



CLEVELAND STATE UNIVERSITY

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## Optimal Control Homework 11

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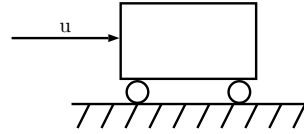
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We found  $E(J)$  for a stochastic LQR problem with cost function  $J$ . Define an LQR problem (maybe an RC circuit, or a two-state Newtonian system, or a linearized version of your project problem). Implement stochastic LQR control. Run your system many times (maybe 100 times or so) and verify that the numerical value for  $E(J)$  matches the analytical expression for  $E(J)$ .

A simple cart system is chosen for this problem. the dynamic equations for this system are :



$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\dot{x} = Ax + Bu + w$$

where  $x = [x_1, x_2]^T$  is the state vector.  $x_1$  and  $x_2$  are the position and velocity respectively. The control  $u$  is the acceleration. The noise vector  $w = [w_1, w_2]^T$  is chosen to be white noise with mean of zero and covariance matrix  $v$

$$w \sim (0, v)$$

Also the initial condition  $x(0)$  is chosen to be white noise with the covariance matrix  $p_0$ :

$$x(0) \sim (0, p_0)$$

where  $v$  and  $p_0$  are chosen to be:

$$v = \begin{bmatrix} 10^{-5} & 0 \\ 0 & 10^{-5} \end{bmatrix}$$

$$p_0 = \begin{bmatrix} 0.1 & 0 \\ 0 & 0.1 \end{bmatrix}$$

in order to simulate the system the covariance matrix  $v$  is scaled by the time step  $\Delta t$ :

$$\frac{v}{\Delta t}$$

The following code has been written for this problem:

```

1  clc
2  clear
3  close all
4
5
6  %state equations
7  A=[0 1;0 0];
8  B=[0;1];
9
10 %Cost function
11 Q=[2 0;0 2];
12 R=1;
13 invR=1; %R^-1
14 H=[10 0;0 0];
15
16
17 dt = 0.001; % Integration step size
18 tf = 10; % Simulation length
19 t=0:dt:tf;
20 x0 = [0;0]; % Initial state
21 N = round(tf/dt) + 1; %Number of steps
22

```

```

23 S=zeros(2,2,N);
24 S(:,:,N)=H; %final conditions for S
25 F(N,:)=invR*B'*S(:,:,N); % F=-R^-1*B'*S
26
27 %Backwards integration in time
28 for i = 1 : N-1
29 Sdot=-Q+S(:,:,N-i+1)*B*invR*B'*S(:,:,N-i+1)-S(:,:,N-i+1)*A-A'*S(:,:,N-i+1); % Riccati ...
    equation
30 S(:,:,N-i)=S(:,:,N-i+1)-Sdot*dt;
31 F(N-i,:)=invR*B'*S(:,:,N-i);
32 end
33
34 %plot
35 plot(t,permute(S(1,1,:),[3,2,1]),t,permute(S(2,1,:),[3,2,1]),t,permute(S(2,2,:),[3,2,1]))
36 xlabel('time(s)')
37 ylabel('S')
38 legend('s1','s2','s3')
39
40 figure
41 plot(t,F(:,1),t,F(:,2))
42 xlabel('time(s)')
43 ylabel('F')
44 legend('f1','f2')
45
46
47 runs=100; %number of runs
48 J=zeros(runs,1);
49
50 for k=1:runs;
51 disp(['#',num2str(k)]) %display run number
52 v=[1e-5,0;0,1e-5]; %covariance matrix
53 %v=zeros(2);
54 p0=[0.1,0;0,0.1]; %covariance matrix for initial conditions
55 x=x0+(p0).^0.5*randn(2,1);
56 %x=x0;
57 xArray = zeros(2, N);
58 uArray = zeros(1, N);
59 wArray = zeros(2, N);
60 %System simulation
61 for i=1:N
62 w=(v/dt).^0.5*randn(2,1);
63 u=F(i,:)*x;
64 xdot=A*x+B*u+w;
65 xArray(:,i)=x;
66 uArray(i)=u;
67 wArray(:,i)=w;
68 x=x+xdot*dt;
69 end
70
71 % compute cost
72 integrand=zeros(1,N-1);
73 for j=1:N-1
74 integrand(j)=xArray(:,j)'*Q*xArray(:,j)+uArray(j)*R*uArray(j);
75 end
76
77 J(k)=xArray(:,end)'*H*xArray(:,end)+trapz(t(1:end-1),integrand);
78
79 end
80
81
82 disp(['Average of numerical cost over ',num2str(k),' runs'])
83 mean(J)
84
85 %compute analytical cost
86 integrand2=zeros(1,N);
87 for j=1:N
88 integrand2(j)=trace(v*S(:,:,j));
89 end

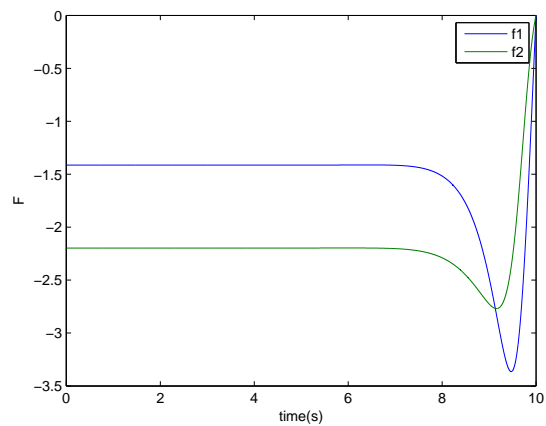
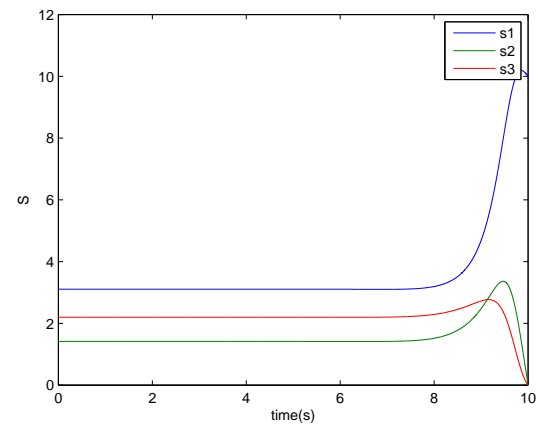
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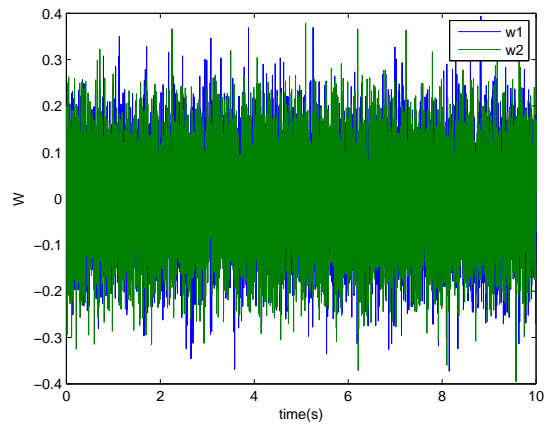
90
91 disp('Analytical cost')
92 J2=trace(S(:, :, 1)*p0)+trapz(t(2:end),integrand2(2:end))
93
94
95 %plot
96 figure
97 plot(t,xArray(1,:),t,xArray(2,:))
98 xlabel('time(s)')
99 ylabel('x')
100 legend('x1','x2')
101
102
103 figure
104 plot(t,uArray)
105 xlabel('time(s)')
106 ylabel('u')
107
108 figure
109 plot(t,wArray(1,:),t,wArray(2,:))
110 xlabel('time(s)')
111 ylabel('W')
112 legend('w1','w2')

```

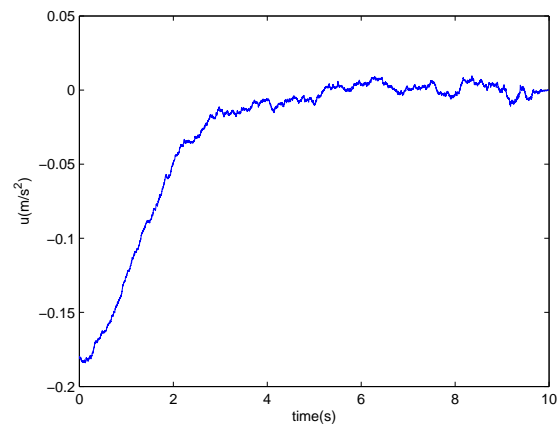
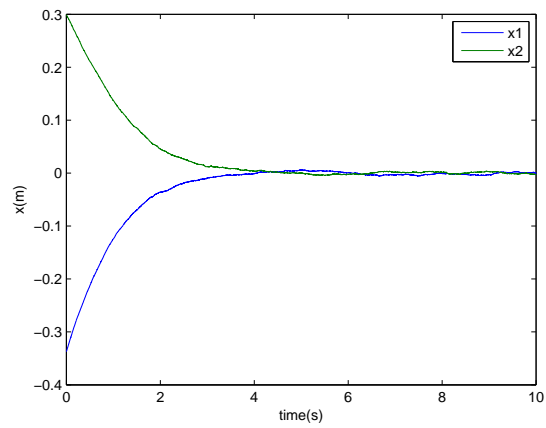
The following figures show the the components of the  $S$  matrix and the  $F$  vector:



the following figure shows the components of the noise vector  $w$ :



the flowing figures show the optimal state trajectory and control for a sample run:



The final results are as follows:

```

1   Average of numerical cost over 100 runs
2   ans =
3       0.54327
4   Analytical cost
5   J2 =
6       0.53108

```