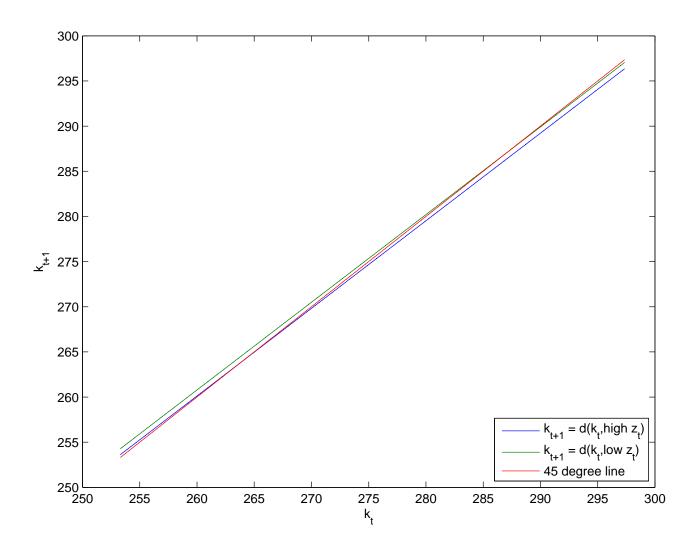
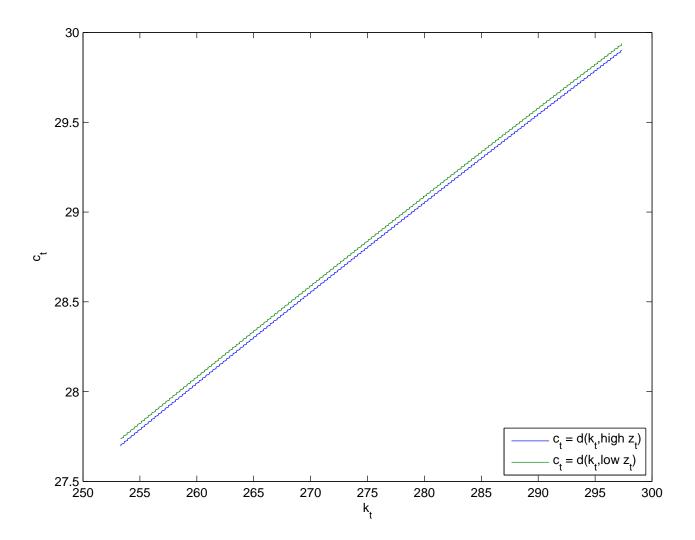
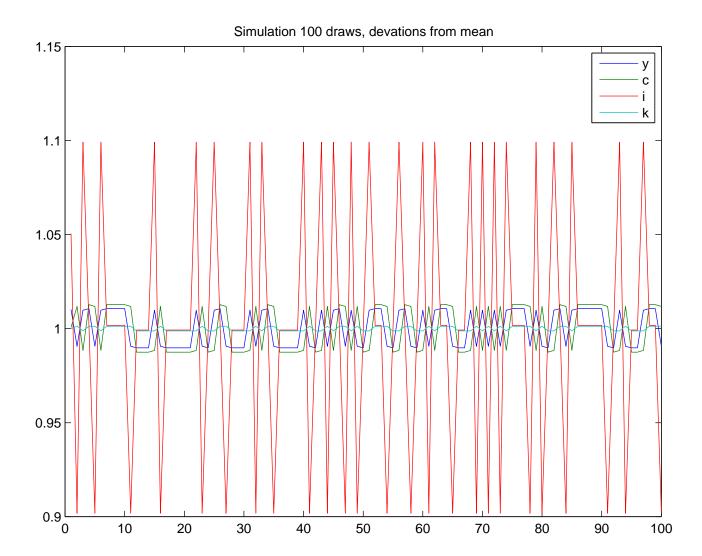
1 Stochastic dynamic programming







```
clear all;
close all;
\% Define the structural parameters of the model,
% i.e. policy invariant preference and technology parameters
% alpha : capital's share of output
% beta : time discount factor
% delta : depreciation rate
\% sigma : risk-aversion parameter, also intertemp. subst. param.
alpha = .35;
beta = .98;
delta = .025;
sigma = 2;
% Number of exogenous states
gz = 2;
% Values of z
z = [4.95]
     5.05];
% Probabilites for z
prh = .5;
prl = 1-prh;
% Expected value of z
zbar = prh*z(1) + prl*z(2);
% Find the steady-state level of capital as a function of
% the structural parameters
kstar = ((1/beta - 1 + delta)/(alpha*zbar))^(1/(alpha-1));
\% Define the number of discrete values k can take
gk = 101;
k = linspace(0.90*kstar,1.10*kstar,gk);
\mbox{\ensuremath{\mbox{\%}}} Compute a (gk x gk x gz) dimensional consumption matrix c
  for all the (gk x gk x gz) values of k_t, k_t+1 and z_t.
for h = 1 : gk
   for i = 1 : gk
       for j = 1 : gz
            c(h,i,j) = z(j)*k(h)^alpha + (1-delta)*k(h) - k(i);
            if c(h,i,j) < 0
                c(h,i,j) = 0;
            end
            \% h is the counter for the endogenous state variable k_t
            \% i is the counter for the control variable k_t+1
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% j is the counter for the exogenous state variable z_t
       end
   end
end
% Compute a (gk x gk x gz) dimensional consumption matrix u
    for all the (gk x gk x gz) values of k_t, k_t+1 and z_t.
for h = 1 : gk
   for i = 1 : gk
       for j = 1 : gz
             if sigma == 1
                 u(h,i,j) = log(c(h,i,j))
             else
                 u(h,i,j) = (c(h,i,j)^(1-sigma) - 1)/(1-sigma);
             end
             % h is the counter for the endogenous state variable k_t
             \% i is the counter for the control variable k_t+1
             % j is the counter for the exogenous state variable z_t
       end
   end
end
% Define the initial matrix v as a matrix of zeros (could be anything)
v = zeros(gk,gz);
% Set parameters for the loop
convcrit = 1E-6; % chosen convergence criterion
diff = 1;
                   % arbitrary initial value greater than convcrit
                   % iterations counter
iter = 0;
while diff > convcrit
    \% for each combination of k\_t and gamma\_t
       find the k_t+1 that maximizes the sum of instantenous utility and
        discounted continuation utility
    for h = 1 : gk
        for j = 1 : gz
             Tv(h,j) = max(u(h,:,j) + beta*(prh*v(:,1)' + prl*v(:,2)'));
        end
    end
    iter = iter + 1;
    diff = norm(Tv - v);
    v = Tv;
end
\mbox{\ensuremath{\mbox{\%}}} Find the implicit decision rule for \mbox{\ensuremath{\mbox{k}}}\mbox{\ensuremath{\mbox{t+1}}} as a function of the state
% variables k_t and z_t
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```
for h = 1 : gk
    for j = 1 : gz
       % Using the [] syntax for max does not only give the value, but
       % also the element chosen.
       [Tv,gridpoint] = \max(u(h,:,j) + beta*(prh*v(:,1)' + prl*v(:,2)'));
       % Find what grid point of the k vector which is the optimal decision
       kgridrule(h,j) = gridpoint;
       \% Find what value for k_t+1 which is the optimal decision
       kdecrule(h,j) = k(gridpoint);
    end
end
\% Plot it and save it as a jpg-file
figure
plot(k,kdecrule,k,k);
xlabel('k_t')
ylabel('k_{t+1}')
print -djpeg kdecrule.jpg
% Compute the optimal decision rule for c as a function of the state
%
    variables
for h = 1 : gk
                   % counter for k_t
    cdecrule(h,j) = z(j)*k(h)^alpha + (1-delta)*k(h) - kdecrule(h,j);
    end
end
figure
plot(k,cdecrule)
xlabel('k_t')
ylabel('c_t')
print -djpeg decrulec.jpg
% Simulation
% Aribrary starting point
enostate = round(gk/2);
ksim(1) = k(enostate);
% Draw a sequence of 100 random variables and use the optimal decision rule
for i = 1 : 100
    draw = rand;
```

```
if draw < prh
        exostate = 1;
    else
        exostate = 2;
    end
    kprimegrid = kgridrule(enostate,exostate);
    kprime = k(kprimegrid);
    zsim(i) = z(exostate);
   ksim(i+1) = kprime;
    ysim(i) = z(exostate)*ksim(i)^alpha;
    isim(i) = ksim(i+1) - (1-delta)*ksim(i);
    csim(i) = ysim(i) - isim(i);
end
figure
plot(1:100,ysim/mean(ysim),1:100,csim/mean(csim),1:100,isim/mean(isim),1:100,ksim(1:100)/mean(ksim(1:100))
legend('y','c','i','k')
title('Simulation 100 draws, devations from mean')
print -djpeg simulation.jpg
```