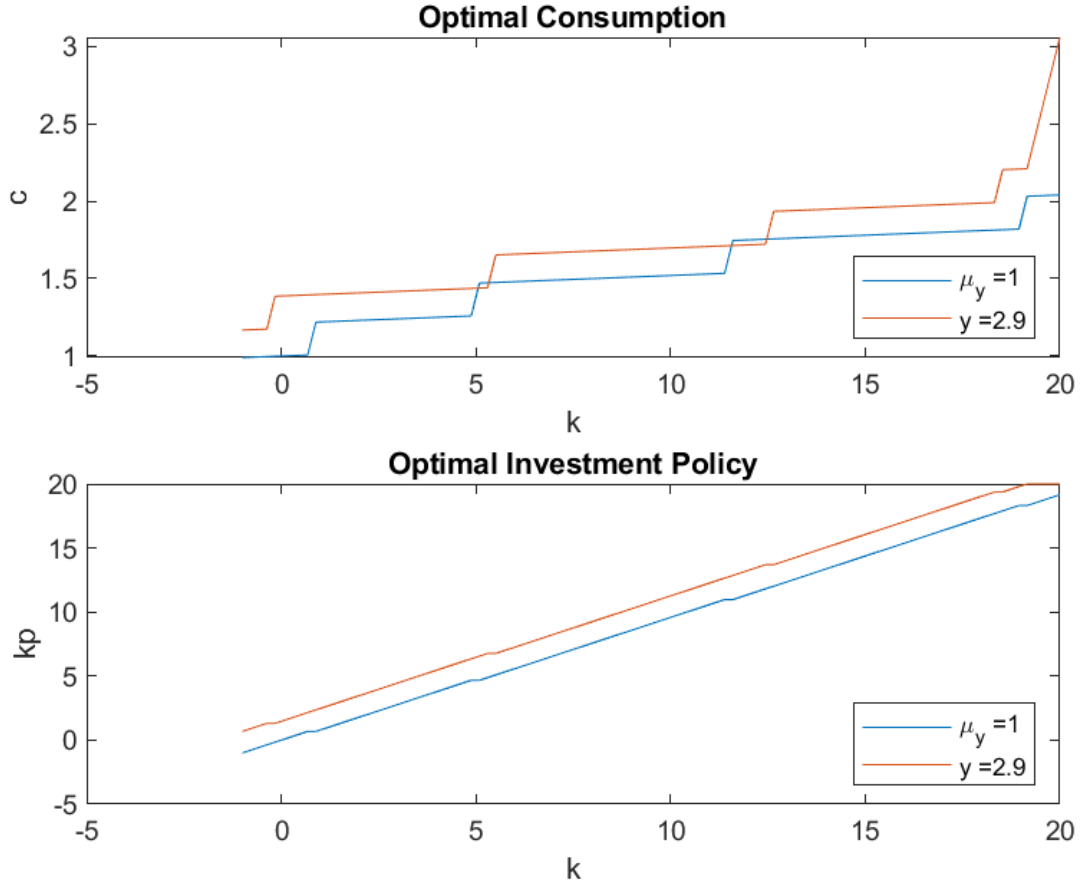


Figure 2: Optimal policies of each household for  $R = 1.01$

Figure 2 gives the optimal policies of consumption and investment of a household for the mean/ median income  $\mu_y$  and for the highest income levels.

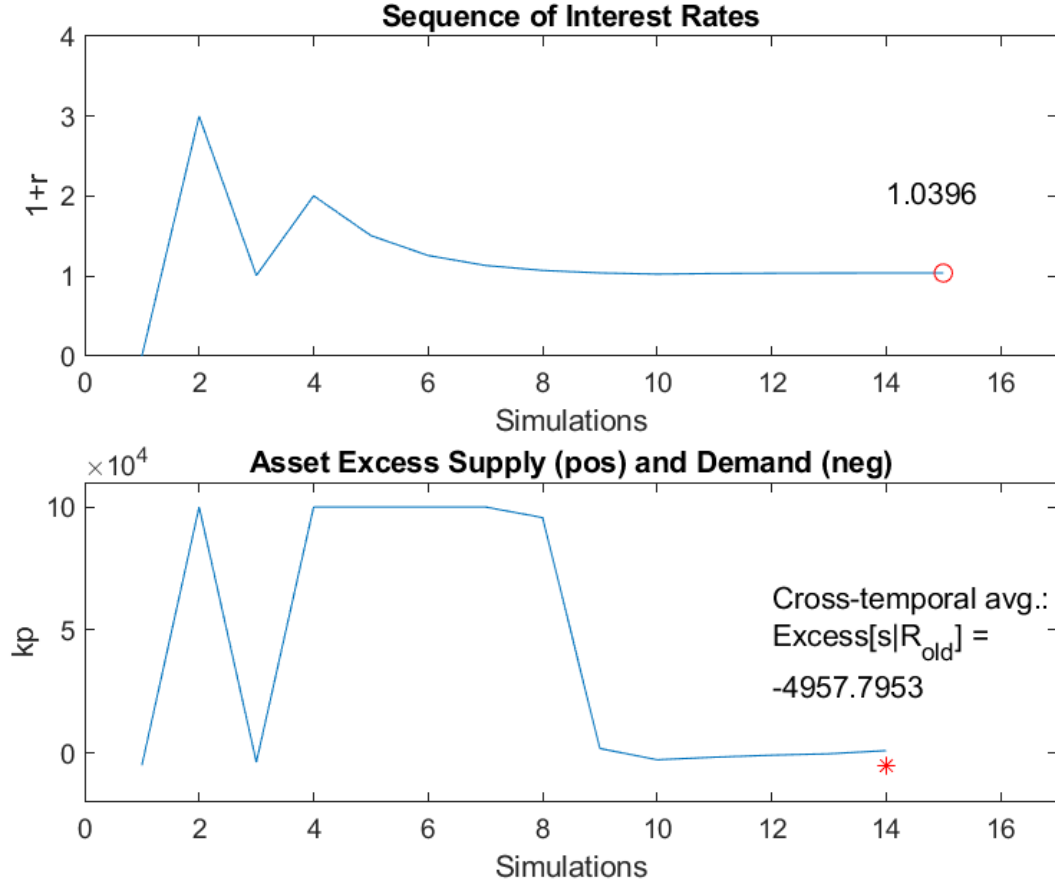


Figure 3: Convergence to Equilibrium with 5000 Households

Firstly, since the (stable) economy comprises of 2000 periods the asset excess demand/ supply in Figure 3 is the following aggregation per simulation  $s$ :

$$Excess(s) = \frac{1}{2000} \sum_{t=1}^{2000} supply_{t,k'}$$

where

$$supply_{t,k'} = \sum_{i=1}^{5000} k'_{t,i}$$

is the economy's aggregate demand across all households  $i$  in period  $t$ .

*Extra Credit* – Note the red  $\star$  in the lower graph of Figure 3: We see that when we allow agents to borrow more,  $k_{min} \geq -2$  rather than the previous lower bound of  $-1$ , agents naturally make use of this newly extended constraint. However, since we left  $1 + r$  at its previous equilibrium level more people borrow than lend with  $Excess$  being clearly negative, i.e. the markets don't clear period-by-period under this price – the asset price  $r$  would need to increase for deterring this new excess borrowing and thus for supply and demand to re-balance.

### Part III: Investment and Asset Prices

a) The one period gross return,  $R_{t,t+1}^K$  to the accumulation of physical capital,  $k$ , via the first order condition for  $k$  by the firm:

$$\begin{aligned}
 v_0 &= E_0 \sum_{t=0}^{\infty} M_{0,t} d_t = E_0 \sum_{t=0}^{\infty} M_{0,t} (\pi_t - I_t - \Phi_t) \text{ s.t. } I_t = K_{t+1} - (1 - \delta)K_t \\
 \frac{\partial v_0}{\partial k_{t+1}} : 0 &= E_0 \left\{ \left[ -\frac{\partial I_t}{\partial k_{t+1}} - \frac{\partial \Phi_t}{\partial I_t} \frac{\partial I_t}{\partial k_{t+1}} \right] + M_{0,t+1} \left[ \frac{\pi_{t+1}}{\partial k_{t+1}} - \frac{\partial I_{t+1}}{\partial k_{t+1}} - \left( \frac{\partial \Phi_{t+1}}{\partial k_{t+1}} + \frac{\partial \Phi_{t+1}}{\partial I_{t+1}} \frac{\partial I_{t+1}}{\partial k_{t+1}} \right) \right] \right\} \\
 1 &= E_0 \left\{ \frac{M_{0,t+1}}{M_{0,t}} \underbrace{\left[ \frac{\pi_{t+1}}{\partial k_{t+1}} - \frac{\partial I_{t+1}}{\partial k_{t+1}} - \left( \frac{\partial \Phi_{t+1}}{\partial k_{t+1}} + \frac{\partial \Phi_{t+1}}{\partial I_{t+1}} \frac{\partial I_{t+1}}{\partial k_{t+1}} \right)}_{\equiv R_{t,t+1}^K} \right]}_{\equiv R_{t,t+1}^K} \right\} \\
 &= E_0 \left[ M_{t,t+1} \left( \frac{a_{t+1} - \gamma(-\delta + \eta_{t+1} + 1)^2 - \delta + \frac{\gamma k_{t+2}^2}{k_{t+1}^2} + 1}{1 + 2\gamma \left( \frac{i_t}{k_t} - \eta_t \right)} \right) \right] \blacksquare
 \end{aligned}$$

b) First, rewrite the problem with  $q$  as the Lagrange multiplier on the investment constraint:

$$\mathcal{L} = E_0 \sum_{t=0}^{\infty} M_{0,t} (\pi_t - I_t + q_t (I_t - k_{t+1} K_t (1 - \delta)))$$

Then we can solve the problem w.r.t.  $I_t$ :  $\Psi_t = \max_{I_t} \{-I_t - \Phi_t + q_t I_t\}$  giving us  $q_t^* = 1 + \frac{\partial \Phi_t}{\partial I_t}$ . Then using the fact from the lecture that under linear homogeneity  $v_t^e = q_t^* k_{t+1}$  we can rewrite the return to the owner of the firm:

$$\begin{aligned}
 R_{t,t+1}^S &= \frac{v_{t+1}^e + d_{t+1}}{v_t^e} \\
 &= \frac{\frac{\partial I_{t+1}}{\partial k_{t+1}} k_{t+1} + k_{t+2}}{(1 + \frac{\partial \Phi_{t+1}}{\partial I_{t+1}}) k_{t+2} + \pi_{t+1} - \overbrace{I_{t+1}}^{\frac{\partial I_{t+1}}{\partial k_{t+1}} k_{t+1}} - \Phi_{t+1}} \\
 &= \frac{(1 + \frac{\partial \Phi_{t+1}}{\partial I_{t+1}}) k_{t+2} + \pi_{t+1} - \overbrace{I_{t+1}}^{\frac{\partial I_{t+1}}{\partial k_{t+1}} k_{t+1}} - \Phi_{t+1}}{(1 + \frac{\partial \Phi_t}{\partial I_t}) k_{t+1}} \\
 &= \frac{\frac{\partial \pi_{t+1}}{\partial k_{t+1}} k_{t+1} - \frac{\partial I_{t+1}}{\partial k_{t+1}} k_{t+1} - \Phi_{t+1} + \overbrace{\frac{\partial \Phi_{t+1}}{\partial I_{t+1}} k_{t+2}}^{\substack{= -\frac{d\Phi_{t+1}}{dk_{t+1}} k_{t+1} \\ = 2\gamma \left[ \frac{I_{t+1}}{k_{t+1}} - \eta_{t+1} \right]}}}{(1 + \frac{\partial \Phi_t}{\partial I_t}) k_{t+1}} \\
 &= \frac{\frac{\pi_{t+1}}{\partial k_{t+1}} - \frac{\partial I_{t+1}}{\partial k_{t+1}} - \left( \frac{\partial \Phi_{t+1}}{\partial k_{t+1}} + \frac{\partial \Phi_{t+1}}{\partial I_{t+1}} \frac{\partial I_{t+1}}{\partial k_{t+1}} \right)}{\frac{\partial I_t}{\partial k_{t+1}} + \frac{\partial \Phi_t}{\partial I_t} \frac{\partial I_t}{\partial k_{t+1}}} \\
 &= \frac{a_{t+1} - \gamma(-\delta + \eta_{t+1} + 1)^2 - \delta + \frac{\gamma k_{t+2}^2}{k_{t+1}^2} + 1}{1 + 2\gamma \left( \frac{i_t}{k_t} - \eta_t \right)} \\
 &= R_{t,t+1}^K \blacksquare
 \end{aligned}$$

## Appendix - Matlab Code

```

1 %%
2 % *%%* Patrick's & Felix' Code*
3 %Task (1a)
4 %% Discretize  $\log(y)$ -AR process
5
6 L = 19;
7 ny = L; % number of grid points
8 rho_lny = 0.6; % autocorrelation coefficient
9 bar_lny = 0/(1-rho_lny); % unconditional mean of lny
10 sigma_eps = 0.35*sqrt(1-rho_lny^2);
11
12 [grid_lny,P,d]=tauchen1(ny,bar_lny,rho_lny,sigma_eps);
13
14 %% Household's problem
15 % VFI
16
17 % Set parameter values
18 r = 1.01;
19 sigma = 1/2; % = 1/risk aversion
20 beta = 0.97;
21
22 S = 2500; % number simulated periods
23 H = 5000; % number of households
24
25 % Set capital grid parameters
26 M = 100;
27 nk = M + 1;
28 nkp = nk;
29 kmin = -1; % borrowing limit
30 kmax = -20*kmin;
31 kstep = (kmax - kmin)/(nk - 1);
32
33 k = [kmin: kstep: kmax]'; % Grid for current wealth
34 kp = k; % Grid for next period wealth
35
36 y = exp(grid_lny);
37
38 rate_seq = [0, 3, r]; %bounds as starting value for interest rate chain
39 suppl_seq = [H*kmin-1, kmax*H+1]; %bounds for capital suppl-demand chain, in line w/ R=1 and
40 R=10
41 alpha = 3; % counter for rates evaluated
42
43 tic;
44 toc_seq = 0;
45 while abs(rate_seq(end)-rate_seq(end-1)) > 0.0005
46     % equivalent result for the following while cond. (H*kstep := max dev. dep. on the
47     % economy):
48     % (abs(suppl_seq(end)-suppl_seq(end-1)) > H*kstep) || (suppl_seq(end)*suppl_seq(end-1)
49     % >0)
50
51 % Compute momentary utility function
52 c = zeros(nkp, nk, ny);
53 U = zeros(nkp, nk, ny);
54
55 for j = 1:ny
56     for i = 1:nk
57         c(:,i,j) = max( (r*k(i) + y(j))*ones(size(kp)) - kp, 0.001);
58     end
59 end
60
61 U = (c.^(1 - 1/sigma) - 1)/(1 - 1/sigma);
62
63 % Initial guess for the value function
64 V0 = zeros(nk,ny);

```

```

64 init = repmat(kp,1,size(y,2))+repmat(y,size(kp,1),1);
65 V1 = (init.^(1 - 1/sigma) - 1)/(1 - 1/sigma);
66
67 kp_opt_ind = zeros(nk,ny);
68
69 err = 1;
70
71 % Start iterations to determine optimal value function
72 while err > 0.0001
73     for i=1:ny
74         [V1(:,i), kp_opt_ind(:,i)] = max(U(:, :, i) + beta*V0(:, :)*P(i, :)*ones(1,nkp));
75         % for every income level y_i pick the max kp along the k grid
76         % use transition matrix to find expected next periods value fct. which
77         % is effectively a probability weighted average value function (expected value)
78     end
79     % for each level of y_i we add last round's optimal kp picks (which is
80     % V0, so it's multiplied by beta etc.) which is constant per level of k
81     % (thus upscaled in the kp dimension
82     temp = (V1 - V0).^2;
83     err = sum(temp(:))/numel(V1);
84     V0 = V1; %we now have a matrix of kp values on a k*y grid
85 end
86
87 nybar = 0.5*(ny+1);
88
89 if r == 1.01
90     figure(1)
91     subplot(2,1,1)
92     plot(k,V1(:,nybar))
93     title(['\mu_y = ' num2str(round(y(nybar),3))])
94     xlabel('k')
95     ylabel('V(\cdot)')
96     subplot(2,1,2)
97     plot(k,V1(:,end))
98     title(['y = ' num2str(round(y(end),3))])
99     xlabel('k')
100    ylabel('V(\cdot)')
101    %saveas (figure(1),[pwd ' /02_Graphs/II_1a_TwoLevels.png'])
102 end
103
104
105 %Task (2c)
106 %% Extract c* and kp*
107
108 kp_opt = kp*ones(1,ny);
109 kp_opt = kp_opt(kp_opt_ind);
110
111 c_opt = transpose(max( (ones(ny,1)*r*k' + y'*ones(1,nk)) - kp_opt', 0.001));
112
113 if r == 1.01
114     figure(2)
115     subplot(2,1,1)
116     plot(k,c_opt(:,nybar), k,c_opt(:,end))
117     title('Optimal Consumption')
118     legend(strcat('\mu_y = ', num2str(round(y(nybar),1))), ...
119           strcat('y = ', num2str(round(y(end),1))), 'Location', 'SouthEast')
120     xlabel('k')
121     ylabel('c')
122     subplot(2,1,2)
123     plot(k,kp_opt(:,nybar), k,kp_opt(:,end))
124     title('Optimal Investment Policy')
125     legend(strcat('\mu_y = ', num2str(round(y(nybar),1))), ...
126           strcat('y = ', num2str(round(y(end),1))), 'Location', 'SouthEast')
127     xlabel('k')
128     ylabel('kp')
129
130 saveas (figure(2),[pwd ' /02_Graphs/II_a_kpAndcPerK.png'])
131 end

```



```

132
133 %Task (2b)
134 %% Simulate the economy
135 chain_length = S;
136 chain = zeros(H,chain_length);
137 chain(:,1) = ones(H,1)*(ny+1)/2; % initial state number to start chain
138 %chain(1:3,1) = [1 4 9]
139
140 % preallocate memory to store y realizations chosen by Markov chain
141 y_markov = zeros(H, chain_length);
142 grid_lny_H = repmat(grid_lny,H,1);
143 y_markov(:,1) = grid_lny_H(1,chain(:,1));
144
145 for i=2:chain_length
146     distribution = P(chain(:,i-1), :);
147     cumulative_distribution = cumsum(distribution,2);
148     j = rand(H,1);
149     chain(:,i) = max(1,L-sum(cumulative_distribution>j,2));
150     % L the number of states in y, sum() gives row-wise no of cols in cdf
151     % matrix P where j is exceeded, so L-sum() is col-number in P, max is
152     % to handle cases where even first row exceeds j - avoiding col=0
153     y_markov(:,i) = grid_lny_H(1,chain(:,i));
154 end
155 y_chain = exp(y_markov);
156 k_chain = zeros(H,S+1); % requires one column more as policy concerns kp
157 k_chain(:,1) = repmat(k(0.5*(nk+1)),H,1);
158 c_chain = zeros(H,S);
159
160 %k_chain(1:3,1) = k(1:3);
161
162 for j = 1:S
163     [lia, loc] = ismember(k_chain(:,j),k);
164     [lib, locy] = ismember(y_chain(:,j),y);
165     k_chain(:,j+1) = diag(kp_opt(loc,locy));
166
167 c_chain(:,j) = (max( (r*k_chain(:,j) + y_chain(:,j) ) - k_chain(:,j+1), 0.001));
168 end
169 % cut out first 20% of obs for stable economy (0.2*S=500, from 501
170 % onwards and accounting for kp being shifted by +1 makes +2)
171
172 k_stable = k_chain(:,round(0.20*S,0)+2:end);
173
174 supply = sum(k_stable,1);
175 suppl_seq = [suppl_seq mean(supply)];
176
177 if suppl_seq(alpha)*suppl_seq(alpha-1)<0 %eq. in the middle of n and n-1
178     r = 0.5*(rate_seq(alpha)+rate_seq(alpha-1)); %avg
179 elseif suppl_seq(alpha)*suppl_seq(alpha-2)<0 %eq. in the middle of n and n-2
180     r = 0.5*(rate_seq(alpha)+rate_seq(alpha-2)); %avg
181 elseif suppl_seq(alpha)<0 %avg with last opposing sign rate
182     r = 0.5*(rate_seq(alpha)+rate_seq(find(suppl_seq>0,1,'last')));
183 else r = 0.5*(rate_seq(alpha)+rate_seq(find(suppl_seq<0,1,'last')));
184 end
185
186 rate_seq = [rate_seq r];
187 alpha = alpha +1
188 toc1 = toc
189 toc_seq = [toc_seq toc1];
190 end
191 eq_rate = r;
192 X = [ 'End of loop, target interest rate is ', num2str(round(eq_rate,2)) ];
193 disp(X)
194
195 %% Change borrowing constraint but not r
196
197 kmin = -2; % *NEW* borrowing limit
198 kstep = (kmax - kmin)/(nk - 1);
199

```

```

200 k = [kmin: kstep: kmax]'; % Grid for current wealth
201 kp = k; % Grid for next period wealth
202
203
204 % Compute momentary utility function
205 c = zeros(nkp, nk, ny);
206 U = zeros(nkp, nk, ny);
207
208 for j = 1:ny
209 for i = 1:nk
210 c(:,i,j) = max( (r*k(i) + y(j))*ones(size(kp)) - kp, 0.001);
211 end
212 end
213
214 U = (c.^(1 - 1/sigma) - 1)/(1 - 1/sigma);
215
216 % Initial guess for the value function
217
218 V0 = zeros(nk,ny);
219 init = repmat(kp,1,size(y,2))+repmat(y,size(kp,1),1);
220 V1 = (init.^(1 - 1/sigma) - 1)/(1 - 1/sigma);
221
222 kp_opt_ind = zeros(nk,ny);
223
224 err = 1;
225
226 % Start iterations to determine optimal value function
227 while err > 0.0001
228 for i=1:ny
229 [V1(:,i), kp_opt_ind(:,i)] = max(U(:,:,i) + beta*V0(:,:,i)*P(i,:)'*ones(1,nkp));
230 % for every income level y_i pick the max kp along the k grid
231 % use transition matrix to find expected next periods value fct. which
232 % is effectively a probability weighted average value function (expected value)
233 end
234 % for each level of y_i we add last round's optimal kp picks (which is
235 % V0, so it's multiplied by beta etc.) which is constant per level of k
236 % (thus upscaled in the kp dimension
237 temp = (V1 - V0).^2;
238 err = sum(temp(:))/numel(V1);
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240 end
241
242 nybar = 0.5*(ny+1);
243
244 %% Extract c* and kp*
245
246 kp_opt = kp*ones(1,ny);
247 kp_opt = kp_opt(kp_opt_ind);
248
249 c_opt = transpose(max( (ones(ny,1)*r*k' + y'*ones(1,nk)) - kp_opt', 0.001));
250
251 %% Simulate the economy
252 chain_length = S;
253 chain = zeros(H,chain_length);
254 chain(:,1) = ones(H,1)*(ny+1)/2; % initial state number to start chain
255 %chain(1:3,1) = [1 4 9]
256
257 % preallocate memory to store y realizations chosen by Markov chain
258 y_markov = zeros(H, chain_length);
259 grid_lny_H = repmat(grid_lny,H,1);
260 y_markov(:,1) = grid_lny_H(1,chain(:,1));
261
262 for i=2:chain_length
263 distribution = P(chain(:,i-1), :);
264 cumulative_distribution = cumsum(distribution,2);
265 j = rand(H,1);
266 chain(:,i) = max(1,L-sum(cumulative_distribution>j,2));
267 % L the number of states in y, sum() gives row-wise no of cols in cdf

```

```

268 % matrix P where j is exceeded, so L-sum() is col-number in P, max is
269 % to handle cases where even first row exceeds j - avoiding col=0
270 y_markov(:,i) = grid_lny_H(1,chain(:,i));
271 end
272 y_chain = exp(y_markov);
273 k_chain = zeros(H,S+1); % requires one column more as policy concerns kp
274 k_chain(:,1) = repmat(k(0.5*(nk+1)),H,1);
275 c_chain = zeros(H,S);
276
277 for j = 1:S
278 [lia, loc] = ismember(k_chain(:,j),k);
279 [lib, locy] = ismember(y_chain(:,j),y);
280 k_chain(:,j+1) = diag(kp_opt(loc,locy));
281
282 c_chain(:,j) = (max( (r*k_chain(:,j) + y_chain(:,j)) - k_chain(:,j+1), 0.001));
283 end
284 % cut out first 20% of obs for stable economy (0.2*S=500, from 501
285 % onwards and accounting for kp being shifted by +1 makes +2)
286
287 k_stable = k_chain(:,round(0.20*S,0)+2:end);
288
289 supply_new = sum(k_stable,1);
290 excess_new = mean(supply_new);
291
292 %% Plot results
293 figure(3)
294 subplot(2,1,1)
295 plot(rate_seq)
296 title('Sequence of Interest Rates')
297 xlabel('Simulations')
298 ylabel('1+r')
299 axis([0 alpha+2 0 4])
300
301 N = [num2str(round(rate_seq(end),4))]; % If "N" is not cellstr or string datatype, must be
    column vector
302 labelinds = [length(rate_seq)];
303 text(labelinds-1,rate_seq(labelinds)+1,N);
304 hold on
305 plot(labelinds,rate_seq(labelinds),'ro')
306
307 subplot(2,1,2)
308 plot(suppl_seq)
309 title('Asset Excess Supply (pos) and Demand (neg)')
310 xlabel('Simulations')
311 ylabel('kp')
312 axis([0 alpha+2 -0.2*H*kmax 1.1*H*kmax])
313
314 N = "Cross-temporal avg.:" + "\n" + "Excess[s|R-{old}] =" + "\n" + num2str(excess_new);
315 N = compose(N);
316 labelinds = [length(suppl_seq)];
317 text(labelinds-2,excess_new+0.5*H*kmax,N);
318 hold on
319 plot(labelinds,mean(supply_new),'r*')
320
321 saveas(figure(3),[pwd '/02_Graphs/II_d-RateConvergence.png'])
322
323 figure(4)
324 plot(toc_seq/60,'o')
325 xlabel('Simulations')
326 ylabel('Minutes')
327 saveas(figure(4),[pwd '/02_Graphs/II_d-Time.png'])

```