

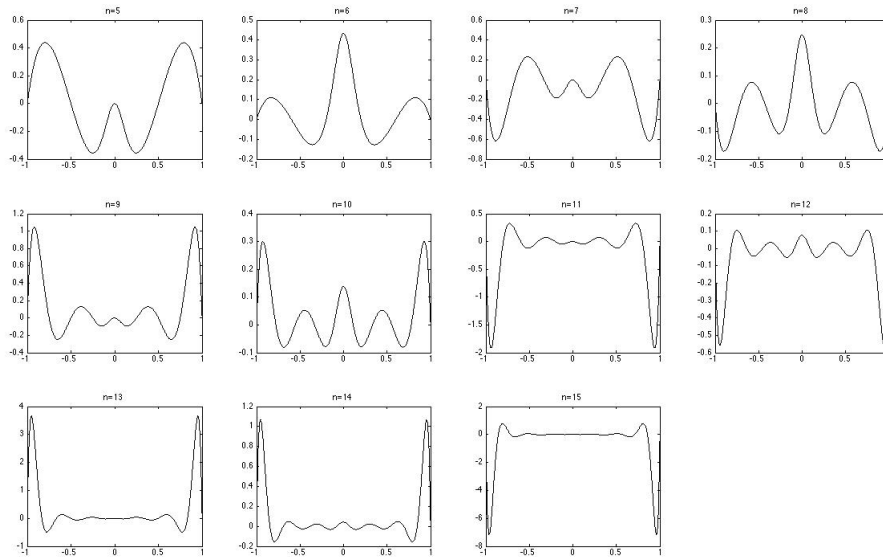
Computational Economics: Problem Set 4

Yangming Bao, ID: 5601239

Cheung Ying Lun, ID: 5441897

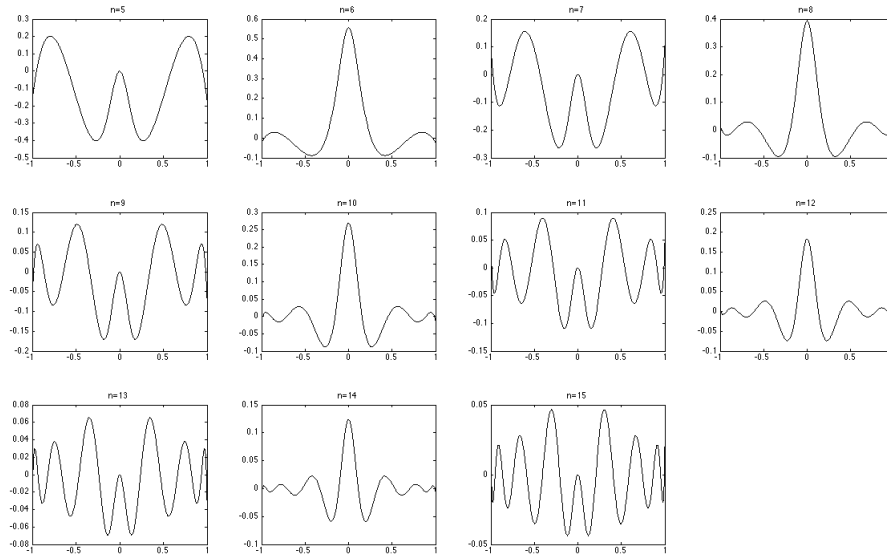
Problem 1: Interpolation of Simple Function

- If using Chebychev polynomials with equidistant nodes, the solution for some large n is out of expectation since the matrix of basis function is close to 0 and it cause the inverse matrix unreasonable large. So if we compare the residual by comparing the solution when n increase from 5 to 15 with the function itself, we can see the pattern as follows. When n increases, the residual converge to zero except the



boundaries. It means the approximate function is more close to the exact function (except the boundaries) when the nodes increase.

- if using Chebychev nodes, the figure looks as follows. When n increase, one significant feature contrast with former one is that the residual at boundaries are small, actually smaller than the points in the middle. Also, the residual are more close to 0 when n increase, which indicates the approximate function is more close to the exact one evenly (including the boundaries).



Problem 2: Simple Optimization Problem

1. The FOC with respect to C_0 reads

$$-(C_0 - \bar{C}) + \mathbb{E}[W_0(1+r) - C_0 - \bar{C}] \stackrel{!}{=} 0 \quad (1)$$

$$\implies C_0^* = \frac{1}{2}\mathbb{E}[W_0(1+r)] = \frac{W_0}{2}(1 + \mathbb{E}[r]). \quad (2)$$

2. The optimal consumption at time 0 only depends on the expected return, irrespective of the variance of r . It does not make economic sense since we expect a risk-averse agent would change one's behavior when risk changes.

3. The FOC with respect to C_0 reads

$$C_0^{-\gamma} - \mathbb{E}[(W_0(1+r) - C_0)^{-\gamma}] \stackrel{!}{=} 0 \quad (3)$$

$$\implies C_0^* = (\mathbb{E}[(W_0(1+r) - C_0)^{-\gamma}])^{-\frac{1}{\gamma}}. \quad (4)$$

4. The maximum percentage deviation is 7.8759e-13%. See Figure 1.

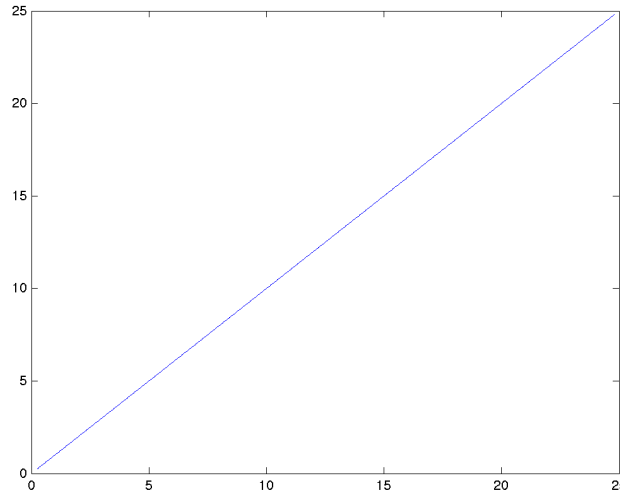


Figure 1: Solution against Chebychev Approximation

5. The maximum percentage deviation increases, but not very much.

Problem 3: Portfolio Choice

1. FOC with respect to α , we can have

$$E \left[\left(1 + r^f + \alpha(r - r^f) \right)^{-\gamma_i} (r - r^f) \right]$$

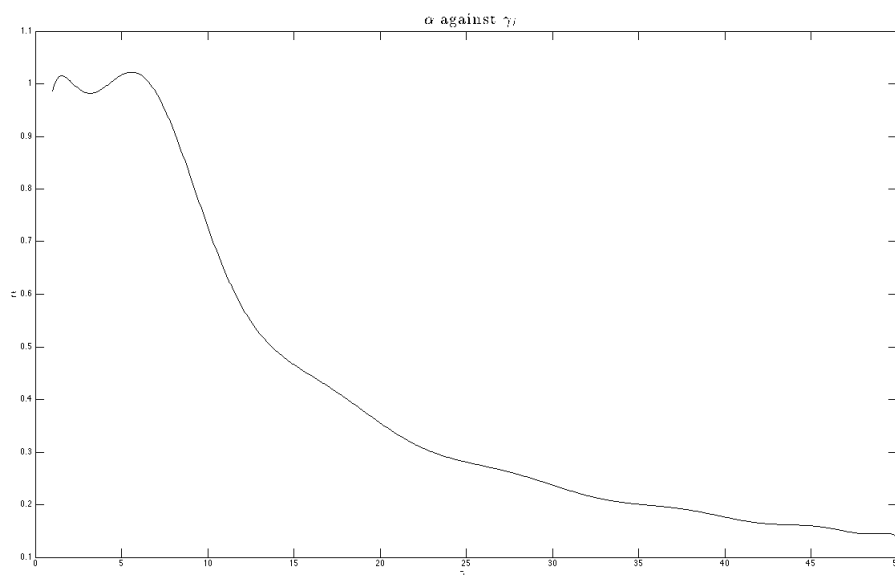
the solution of the equation is the optimal portfolio share for each agent i .

2. If we use the parametrization of problem 2, which is $r^f = 0.02$, $r^{\min} = -0.08$ and $r^{\max} = 0.12$ with the probability of 0.5, respectively, then the problem can be written as

$$\begin{aligned} 0.5 \times [(1.02 + 0.1\alpha)^{-\gamma_i} \times 0.1] + 0.5 \times [(1.02 + (-0.1)\alpha)^{-\gamma_i} \times 0.1] &= 0 \\ \Leftrightarrow (1.02 + 0.1\alpha)^{-\gamma_i} &= (1.02 - 0.1\alpha)^{-\gamma_i} \end{aligned}$$

Then it will lead to $\alpha = 0$, regardless of γ_i .

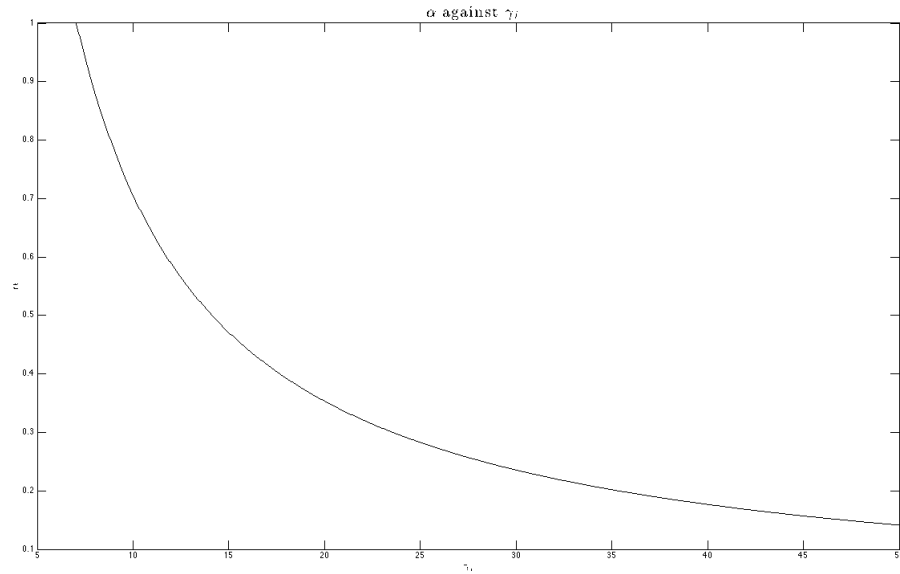
When I set $p = 0.8$, the plot of α against γ_i looks as follows, From the figure, we



can observe that in the neighborhood of the point where α is binding, the points are not exactly smooth at 1 due to the characteristics of the basic function.

3. $\bar{\gamma} = 7.04$ when α is just binding.

The graph now looks smoother than before.



Problem 4: Policy Function Approximation in the Neoclassical Growth Model

1. The Lagrangian for households is

$$\mathcal{L} = \sum_{t=0}^{\infty} \beta^t \ln C_t - \lambda_t [C_t + K_{t+1} - (1 + r_t - \delta)K_t + w_t] \quad (5)$$

The household's FOCs are

$$\beta^t C_t^{-1} - \lambda_t \stackrel{!}{=} 0 \implies \lambda_t = \beta^2 C_t^{-1} \quad (C_t)$$

$$-\lambda_t + (1 + r_{t+1} - \delta)\lambda_{t+1} \stackrel{!}{=} 0 \implies \lambda_t = (1 + r_{t+1} - \delta)\lambda_{t+1} \quad (K_{t+1})$$

Thus we have

$$C_{t+1} = \beta(1 + r_{t+1} - \delta)C_t. \quad (6)$$

For firms, the FOC is

$$\alpha K_t^{\alpha-1} - r_t \stackrel{!}{=} 0 \implies r_t = \alpha K_t^{\alpha-1}. \quad (7)$$

2. At equilibrium, both firm's and household's problems are solved. Moreover, at

steady state $C_t = C_{t+1} = C$ and $K_t = K_{t+1} = K$. Thus, we have

$$1 = \beta(1 + \alpha K^{\alpha-1} - \delta) \implies K^* = \left(\frac{\beta^{-1} + \delta - 1}{\alpha} \right)^{\alpha-1} \approx 7.2112. \quad (8)$$

Due to the market clearing, consumption equals production and

$$\tilde{C}(K) = K^\alpha. \quad (9)$$

3. See code file `Q4.m`.

4. See code file `Q4.m`.

5. See Figure 2.

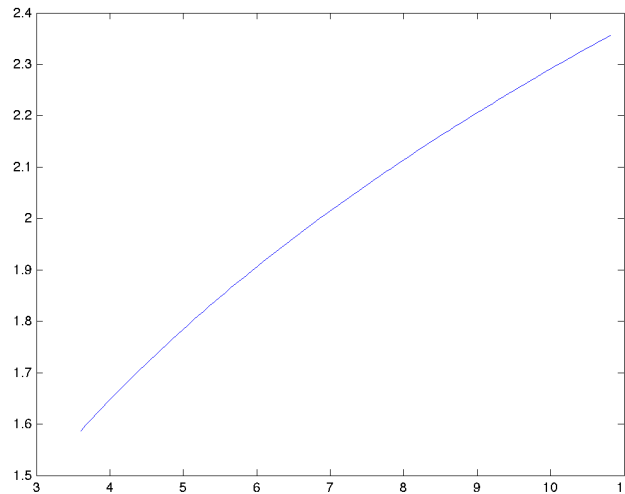


Figure 2: Consumption Policy

6. Capital grows monotonically with decreasing speed towards the steady state. See Figure 3.

7. Maximum absolute error = 3.8648e-07.

Average absolute error = 7.315e-08.

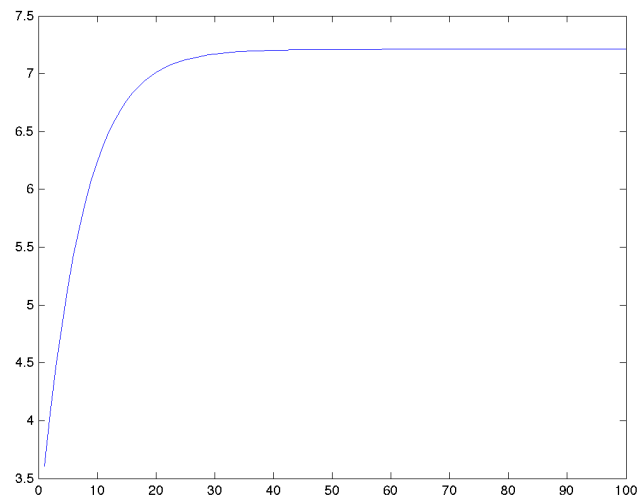


Figure 3: Capital

Question 1

Q1.m

```

1  clc; clear;
2  % cepath='/Users/baoyangming/Dropbox/gsefm/2015SoSe/Computational Economics
    /Applied Computational Economics and Finance/comecon/';
3  % path([cepath 'cetools;' cepath 'cedemos'],path);
4
5  % Set domain of interpolation
6
7  a = -1;
8  b = 1;
9
10 %% using Chebychev polynomials for n equidistant nodes.
11 % for some large n
12 n_big=1000;
13
14 % define a vector of equidistant nodes, x
15 x_big = nodeunif(n_big,a,b);
16 % define the function space for Chebychev polynomials and
17 % calculate the matrix of basis functions, T
18 fspace_big = fundefn('cheb',n_big,a,b);

```

```

19 T_big = funbas(fspace_big, x_big);
20 % calculate the function values at x
21 y_big = feval(@func, x_big');
22 % finally get the polynomial coefficients.
23 c_big = (T_big' * T_big)^(-1) * (T_big' * y_big');
24 y_bar = T_big * c_big;
25
26 figure('Name', 'Comparison using equidistant nodes');
27
28 for n=5:15
29
30     xnode = nodeunif(n, a, b);
31     fspace = fundefn('cheb', n, a, b);
32     T = funbas(fspace, xnode);
33     % calculate the function values at x
34     y = feval(@func, xnode);
35     c = (T' * T)^(-1) * (T' * y);
36
37     % plot the residual
38     T_use = funbas(fspace, x_big);
39     y_tilda = T_use * c;
40     res = y_big' - y_tilda;
41
42     subplot(3, 4, n-5+1);
43     plot(x_big, res, 'k', 'LineWidth', 1.2);
44     title(strcat('n=', num2str(n)));
45 end
46
47
48
49 %% Repeat the exercise using Chebychev nodes
50 clearvars -except a b;
51 % for some large n
52 n_big = 1000;
53 % define the function space for Chebychev polynomials and
54 % calculate the matrix of basis functions, T
55 fspace_big = fundefn('cheb', n_big, a, b);
56 x_big = funnode(fspace_big);
57 T_big = funbas(fspace_big, x_big);
58
59 % calculate the function values at x

```



```

60 y_big = feval(@func,x_big);
61
62 % finally get the polynomial coefficients.
63 c_big=(T_big'*T_big)^(-1)*T_big'*y_big;
64 y_bar = T_big*c_big;
65
66 figure('Name','Comparison using Chebychev nodes');
67 for n=5:15
68     % calculate the matrix of basis functions, B
69     fspace = fundefn('cheb',n,a,b);
70     xnode = funnode(fspace);
71     T = funbas(fspace,xnode);
72
73     % calculate the function values at x
74     y = feval(@func,xnode);
75
76     % finally get the polynomial coefficients.
77     c=(T'*T)^(-1)*(T'*y);
78
79
80     % plot the residual
81     T_use=funbas(fspace,x_big);
82     y_tilda=T_use*c;
83     res = y_bar-y_tilda;
84
85     subplot(3,4,n-5+1);
86     plot(x_big,res,'k','LineWidth',1.2);
87     title(strcat('n=',num2str(n)));
88 end

```

func.m

```

1 function y=func(x)
2 %function in Q1
3 y = 1./(1+25*x.^2);
4 end

```

Question 2

Q2.m

```

1  clc ; clear ;
2  cepath=' /Users/YingLun/Documents/Dropbox/Academic/Postgraduate/GSEFM/PhD/
    Year 2/Summersemester/computational economics/Applied Computational
    Economics and Finance/compecon/';
3  path([cepath 'cetools;' cepath 'cedemos'],path);
4
5  %% Parameters
6
7  r_min    = -0.08;
8  r_max    = 0.12;
9  gamma    = 2;
10 p        = 0.5;
11
12 func      = @(W0)Opt_Comp(W0,r_min,r_max,gamma,p);
13
14 W_max     = 50;
15 W_min     = 0.5;
16
17 n_node    = 15;
18
19 %% Compute the Chebyshev approximation
20
21 fspace    = fundefn('cheb',n_node,W_min,W_max);
22 W_grid    = funnode(fspace);
23
24 C0_hat    = func(W_grid);
25 T_hat     = funbas(fspace,W_grid);
26 alpha     = (T_hat'*T_hat)\(T_hat'*C0_hat);
27
28 W_new     = linspace(0.5,50,1000)';
29 T_new     = funbas(fspace,W_new);
30
31 C0_new    = T_new*alpha;
32
33 true_C    = func(W_new);
34 fig        = plot(true_C,C0_new);
35 % print('-dpng',fig)
36
37 disp(['Max. percentage error = ',num2str(max((true_C-C0_new)./true_C)*100),
    '%'])

```

Obj.m

```

1 function C0 = Opt.Comp(W0,r_min,r_max,gamma,p)
2 %This function computes the optimal consumption C0*
3 % INPUT:
4 %     W0: initial wealth
5 %     r_min: return at bad state
6 %     r_max: return at good state
7 %     gamma: risk aversion coefficient
8 %     p: probability of good state
9 %
10 % OUTPUT:
11 %     C0: optimal consumption at t=0
12
13 %% Initialization
14 eps = 1e-5;
15 del = 1e-5;
16 max_it = 1e6;
17 ini_Jac = eye(length(W0));
18 ini_val = W0/2;
19 func = @(C)((p*(W0*(1+r_max)-C).^(-gamma)+(1-p)*(W0*(1+r_min)-C).^(-
    gamma)).^(-1/gamma)-C);
20 stop_crit = [eps,del,max_it];
21
22 %% Compute C0
23 C0 = Inverse_Broyden_Method(func,ini_Jac,ini_val,stop_crit);
24
25 end

```

NonLCon.m

```

1 function roots = Inverse_Broyden_Method(func,ini_Jac,...
2     ini_val,...
3     stop_crit)
4 %This function perform the Newton's method for root-finding problem.
5 %     func: a function handle for value of the root-finding problem.
6 %     Jac: a function handle of the Jacobian function of the problem
7 %     ini_val: initial value
8 %     stop_crit: stopping criteria = [eps,del,max_it]
9
10 eps = stop_crit(1);
11 del = stop_crit(2);
12 max_it = stop_crit(3);

```

```
13 it = 0;
14 cont = true;
15
16 if length(ini_val)==size(ini_val,2)
17     xold = ini_val';
18 else
19     xold = ini_val;
20 end
21
22 B = inv(ini_Jac);
23 fold = func(xold);
24
25 while cont
26     it = it+1;
27
28     xnew = xold-B*fold;
29     fnew = func(xnew);
30     dx = xnew-xold;
31     df = fnew-fold;
32     B = B+((dx-B*df)*dx'*B)/(dx'*B*df);
33     if (norm(xold-xnew)<=eps*(1+norm(xnew))) || (it==max_it)
34         cont = false;
35     end
36     xold = xnew;
37     fold = fnew;
38 end
39
40 if norm(func(xnew))<=del
41     disp(['Convergence after ',num2str(it),' iterations.'])
42     roots = xnew;
43 else
44     disp('Convergence failed.')
45     roots = [];
46 end
47
48 end
```

Question 3

Q3.m

```
1
2  clc ; clear ;
3
4
5  % parameters
6  p = 0.8; % probability
7  gamma_min = 1;
8  gamma_max = 50;
9  n_node = 15;
10 fspace = fundefn('cheb',n_node,gamma_min,gamma_max);
11 gamma_grid = funnode(fspace);
12 T = funbas(fspace,gamma_grid);
13 y = func(p,gamma_grid);
14 c = (T'*T)\(T'*y);
15
16 gamma = linspace(1,50,1000);
17 T_new = funbas(fspace,gamma);
18 alpha = T_new*c;
19
20
21 % plot alpha against gamma
22 plot(gamma,alpha,'k','LineWidth',1.2);
23 ylabel('$\alpha$','Interpreter','latex');
24 xlabel('$\gamma_i$','Interpreter','latex');
25 title('$\alpha$ against $\gamma_i$','Interpreter','latex');
```

func.m

```
1 function alpha = func(p,gamma)
2 % function for Q3
3
4 c = (p/(1-p)).^(-1./gamma);
5 alpha = max(min(10.2.*(1-c)./(1+c),1),0);
6
7 end
```

Question 4

Q4.m

```

1  clear , clc
2
3  %% Parameters
4
5  beta    = 0.96;
6  alpha   = 0.36;
7  delta   = 0.06;
8
9  m       = 9;
10 N       = 100;
11
12 %% Chebyshev approximation
13
14 Kstar    = ((1/beta+delta-1)/alpha)^(1/(alpha-1));
15 fspace   = fundfn('cheb',m,0.5*Kstar,1.5*Kstar);
16 K_grid   = funnode(fspace);
17 T_hat    = funbas(fspace,K_grid);
18
19 C_hat    = K_grid.^alpha;
20 theta    = (T_hat'*T_hat)\(T_hat'*C_hat);
21
22 K_uni     = nodeunif(N,0.5*Kstar,1.5*Kstar);
23 T_new     = funbas(fspace,K_uni);
24 C_new     = T_new*theta;
25
26 fig       = figure;
27 plot(K_uni,C_new);
28 print(fig,'-dpng','Q4.5')
29
30 %% Path of capital
31 K_t       = zeros(100,1);
32 K_t(1)    = 0.5*Kstar;
33 for t=1:99
34     K_t(t+1) = K_Policy(K_t(t),alpha,beta,delta);
35 end
36 fig       = figure;
37 plot(1:100,K_t);

```

```

38 print(fig, '-dpng', 'Q4.6')
39
40 %% Approximation error
41 % K
42 K_rand = 0.5*Kstar+Kstar.*rand(1000,1);
43 T_rand = funbas(fspace, K_rand);
44 C_rand = T_rand*theta;
45 % K'
46 Kprime = K_Policy(K_rand, alpha, beta, delta);
47 Tprime = funbas(fspace, Kprime);
48 Cprime = Tprime*theta;
49 % Error
50 R = Cprime-C_rand*beta.*(1+alpha*Kprime.^(alpha-1)-delta);
51 E = R./C_rand;
52
53 disp(['Max. absolute error = ', num2str(max(abs(E)))])
54 disp(['Avg. absolute error = ', num2str(mean(abs(E)))])

```

obj_fun.m

```

1 function Kprime = K_Policy(K, alpha, beta, delta)
2 %%This function computes the capital policy rule.
3
4
5 %% Initialization
6 eps = 1e-6;
7 del = 1e-4;
8 max_it = 1e6;
9 ini_Jac = eye(length(K));
10 ini_val = K;
11 func = @(Kprime)((beta*(1+alpha*Kprime.^(alpha-1)-delta)).^(1/alpha)
    .*K-Kprime);
12 stop_crit = [eps, del, max_it];
13
14 %% Compute C0
15 Kprime = Inverse_Broyden_Method(func, ini_Jac, ini_val, stop_crit);
16
17 end

```