ECON 6140 - Problem Set # 3

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Open Sector Growth Model via Dynamic Programming

(1) Let $\tau_t^c = \tau^c$ and $\tau_t^x = \tau^x$ be constant. The functional equation is given by

$$(Tv)(k) = \max_{0 \le k' \le \frac{f(k)}{(1+\tau^x)} + (1-\delta)k} \left\{ u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (k' - (1-\delta)k) \right) + \beta v(k') \right\}$$

- (2) Our state variable is the current value of k and our control is the future value of capital
 - In our setting, this makes sense since we can't affect the current value of capital, but we can decide how much to consumes/invest to pinpoint capital in the next period.
 - Note that we can't set our state space equal to [0, 1]. In fact, part 4-7 shows a steady state value of capital that is not even remotely close to be inside [0,1].
- (3) Note that T is defined on a the set of bounded function defined on $[0,\infty)$.

To show that T is a contraction, we can use Blackwell's theorem. In fact, we only need to show monotonicity and discounting for T

(i) Monotonicity

Let $v(x) \leq w(x)$ and $g_v(k)$, $g_w(k)$ be the respective policy function. Then,

$$(Tv)(k) = \max_{0 \le k' \le \frac{f(k)}{(1+\tau^x)} + (1-\delta)k} \left\{ u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (k' - (1-\delta)k) \right) + \beta v(k') \right\}$$

$$= u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (g_v(k) - (1-\delta)k) \right) + \beta v(g_v(k))$$

$$\le u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (g_v(k) - (1-\delta)k) \right) + \beta w(g_v(k))$$

$$\le u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (g_f(k) - (1-\delta)k) \right) + \beta w(g_w(k))$$

$$= \max_{0 \le k' \le \frac{f(k)}{(1+\tau^x)} + (1-\delta)k} \left\{ u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (k' - (1-\delta)k) \right) + \beta w(k') \right\}$$

$$= (Tv)(k) \le (Tw)(k)$$

$$\Rightarrow (Tv)(k) \le (Tw)(k)$$

(ii) Discounting

$$(Tv + a)(k) = \max_{0 \le k' \le \frac{f(k)}{(1+\tau^x)} + (1-\delta)k} \left\{ u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (k' - (1-\delta)k) \right) + \beta(v(k') + a) \right\}$$

$$= \max_{0 \le k' \le \frac{f(k)}{(1+\tau^x)} + (1-\delta)k} \left\{ u \left(\frac{f(k)}{(1+\tau^c)} - \frac{(1+\tau^x)}{(1+\tau^c)} (k' - (1-\delta)k) \right) + \beta v(k') \right\} + \beta a$$

$$= (Tv)(k) + \beta a$$

Hence, T is a contraction.

(4) Figure 1 show the value function around the steady state and Figure 2 shows the policy function.

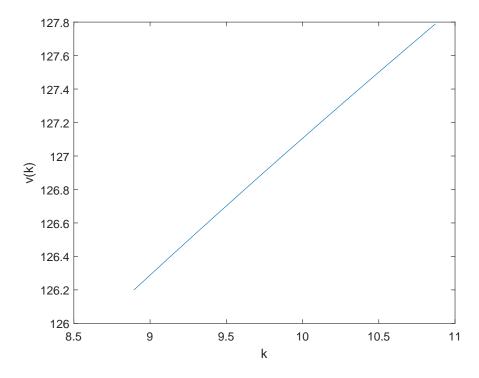


Figure 1: Value function

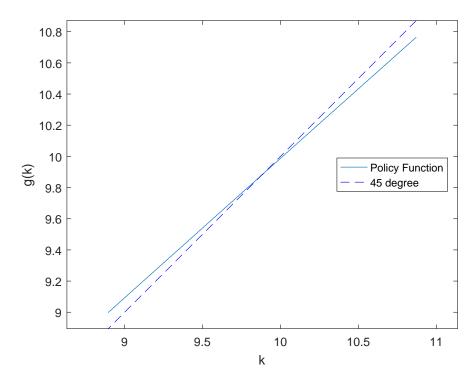


Figure 2: Policy function

(5) Figure 3 show the time path of capital starting from $k_0 = 0.9k^*$.

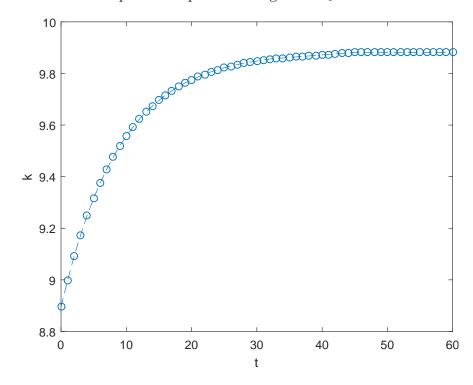


Figure 3: Time path for consumption and capital

(6) We use the shooting algorithm for this question. For sake of clarity, we provide the

dynamic equations:

$$k_{t+1} = \frac{k_t^{\alpha}}{(1+\tau_t^x)} + (1-\delta)k_t - c_t \frac{(1+\tau_t^c)}{(1+\tau_t^x)}$$

$$c_{t+1} = c_t \left(\frac{(1+\tau_t^c)(1+\tau_{t+1}^x)}{(1+\tau_t^x)(1+\tau_{t+1}^c)}\right)^{\frac{1}{\sigma}} \left(\frac{\alpha\beta k_t^{\alpha-1}}{(1+\tau_{t+1}^x)} + \beta(1-\delta)\right)^{\frac{1}{\sigma}}$$

Figure 4 show the time path of capital for a permanent tax increase. Note that the consumption jumps at t = 1 to get back to the steady path.

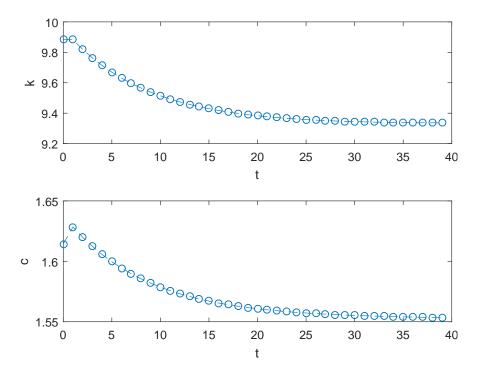


Figure 4: Time path for consumption and capital

(7) Again we use the shooting algorithm for this question. Figure 5 show the time path of capital for a temporary tax increase.

In this setting, the steady path moves both at t = 1 and t = 10 explaining the two jumps in consumption to insure that we will reach the steady state in the future.

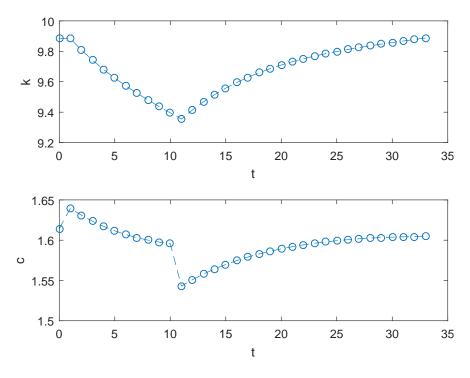


Figure 5: Time path for consumption and capital

Ben-Porath

(1) $(i) \Rightarrow$

Let $Y^* = \max_{s(t),h(t)} \int_0^\infty \exp(-rt)w(t)(1-s(t))h(t)dt$ and $\{c(t),s(t),h(t)\}_{t=0}^\infty$ solve the household problem.

Let's assume that $\{s(t), h(t)\}$ does not maximizes $\int_0^\infty \exp(-rt)w(t)(1-s(t))h(t)dt$, i.e. $\exists Y$ such that

$$\int_0^\infty \exp(-rt)w(t)(1-s(t))h(t)dt = Y < Y^*$$

Let $c'(t) = c(t) + \epsilon$ for some $\epsilon > 0$. Clearly, c' > c due to strictly increasing nature of $u(\cdot)$. If, we can find an ϵ such that c'(t) is feasible, then we have a contradiction with $\{c(t), s(t), h(t)\}_{t=0}^{\infty}$ solving the household problem.

$$\begin{split} \int_0^\infty \exp(-rt)c'(t)dt &= \int_0^\infty \exp(-rt)c(t)dt + \int_0^\infty \exp(-rt)\epsilon dt \\ &\leq \int_0^\infty \exp(-rt)w(t)(1-s(t))h(t)dt + \frac{\epsilon}{r} \\ &\leq Y + \frac{\epsilon}{r} \end{split}$$

Since Y < Y', there exists an $\epsilon > 0$ such that $Y + \frac{\epsilon}{r} \le Y^*$. Hence, there exists some $\{s'(t),h'(t)\}_{t=0}^{\infty}$ such that $\int_{0}^{\infty} \exp(-rt)w(t)(1-s'(t))h'(t)dt = Y^*$ which implies that $\{c'(t),s'(t),h'(t)\}_{t=0}^{\infty}$ is feasible, i.e. we have a contradiction.

(ii) **⇐**

Let $\{s(t), h(t)\}_{t=0}^{\infty}$ solve $\max_{s(t), h(t)} \int_{0}^{\infty} \exp(-rt)w(t)(1-s(t))h(t)dt = Y$. Let $\{c(t)\}_{t=0}^{\infty}$ solve the household problem and $\{c'(t)\}_{t=0}^{\infty}$ be any solution such that c' > c

Since $\{c(t)\}_{t=0}^{\infty}$ solves the household problem with the following budget constraint,

$$\int_0^\infty \exp(-rt)c(t)dt \le Y$$

we have that

$$\int_0^\infty \exp(-rt)c'(t)dt > Y$$

Therefore, any $\{s'(t),h'(t)\}_{t=0}^{\infty}$ that supports $\{c'(t)\}_{t=0}^{\infty}$ it must be the case that $int_0^{\infty}\exp(-rt)w(t)(1-s'(t))h'(t)dt > Y$ which is a contradiction.

Hence, $\{c(t), s(t), h(t)\}_{t=0}^{\infty}$ solves the household problem.

(2) The current Hamiltonian is given by

$$\mathcal{H}(s,h,\mu) = w(t)(1-s(t))h(t) + \mu(t)(\phi(s(t)h(t)) - \delta_h h(t))$$

where h is the state variable, s the control variable and μ the costate variable.

The FOCs, TVC, and constraint are given by

$$s: -w(t)h(t) + \mu(t)h(t)\phi'(s(t)h(t)) = 0$$

$$h: w(t)(1 - s(t)) + \mu(t)(s(t)\phi'(s(t)h(t)) - \delta_h) = r\mu(t) - \dot{\mu}(t)$$

$$TVC: \lim_{t \to \infty} \exp(-rt)\mu(t)h(t) = 0$$

$$\vdots \dot{h} = \phi(s(t)h(t)) - \delta_h h(t)$$

(3) First, we differentiate the first FOC

$$\mu(t)\phi''(x(t))\dot{x}(t) + \phi'(x(t))\dot{\mu}(t) = \dot{w}(t)$$

We can then combine the FOCs in the following way

$$-\dot{\mu}(t) = w(t)(1 - s(t)) + \frac{w(t)}{\phi'(x(t))}(s(t)\phi'(x(t)) - \delta_h - r)$$

$$-\dot{\mu}(t) = w(t) + \frac{w(t)}{\phi'(x(t))}(-\delta_h - r)$$

$$-\phi'(x(t))\dot{\mu}(t) = \phi'(x(t))w(t) + w(t)(-\delta_h - r)$$

$$\mu(t)\phi''(x(t))\dot{x}(t) - \dot{w}(t) = \phi'(x(t))w(t) - w(t)(\delta_h + r)$$

$$\Rightarrow \dot{x}(t) = \frac{1}{\mu(t)\phi''(x(t))}(\dot{w}(t) + w(t)(\phi'(x(t)) - \delta_h - r))$$

(4) In steady state, $\dot{\mu} = \dot{h} = 0$. Hence,

$$0 = \phi(x^*(t)) - \delta_h h^*(t)$$

$$\Rightarrow h^*(t) = \frac{\phi(x^*(t))}{\delta_h}$$

and

$$\phi'(x^*(t)) = \delta_h + r$$

$$\Rightarrow x^*(t) = \phi'^{-1}(\delta_h + r)$$

This implies that

$$\Rightarrow h^*(t) = \frac{\phi(\phi'^{-1}(\delta_h + r))}{\delta_h}$$
$$\Rightarrow s^*(t) = \frac{x^*(t)}{h^*(t)} = \frac{\delta_h \phi'^{-1}(\delta_h + r)}{\phi(\phi'^{-1}(\delta_h + r))}$$

We have that h^* and s^* are uniquely determined by δ_h , r and the shape of $\phi(\cdot)$. Note that $\phi(\cdot)$ is strictly increasing while $\phi'^{-1}(\cdot)$ is strictly decreasing.

Additionally, w(t) does not influence the steady state human capital accumulation and the schooling decision. This contrasts the fact that w(t) does influence the path of x(t), but not the steady state.

Code

```
1 % Value Function Iteration + Shooting Algorithm
 % Course: ECON 6140
  % Version: 1.0
4 % Author: Julien Neves
  clear, clc;
  % Question 4
  n = 1000;
                  % Size of grid
                  % labor share
  alpha = 0.33;
                   % depreciation of capital
  delta = 0.05;
11
  sigma = 0.5;
                   % CRRA
12
                  % discount factor
  beta = 0.98;
  tau_x = 0.01;
                   % technology
                  % technology
  tau_c = 0.01;
15
16
  crit = 1;
                  % Initialize convergence criterion
```

```
tol = .001;
                   % Convergence tolerance
19
  T = 60;
20
21
  % Grid
22
  kstar = ((alpha*beta)/((1-beta*(1-delta))*(1+tau_x)))^(1/(1-alpha)
     ); % Steady state
  cstar = (kstar^alpha-delta*kstar*(1+tau_x))/(1+tau_c);
25
  ub_k = 0.9 * kstar:
                            % Upper bound
26
  lb_k = 1.1 * kstar;
                            % Lower bound
27
  kgrid = linspace(lb_k, ub_k, n)'; % Create grid
28
29
  k0 = 0.9 * kstar;
30
31
  % Empty
                           % Initialize temporary value function
  val_temp = zeros(n,1);
     vector
  val_fun = zeros(n,1);
                            % Initialize value function vector
  pol_fun = zeros(n,1);
                           % Initialize policy function vector
36
               % Initialize iteration counter
  ite = 0;
37
38
  % Value function iteration
  while crit>tol;
40
      % Iterate on k
41
       for i=1:n
42
           c = (kgrid(i)^alpha + (1+tau_x)*((1-delta)*kgrid(i) -
43
              kgrid))/(1+tau_c); % Compute consumption for kt
           utility_c = c.^(1-sigma)/(1-sigma); % Compute utility for
44
              every ct
           utility_c (c<=0) = -Inf; % Set utility to -Inf for c<=0
45
           [val\_fun(i), pol\_fun(i)] = max(utility\_c + beta*val\_temp);
46
                % Solve bellman equation
      end
47
       crit = max(abs(val_fun-val_temp));
                                             % Compute convergence
48
          criterion
       val_temp = val_fun; % Update value function
49
       ite = ite + 1 % update iteration
  end
51
  % Value function
53
  figure (1)
  plot (kgrid , val_fun )
  xlabel('k')
```

```
ylabel('v(k)')
  print -depsc fig1.eps
  % Policy function
60
  figure (2)
  plot(kgrid, kgrid(pol_fun))
  axis equal
  hold on
  plot (kgrid, kgrid, '---b')
  xlabel('k')
  ylabel('g(k)')
67
  legend('Policy Function', '45 degree', 'Location', 'best')
  print -depsc fig2.eps
70
71
  % Question 5
  kpath=zeros(1,T);
  [\tilde{\ }, kpath(1)] = min(abs(kgrid-k0));
74
75
  for i = 1:T
       kpath(i+1) = pol_fun(kpath(i));
77
  end
78
  kpath = kgrid (kpath);
79
  % Capital path
  figure (3)
  plot (0: length (kpath) -1, kpath, '---o');
  xlabel('t') % x-axis label
  ylabel('k') % y-axis label
  print -depsc fig3.eps
86
87
  % Question 6
  kstar_old = kstar;
89
  cstar_old = cstar;
90
91
  N = 100000; % grid size
  T = 100:
               % time periods
  tol = .01;
                    % Convergence tolerance
94
  tau_c = repmat(.03, 1, T);
                                % technology
96
  tau_x = repmat(.05, 1, T);
                                % technology
97
98
  kstar = ((alpha*beta)/((1-beta*(1-delta))*(1+tau_x(T))))^(1/(1-beta*(1-delta)))
      alpha)); % Steady state
  cstar = (kstar^alpha-delta*kstar*(1+tau_x(T)))/(1+tau_c(T));
                                                                         % c
```

```
steady state
101
   k0 = kstar_old; % starting k value
102
103
                          % initial k path vector
   k = zeros(1,T+1);
104
                          % initial c path vector
   c = zeros(1, T+1);
105
106
   lb_k = 0.9 * kstar;
                          % lower bound of k axis
107
   ub_k = 1.1 * kstar;
                          % upper bound of k axis
108
109
   lb_c = 0.9*cstar;
                          % lower bound of c axis
110
   ub_c = 1.1 * cstar;
                          % lower bound of c axis
111
112
   kgrid = linspace(lb_k, ub_k, n);
                                           % k axis
113
   cgrid = linspace(lb_c, ub_c, n);
                                           % c axis
114
115
                 % initialize tolerance criteria
   crit = 1;
116
   ite = 1:
                 % initialize iteration
117
118
   while (crit>tol && ite <=N)
119
        k(1) = k0; % set starting k0
120
        c(1) = cgrid(ite); \% pick c0
121
        for t = 1:T-1
122
            123
                tau_c(t))/(1+tau_x(t)); % compute k(t+1)
            A = (1 + tau_c(t)) * (1 + tau_x(t+1)) / ((1 + tau_x(t))) * (1 + tau_c(t))
124
                +1)); % tax scaling factor
            c(t+1) = c(t)*(A*(alpha*beta*k(t)^(alpha-1)/(1+tau_x(t+1)))
125
                +beta*(1-delta)))^(1/sigma); %compute c(t+1)
             \operatorname{crit} = \max(\operatorname{abs}(\operatorname{kstar} - \operatorname{k}(t+1)), \operatorname{abs}(\operatorname{cstar} - \operatorname{c}(t+1)));
                                                                         %
126
                deviation from steady state
             if crit <=tol
127
                 % if close to steady state stop algorithm
128
                 k = k(1:t+1); % cut path after convergences
129
                 c = c(1:t+1); % cut path after convergences
130
                 break
131
             else
132
                 continue
133
            end
134
        end
135
        ite = ite + 1\% update iteration
136
   end
137
138
   k = [kstar_old, k]; \% add k0
   c = [cstar_old, c]; \% add c0
```

```
141
   % plot time path of k and c
142
   figure (4)
143
   subplot (2,1,1)
144
   plot(0:length(k)-1,k, '--o');
   xlabel('t') % x-axis label
146
   ylabel ('k') % y-axis label
147
   subplot (2,1,2)
148
   plot(0:length(c)-1,c, '--o');
149
   xlabel('t') % x-axis label
150
   ylabel('c') % y-axis label
151
   print -depsc fig4.eps
152
153
   \% Question 7
154
   tau_c = [repmat(.03, 1, 10), repmat(.01, 1, T-10)];
                                                           % technology
155
   tau_x = [repmat(.05, 1, 10), repmat(.01, 1, T-10)];
                                                           % technology
156
157
   kstar = ((alpha*beta)/((1-beta*(1-delta))*(1+tau_x(T))))^(1/(1-beta*(1-delta)))
158
      alpha)); % Steady state
   cstar = (kstar^alpha-delta*kstar*(1+tau_x(T)))/(1+tau_c(T));
                                                                            % c
159
       steady state
160
   k0 = kstar_old; % starting k value
161
162
   k = zeros(1,T+1);
                          % initial k path vector
163
   c = zeros(1,T+1);
                          % initial c path vector
164
165
   lb_k = 0.9 * kstar;
                          % lower bound of k axis
166
   ub_k = 1.1 * kstar;
                          % upper bound of k axis
167
168
   lb_c = 0.9*cstar;
                          % lower bound of c axis
169
   ub_c = 1.1*cstar;
                          % lower bound of c axis
170
171
   kgrid = linspace(lb_k, ub_k, n);
                                          % k axis
172
   cgrid = linspace(lb_c, ub_c, n);
                                          % c axis
173
174
                 % initialize tolerance criteria
   crit = 1:
175
   ite = 1;
                % initialize iteration
176
177
   while (crit > tol \&\& ite <= N)
178
        k(1) = k0; % set starting k0
179
        c(1) = cgrid(ite); \% pick c0
180
        for t = 1:T-1
181
            k(t+1) = k(t) \cdot alpha/(1+tau_x(t))+(1-delta)*k(t)-c(t)*(1+tau_x(t))
182
                tau_c(t))/(1+tau_x(t)); % compute k(t+1)
```

```
A = (1 + tau_c(t)) * (1 + tau_x(t+1)) / ((1 + tau_x(t))) * (1 + tau_c(t+1)) / ((1 + tau_x(t))) * (1 + tau_c(t+1)) / ((1 + tau_x(t+1))) / ((1 + tau_x(t+1))) * (1 + tau_c(t+1)) / ((1 + tau_x(t+1))) / ((1 + tau_x(t+
183
                                                   +1)); % tax scaling factor
                                        c(t+1) = c(t)*(A*(alpha*beta*k(t)^(alpha-1)/(1+tau_x(t+1)))
184
                                                  +beta*(1-delta)))^(1/sigma); %compute c(t+1)
                                         \operatorname{crit} = \max(\operatorname{abs}(\operatorname{kstar} - \operatorname{k}(t+1)), \operatorname{abs}(\operatorname{cstar} - \operatorname{c}(t+1)));
                                                                                                                                                                                                                                        %
185
                                                   deviation from steady state
                                         if crit <=tol
186
                                                      % if close to steady state stop algorithm
187
                                                       k = k(1:t+1); % cut path after convergences
188
                                                       c = c(1:t+1); % cut path after convergences
189
                                                       break
190
                                         else
191
                                                       continue
192
                                        end
193
                         end
194
                          ite = ite + 1\% update iteration
195
          end
196
197
          k = [kstar\_old, k]; \% add k0
198
           c = [cstar_old, c]; \% add c0
199
200
          \% plot time path of k and c
201
           figure (5)
202
           subplot (2,1,1)
           plot(0:length(k)-1,k, '--o');
204
           xlabel('t') % x-axis label
205
           ylabel ('k') % y-axis label
206
           subplot (2,1,2)
207
           plot (0: length (c) -1,c, '--o');
208
           xlabel('t') % x-axis label
209
           ylabel('c') % y-axis label
210
           print -depsc fig5.eps
```