

```
clear all
sympref('FloatingPointOutput', false)
```

```
ans = logical
0
```

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Problem 2

Part A

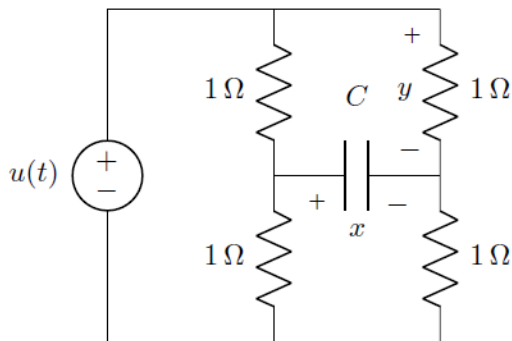


Figure 1: Capacitive bridge.

```
syms i_1 i_2 i_3 u x x_dot y C
%Loop 1 equation
l1eqn = i_1 + i_2 == u
```

```
l1eqn = i_1 + i_2 = u
```

%Node 1 equation

$$n1eqn = i_1 - i_2 == C \cdot \dot{x}$$

$$n1eqn = i_1 - i_2 = C \dot{x}$$

%Loop 2 equation

$$l2eqn = -i_1 + y - x == 0$$

$$l2eqn = -x + y - i_1 = 0$$

%Node 2 equation

$$n2eqn = -y + i_3 == C \cdot \dot{x}$$

$$n2eqn = -y + i_3 = C \dot{x}$$

%Loop 2 equation

$$l3eqn = -i_2 + x + i_3 == 0$$

$$l3eqn = x - i_2 + i_3 = 0$$

$$eqns = [l1eqn; n1eqn; l2eqn; n2eqn; l3eqn]$$

eqns =

$$\begin{bmatrix} i_1 + i_2 = u \\ i_1 - i_2 = C \dot{x} \\ -x + y - i_1 = 0 \\ -y + i_3 = C \dot{x} \\ x - i_2 + i_3 = 0 \end{bmatrix}$$

$$state_eqn = 0 == \text{eliminate}(eqns, [i_1 \ i_2 \ i_3 \ y]);$$

$$state_eqn = \text{isolate}(state_eqn, \dot{x})$$

state_eqn =

$$\dot{x} = -\frac{x}{C}$$

$$output_eqn = 0 == \text{eliminate}(eqns, [i_1 \ i_2 \ i_3 \ \dot{x}])$$

$$output_eqn = 0 = x - 2y + u$$

$$output_eqn = \text{isolate}(output_eqn, y)$$

output_eqn =

$$y = \frac{x}{2} + \frac{u}{2}$$

Part B

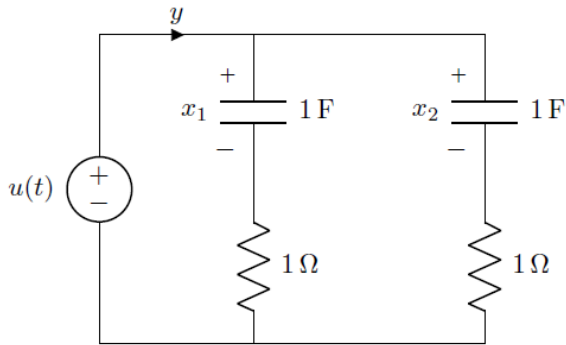


Figure 2: Parallel circuit.

```
syms x_1 x_2 x_dot_1 x_dot_2
```

```
%Loop 1 equation
```

```
l1eqn = x_1 + i_1 - u == 0
```

```
l1eqn = x_1 - u + i_1 = 0
```

```
%Loop 2 equation
```

```
l2eqn = x_2 + i_2 - u == 0
```

```
l2eqn = x_2 - u + i_2 = 0
```

```
%Node 1 equation
```

```
n1eqn = y == x_dot_1 + x_dot_2
```

```
n1eqn = y = x_dot_1 + x_dot_2
```

```
%Node 2 equation
```

```
n2eqn = i_1 == x_dot_1
```

```
n2eqn = i_1 = x_dot_1
```

```
%Node 3 equation
```

```
n3eqn = i_2 == x_dot_2
```

```
n3eqn = i_2 = x_dot_2
```

```
eqns = [l1eqn;l2eqn;n1eqn;n2eqn;n3eqn]
```

```
eqns =
```

$$\begin{bmatrix} x_1 - u + i_1 = 0 \\ x_2 - u + i_2 = 0 \\ y = \dot{x}_1 + \dot{x}_2 \\ i_1 = \dot{x}_1 \\ i_2 = \dot{x}_2 \end{bmatrix}$$

```
solved_state_eqns = eliminate(eqns, [i_1 i_2 y])
```

```
solved_state_eqns = [-x_2 - \dot{x}_2 + u, x_1 - x_2 + \dot{x}_1 - \dot{x}_2]
```

```
solved_state_eqns = [0;0] == [solved_state_eqns(1);solved_state_eqns(2)];
solved_state_eqns = [solved_state_eqns(1)-solved_state_eqns(2);solved_state_eqns(1)];
solved_state_eqns = [isolate(solved_state_eqns(1), x_dot_1);isolate(solved_state_eqns(2), x_dot_2)]
```

```
solved_state_eqns =
```

$$\begin{bmatrix} \dot{x}_1 = -x_1 + u \\ \dot{x}_2 = -x_2 + u \end{bmatrix}$$

```
solved_output_eqn = 0 == eliminate(eqns, [i_1 i_2 x_dot_1 x_dot_2]);
solved_output_eqn = isolate(solved_output_eqn, y)
```

```
solved_output_eqn = y = -x_1 - x_2 + 2 u
```

Problem 3

Part A

State space model of spring mass system:

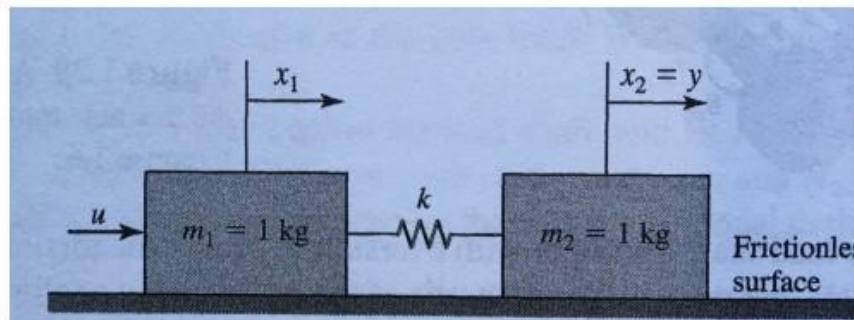


Figure 3: Two-mass-spring system. Assume a frictionless surface. Source: [Za] Fig. 1.31.

Here, the state space model is $\dot{x} = Ax + Bu$ where $x = [x_1 \ x_2 \ \dot{x}_1 \ \dot{x}_2]$

```
m1 = 1;
m2 = 1;
k = 1;
```

```
A = [0      0      1      0
```

```

0      0      0      1
-k/m1  k/m1  0      0
k/m2  -k/m2  0      0];

```

```

B = [ 0
      0
      1/m1
      0 ];

```

```

C = [ 0 1 0 0];

```

```

D = 0;

```

```

sys = ss(A,B,C,D);

```

Controllability Gramian W_c

$$W_c(t) = \int_0^t e^{A\tau} B B^T e^{A^T \tau} d\tau$$

```

syms Tau t n

```

Symbolic Integrand for W_c

```

Wc_integrand = expm(A*Tau)*(B*B. ')*expm(A. '*Tau);

```

Symbolically Integrate (too large to display)

```

Wc(t) = int(Wc_integrand,Tau, 0, t);

```

Evaluate W_c at the time values $t_1 = 10$, $t_2 = 20$, and $t_3 = 100$ and determine whether or not $W_c(t_1)$, $W_c(t_2)$, $W_c(t_3)$ are convergent.

Via a proof shown in the hand calculations, any real diagonalizable matrix is convergent if and only if all of its eigenvalues are less than 1. The controllability grammians of this system are all real and diagonalizable and are shown to each have at least one eigenvalue greater than 1.

```

t1 = 10;
t2 = 20;
t3 = 100;
Wc1 = double(Wc(t1))

```

```

Wc1 = 4x4
 84.1477  82.7081  14.3302  10.9210
 82.7081  83.7694  13.9540  10.7948
 14.3302  13.9540   4.1031   1.2504
 10.9210  10.7948   1.2504   3.3960

```

```

Wc2 = double(Wc(t2))

```

```

Wc2 = 4x4

```

672.9142	665.4171	49.9649	50.5351
665.4171	662.9182	49.4649	50.0351
49.9649	49.4649	7.4974	2.4991
50.5351	50.0351	2.4991	7.5044

```
Wc3 = double(Wc(t3))
```

```
Wc3 = 4x4
```

```
104 ×
```

8.3365	8.3327	0.1249	0.1251
8.3327	8.3315	0.1249	0.1251
0.1249	0.1249	0.0037	0.0012
0.1251	0.1251	0.0012	0.0038

t1 = 10 sec

```
% V is the matrix of eigenvectors and D is a diagonal matrix of eigenvalues
[V1, Wc_bar_1] = eig(Wc1);
%Make sure they are ordered from largest to smallest
[eigvals1, ind1] = sort(diag(Wc_bar_1), 'descend');
disp(["Reordered Eigenvalue of Wc1 = " + eigvals1 "Index = " + ind1]);
```

"Reordered Eigenvalue of Wc1 = 170.5095"	"Index = 4"
"Reordered Eigenvalue of Wc1 = 2.458196"	"Index = 3"
"Reordered Eigenvalue of Wc1 = 1.456741"	"Index = 2"
"Reordered Eigenvalue of Wc1 = 0.9917897"	"Index = 1"

```
Wc_bar_s1 = Wc_bar_1(ind1, ind1)
```

```
Wc_bar_s1 = 4x4
```

170.5095	0	0	0
0	2.4582	0	0
0	0	1.4567	0
0	0	0	0.9918

Since the matrix \bar{W}_{c_1} , which is in the eigenbasis of W_{c_1} , has at least one eigenvalue greater than 1,

$\lim_{n \rightarrow \infty} W_{c_1}^n \neq \tilde{0}$. I.e. W_{c_1} does not converge (see hand calculations for proof).

```
%Reordered Eigenvectors
Vs1 = V1(:, ind1)
```

```
Vs1 = 4x4
```

0.7000	0.0540	-0.4444	-0.5565
0.6981	-0.0605	0.5900	0.4011
0.1195	0.6300	-0.4824	0.5967
0.0917	-0.7723	-0.4708	0.4164

t2 = 20 sec

```
% V is the matrix of eigenvectors and D is a diagonal matrix of eigenvalues
[V2, Wc_bar_2] = eig(Wc2);
%Make sure they are ordered from largest to smallest
[eigvals2, ind2] = sort(diag(Wc_bar_2), 'descend');
disp(["Reordered Eigenvalue of Wc2 = " + eigvals2 "Index = " + ind2]);
```

"Reordered Eigenvalue of Wc2 = 1340.8664"	"Index = 4"
"Reordered Eigenvalue of Wc2 = 5.0021324"	"Index = 3"

```
"Reordered Eigenvalue of Wc2 = 2.6068916"    "Index = 2"
"Reordered Eigenvalue of Wc2 = 2.3587624"    "Index = 1"
```

```
Wc_bar_s2 = Wc_bar_2(ind2,ind2)
```

```
Wc_bar_s2 = 4x4
```

```
103 x
    1.3409         0         0         0
         0    0.0050         0         0
         0         0    0.0026         0
         0         0         0    0.0024
```

Since the matrix \bar{W}_{c_2} , which is in the eigenbasis of W_{c_2} , has at least one eigenvalue greater than 1,

$\lim_{n \rightarrow \infty} W_{c_2}^n \neq \tilde{0}$. I.e. W_{c_2} does not converge (see hand calculations for proof).

```
%Reordered Eigenvectors
```

```
Vs2 = V2(:,ind2)
```

```
Vs2 = 4x4
```

```
    0.7078   -0.0006    0.4551    0.5403
    0.7025    0.0017   -0.5345   -0.4699
    0.0527   -0.7181    0.4948   -0.4866
    0.0533    0.6959    0.5122   -0.5005
```

t3 = 100 sec

```
% V is the matrix of eigenvectors and D is a diagonal matrix of eigenvalues
```

```
[V3, Wc_bar_3] = eig(Wc3);
```

```
%Make sure they are ordered from largest to smallest
```

```
[eigvals3,ind3] = sort(diag(Wc_bar_3), 'descend');
```

```
disp(["Reordered Eigenvalue of Wc3 = " + eigvals3 "Index = " + ind3]);
```

```
"Reordered Eigenvalue of Wc3 = 166704.1732"    "Index = 4"
"Reordered Eigenvalue of Wc3 = 25.00875972"    "Index = 3"
"Reordered Eigenvalue of Wc3 = 12.61979239"    "Index = 2"
"Reordered Eigenvalue of Wc3 = 12.36925083"    "Index = 1"
```

```
Wc_bar_s3 = Wc_bar_3(ind3,ind3)
```

```
Wc_bar_s3 = 4x4
```

```
105 x
    1.6670         0         0         0
         0    0.0003         0         0
         0         0    0.0001         0
         0         0         0    0.0001
```

Since the matrix \bar{W}_{c_3} , which is in the eigenbasis of W_{c_3} , has at least one eigenvalue greater than 1,

$\lim_{n \rightarrow \infty} W_{c_3}^n \neq \tilde{0}$. I.e. W_{c_3} does not converge (see hand calculations for proof).

```
%Reordered Eigenvectors
```

```
Vs3 = V3(:,ind3)
```

```
Vs3 = 4x4
```

```
    0.7071    0.0000    0.4870   -0.5126
    0.7069   -0.0000   -0.5023    0.4979
```

0.0106	0.7080	0.5045	0.4940
0.0106	-0.7062	0.5059	0.4952

Which States Are Most Controllable?

The degree of controllability of any state in this system is determined, roughly, by how much of the eigenvector associated with the dominant eigenvalue lies in the direction of that state. Since for each of these controllability Gramian matrices, there is one eigenvalue that is orders of magnitude larger than the others, this is a good criteria.

For $t_1 = 10s$, the dominant eigenvector is:

```
Vs1(:,1)
```

```
ans = 4×1
    0.7000
    0.6981
    0.1195
    0.0917
```

Which has the associated eigenvalue:

```
disp(eigvals1(1))
```

```
170.5095
```

The eigenvalue points largely in the directions of x_1 and x_2 , but is slightly more in x_1 than it is in x_2 . These states are the most controllable.

For $t_2 = 20s$, the dominant eigenvector is:

```
Vs2(:,1)
```

```
ans = 4×1
    0.7078
    0.7025
    0.0527
    0.0533
```

Which has the associated eigenvalue:

```
disp(eigvals2(1))
```

```
1.3409e+03
```

The eigenvalue points largely in the directions of x_1 and x_2 , but is slightly more in x_1 than it is in x_2 . These states are the most controllable.

For $t_3 = 100s$, the dominant eigenvector is:


```
Vs3(:,1)
```

```
ans = 4x1
    0.7071
    0.7069
    0.0106
    0.0106
```

Which has the associated eigenvalue:

```
disp(eigvals3(1))
```

```
1.6670e+05
```

The eigenvalue points largely in the directions of x_1 and x_2 , but is slightly more in x_1 than it is in x_2 . These states are the most controllable.

The clear trend is that the position states tend to be much more controllable than the velocity states and x_1 , where the input most directly affects the mass, is slightly more controllable than x_2 .

Observability Gramian W_o

$$W_o(t) = \int_0^t e^{A^T \tau} C^T C e^{A \tau} d\tau$$

Symbolic Integrand for W_o

```
Wo_integrand = expm(A'*Tau)*(C'*C)*expm(A*Tau);
```

Symbolically Integrate (too large to display)

```
Wo(t) = int(Wo_integrand,Tau, 0, t);
```

Evaluate W_o at the time values $t_1 = 10$, $t_2 = 20$, and $t_3 = 100$ and determine whether or not $W_o(t_1)$, $W_o(t_2)$, $W_o(t_3)$ are convergent.

Via a proof shown in the hand calculations, any real diagonalizable matrix is convergent if and only if all of its eigenvalues are less than 1. The observability grammians of this system are all real and diagonalizable and are shown to each have at least one eigenvalue greater than 1.

```
Wo1 = double(Wo(t1))
```

```
Wo1 = 4x4
    3.3960    1.2504    10.7948    10.9210
    1.2504    4.1031    13.9540    14.3302
    10.7948    13.9540    83.7694    82.7081
    10.9210    14.3302    82.7081    84.1477
```

```
Wo2 = double(Wo(t2))
```

```
Wo2 = 4x4
    7.5044    2.4991    50.0351    50.5351
    2.4991    7.4974    49.4649    49.9649
    50.0351    49.4649    662.9182    665.4171
    50.5351    49.9649    665.4171    672.9142
```

```
Wo3 = double(Wo(t3))
```

```
Wo3 = 4x4
10^4 x
    0.0038    0.0012    0.1251    0.1251
    0.0012    0.0037    0.1249    0.1249
    0.1251    0.1249    8.3315    8.3327
    0.1251    0.1249    8.3327    8.3365
```

t1 = 10 sec

```
% V is the matrix of eigenvectors and D is a diagonal matrix of eigenvalues
[V1, Wo_bar_1] = eig(Wo1);
%Make sure they are ordered from largest to smallest
[eigvals1,ind1] = sort(diag(Wo_bar_1), 'descend');
disp(["Reordered Eigenvalue of Wo1 = " + eigvals1 "Index = " + ind1]);
```

```
"Reordered Eigenvalue of Wo1 = 170.5095"    "Index = 4"
"Reordered Eigenvalue of Wo1 = 2.458196"    "Index = 3"
"Reordered Eigenvalue of Wo1 = 1.456741"    "Index = 2"
"Reordered Eigenvalue of Wo1 = 0.9917897"    "Index = 1"
```

```
Wo_bar_s1 = Wo_bar_1(ind1,ind1)
```

```
Wo_bar_s1 = 4x4
    170.5095         0         0         0
         0     2.4582         0         0
         0         0     1.4567         0
         0         0         0     0.9918
```

Since the matrix \bar{W}_{o_1} , which is in the eigenbasis of W_{o_1} , has at least one eigenvalue greater than 1,

$\lim_{n \rightarrow \infty} W_{o_1}^n \neq \tilde{0}$. i.e. W_{o_1} does not converge (see hand calculations for proof).

```
%Reordered Eigenvectors
Vs1 = V1(:,ind1)
```

```
Vs1 = 4x4
    0.0917    0.7723   -0.4708    0.4164
    0.1195   -0.6300   -0.4824    0.5967
    0.6981    0.0605    0.5900    0.4011
    0.7000   -0.0540   -0.4444   -0.5565
```

t2 = 20 sec

```
% V is the matrix of eigenvectors and D is a diagonal matrix of eigenvalues
[V2, Wo_bar_2] = eig(Wo2);
%Make sure they are ordered from largest to smallest
[eigvals2,ind2] = sort(diag(Wo_bar_2), 'descend');
disp(["Reordered Eigenvalue of Wo2 = " + eigvals2 "Index = " + ind2]);
```

```

"Reordered Eigenvalue of Wo2 = 1340.8664"    "Index = 4"
"Reordered Eigenvalue of Wo2 = 5.0021324"    "Index = 3"
"Reordered Eigenvalue of Wo2 = 2.6068916"    "Index = 2"
"Reordered Eigenvalue of Wo2 = 2.3587624"    "Index = 1"

```

```
Wo_bar_s2 = Wo_bar_2(ind2,ind2)
```

```
Wo_bar_s2 = 4x4
```

```

103 x
    1.3409         0         0         0
         0    0.0050         0         0
         0         0    0.0026         0
         0         0         0    0.0024

```

Since the matrix \bar{W}_{o_2} , which is in the eigenbasis of W_{o_2} , has at least one eigenvalue greater than 1,

$\lim_{n \rightarrow \infty} W_{o_2}^n \neq \tilde{0}$. I.e. W_{o_2} does not converge (see hand calculations for proof).

```
%Reordered Eigenvectors
```

```
Vs2 = V2(:,ind2)
```

```
Vs2 = 4x4
```

```

    0.0533    0.6959   -0.5122    0.5005
    0.0527   -0.7181   -0.4948    0.4866
    0.7025    0.0017    0.5345    0.4699
    0.7078   -0.0006   -0.4551   -0.5403

```

t3 = 100 sec

```
% V is the matrix of eigenvectors and D is a diagonal matrix of eigenvalues
```

```
[V3, Wo_bar_3] = eig(Wo3);
```

```
%Make sure they are ordered from largest to smallest
```

```
[eigvals3,ind3] = sort(diag(Wo_bar_3), 'descend');
```

```
disp(["Reordered Eigenvalue of Wo3 = " + eigvals3 "Index = " + ind3]);
```

```

"Reordered Eigenvalue of Wo3 = 166704.1732"    "Index = 4"
"Reordered Eigenvalue of Wo3 = 25.00875972"    "Index = 3"
"Reordered Eigenvalue of Wo3 = 12.61979239"    "Index = 2"
"Reordered Eigenvalue of Wo3 = 12.36925083"    "Index = 1"

```

```
Wo_bar_s3 = Wo_bar_3(ind3,ind3)
```

```
Wo_bar_s3 = 4x4
```

```

105 x
    1.6670         0         0         0
         0    0.0003         0         0
         0         0    0.0001         0
         0         0         0    0.0001

```

Since the matrix \bar{W}_{o_3} , which is in the eigenbasis of W_{o_3} , has at least one eigenvalue greater than 1,

$\lim_{n \rightarrow \infty} W_{o_3}^n \neq \tilde{0}$. I.e. W_{o_3} does not converge (see hand calculations for proof).

```
%Reordered Eigenvectors
```

```
Vs3 = V3(:,ind3)
```

```
Vs3 = 4x4
    0.0106    0.7062   -0.5059   -0.4952
    0.0106   -0.7080   -0.5045   -0.4940
    0.7069    0.0000    0.5023   -0.4979
    0.7071   -0.0000   -0.4870    0.5126
```

Which States Are Most Observable?

The degree of observability of any state in this system is determined, roughly, by how much of the eigenvector associated with the dominant eigenvalue lies in the direction of that state. Since for each of these observability Gramian matrices, there is one eigenvalue that is orders of magnitude larger than the others, this is a good criteria.

For $t_1 = 10s$, the dominant eigenvector is:

```
Vs1(:,1)
```

```
ans = 4x1
    0.0917
    0.1195
    0.6981
    0.7000
```

Which has the associated eigenvalue:

```
disp(eigvals1(1))
```

```
170.5095
```

The eigenvalue points largely in the directions of \dot{x}_1 and \dot{x}_2 , but is slightly more in \dot{x}_2 than it is in \dot{x}_1 . These states are the most observable.

For $t_2 = 20s$, the dominant eigenvector is:

```
Vs2(:,1)
```

```
ans = 4x1
    0.0533
    0.0527
    0.7025
    0.7078
```

Which has the associated eigenvalue:

```
disp(eigvals2(1))
```

```
1.3409e+03
```

The eigenvalue points largely in the directions of \dot{x}_1 and \dot{x}_2 , but is slightly more in \dot{x}_2 than it is in \dot{x}_1 . These states are the most observable.

For $t_3 = 100s$, the dominant eigenvector is:

```
Vs3(:,1)
```

```
ans = 4x1
    0.0106
    0.0106
    0.7069
    0.7071
```

Which has the associated eigenvalue:

```
disp(eigvals3(1))
```

```
1.6670e+05
```

The eigenvalue points largely in the directions of \dot{x}_1 and \dot{x}_2 , but is slightly more in \dot{x}_2 than it is in \dot{x}_1 . These states are the most observable.

The clear trend is that the velocity states tend to be much more observable than the position states and \dot{x}_2 , is slightly more observable than \dot{x}_1 .

Part B

Three Linear Quadratic Regulators

Linear quadratic regulators are used to choose stabilizing k matrices for full state feedback. Bryson's rule for various values of the maximum acceptable state and input are utilized to get three different K matrices.

```
%Bryson's rule for xmax = 3 meters, x_dot_max = 10 m/s, and Umax = 2 newtons - because why
%not?
Q1 = [1/3 0 0 0; 0 1/3 0 0; 0 0 1/10 0; 0 0 0 1/10];
R1 = 1/2;

[K1,S1,e1] = lqr(A,B,Q1,R1);
K1
```

```
K1 = 1x4
    1.1488    0.0059    1.5804    0.9373
```

```
%Bryson's rule for xmax = 1 meters, x_dot_max = 15 m/s, and Umax = 20 newtons
Q2 = [1/1 0 0 0; 0 1/1 0 0; 0 0 1/15 0; 0 0 0 1/15];
R2 = 1/20;

[K2,S2,e2] = lqr(A,B,Q2,R2);
K2
```

```
K2 = 1x4
    6.4807   -0.1562    3.7809    4.5822
```

```
%Bryson's rule for xmax = .05 meters, x_dot_max =2 m/s, and Umax = 3 newtons
Q3 = [1/.05 0 0 0; 0 1/.05 0 0; 0 0 1/2 0; 0 0 0 1/2];
R3 = 1/3;

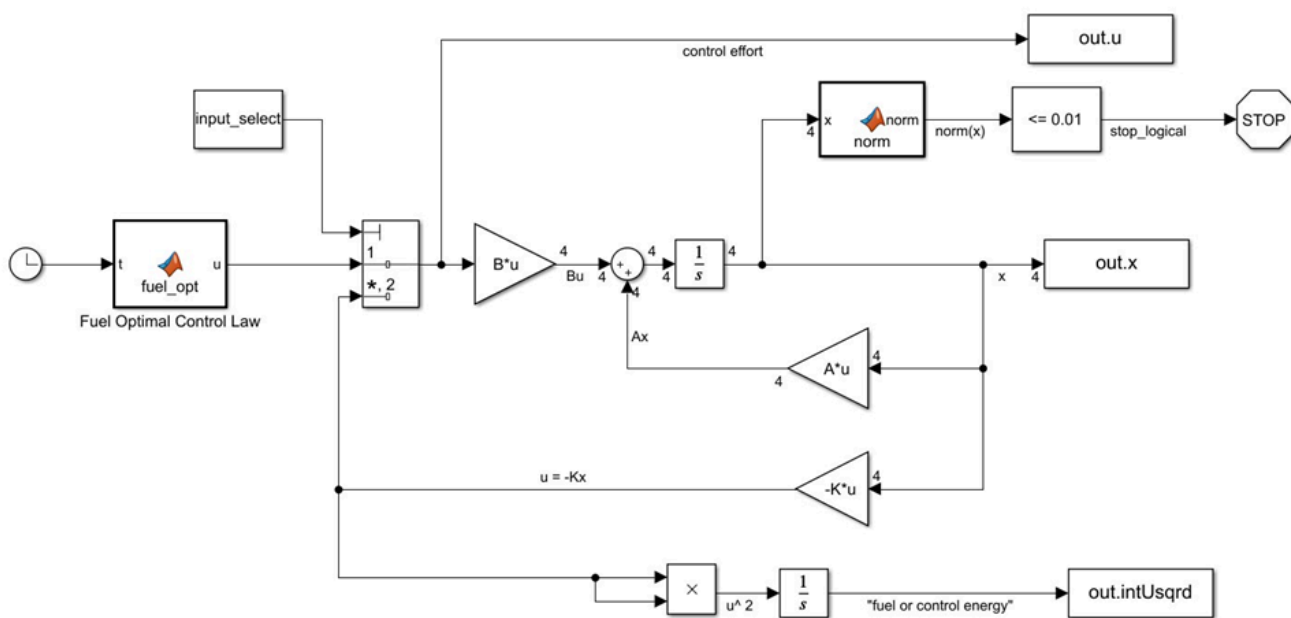
[K3,S3,e3] = lqr(A,B,Q3,R3);
K3
```

```
K3 = 1x4
    10.6433    0.3111    4.7735    7.8756
```

Simulate the Full State Feedback Controllers from $\tilde{x}_0 = [1 \ 0 \ 2 \ 0]$

The full state feedback state space model was implemented in Simulink to allow for easy calculation of integrated controller fuel and determination of steady state from the euclidean norm of the state vector. The final simulation time is the time when the state vector enters a unit hypersphere of radius 0.01.

$$\dot{x} = (A - BK)x$$



Run 'initial' simulations and compare the energy cost and time to steady state

```
x0 = [1 0 2 0]'; %Initial state vector
tstop = 25;
input_select = 2; %Full state feedback input NOT fuel optimal control input
t_end = 3; %Not used now but need to initialize the variable so it's not undefined
Wc_sim = double(Wc(5)); %Grammian for 5 seconds
```

LQR 1

```
K = K1;
out1 = sim('FSF');
x1 = out1.x;
```

```
tsim1 = out1.tout;  
energy1 = out1.intUsqrd(end);  
ss_time1 = tsim1(end);
```

LQR 2

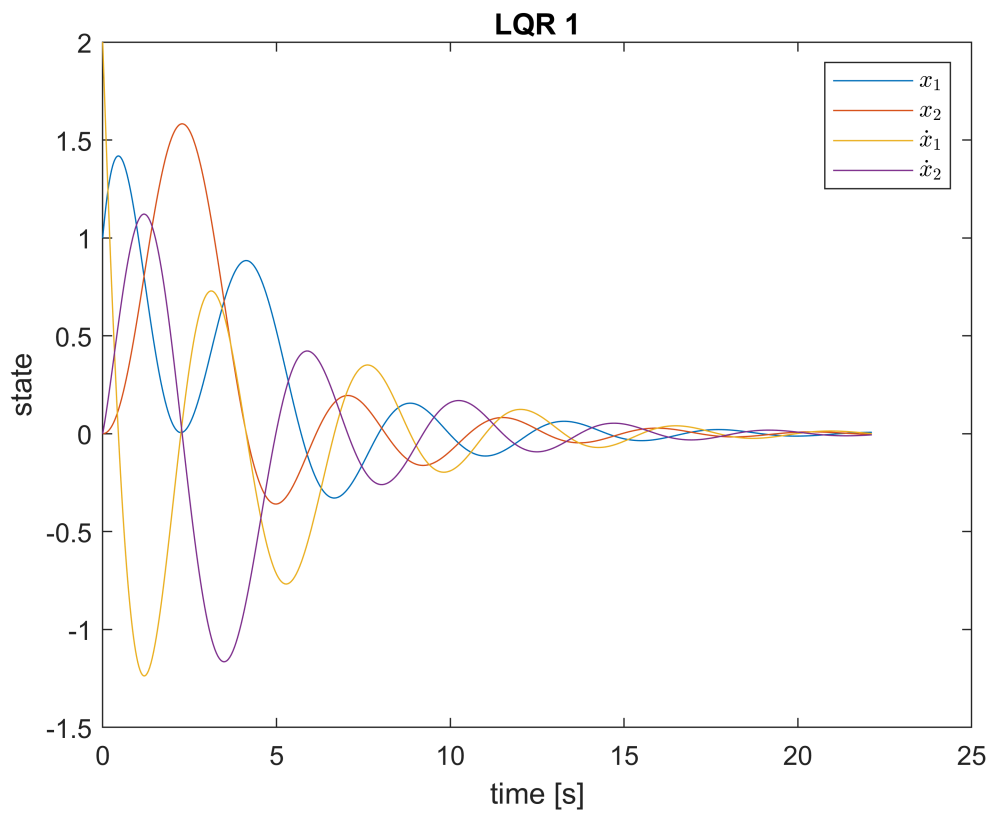
```
K = K2;  
out2 = sim('FSF');  
x2 = out2.x;  
tsim2 = out2.tout;  
energy2 = out2.intUsqrd(end);  
ss_time2 = tsim2(end);
```

LQR3

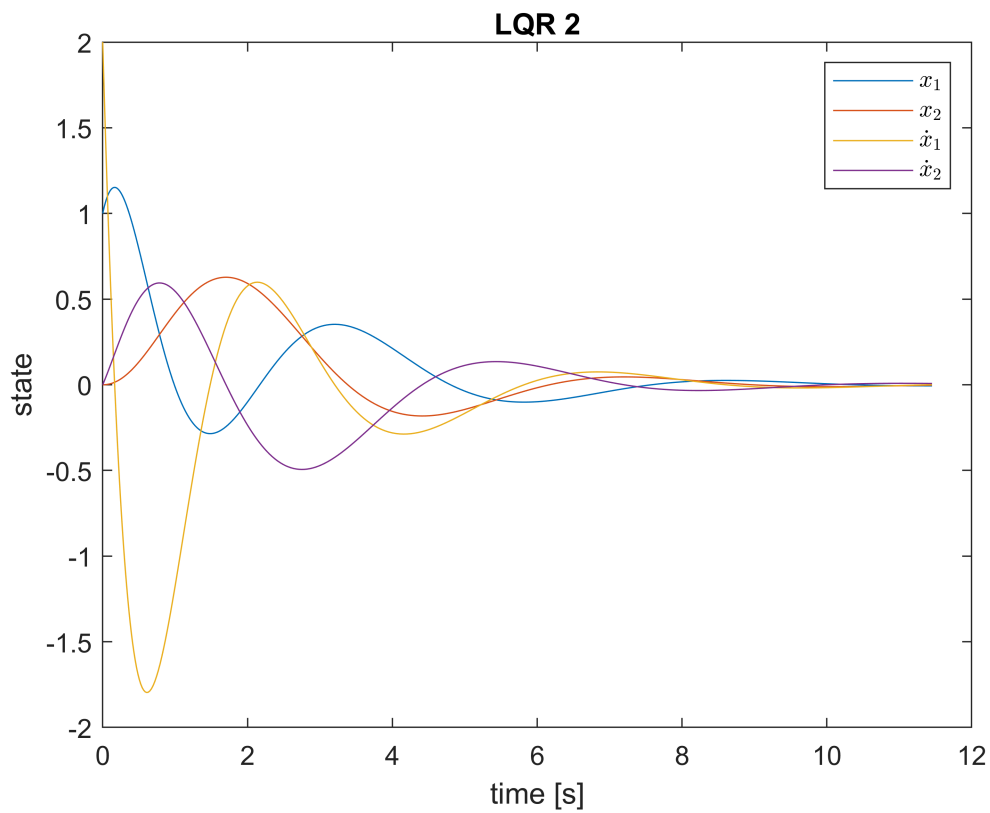
```
K = K3;  
out3 = sim('FSF');  
x3 = out3.x;  
tsim3 = out3.tout;  
energy3 = out3.intUsqrd(end);  
ss_time3 = tsim3(end);
```

Plot results

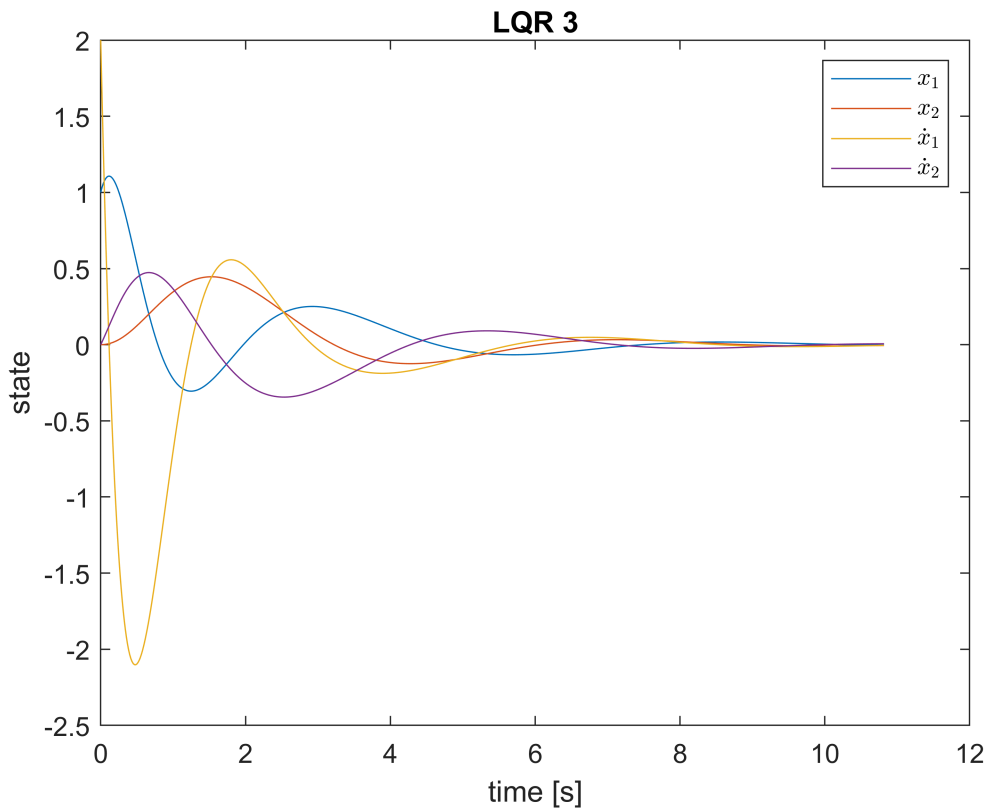
```
figure  
plot(tsim1,x1)  
legend('$x_1$', '$x_2$', '$\dot{x}_1$', '$\dot{x}_2$', 'interpreter', 'latex')  
xlabel('time [s]')  
ylabel('state')  
title('LQR 1')
```



```
figure
plot(tsim2,x2)
legend('$x_1$', '$x_2$', '$\dot{x}_1$', '$\dot{x}_2$', 'interpreter', 'latex')
xlabel('time [s]')
ylabel('state')
title('LQR 2')
```

```
figure
plot(tsim3,x3)
legend('$x_1$', '$x_2$', '$\dot{x}_1$', '$\dot{x}_2$', 'interpreter', 'latex')
xlabel('time [s]')
ylabel('state')
title('LQR 3')
```



Discussion

Controller 1 utilized the least energy but took the most amount of time. Controller 3 utilized the most energy but took the least amount of time. Controller 2 landed between those two scenarios.

```
disp("Energy cost controller 1 was " + num2str(energy1,2) + ".")
```

Energy cost controller 1 was 6.9.

```
disp("Time to steady state controller 1 was " + num2str(ss_time1,3) + " seconds.")
```

Time to steady state controller 1 was 22.1 seconds.

```
disp("Energy cost controller 2 was " + num2str(energy2,3) + ".")
```

Energy cost controller 2 was 31.2.

```
disp("Time to steady state controller 2 was " + num2str(ss_time2,3) + " seconds.")
```

Time to steady state controller 2 was 11.5 seconds.

```
disp("Energy cost controller 3 was " + num2str(energy3,3) + ".")
```

Energy cost controller 3 was 51.1.

```
disp("Time to steady state controller 3 was " + num2str(ss_time3,3) + " seconds.")
```

Time to steady state controller 3 was 10.8 seconds.

Part C

Control law is now the open loop law

$$u(t) = -B^T e^{A^T(t_1-t)} W_c^{-1}(t_1) [e^{At} x_0 - x_1]$$

Instead of the closed loop law

$$u(t) = -Kx(t)$$

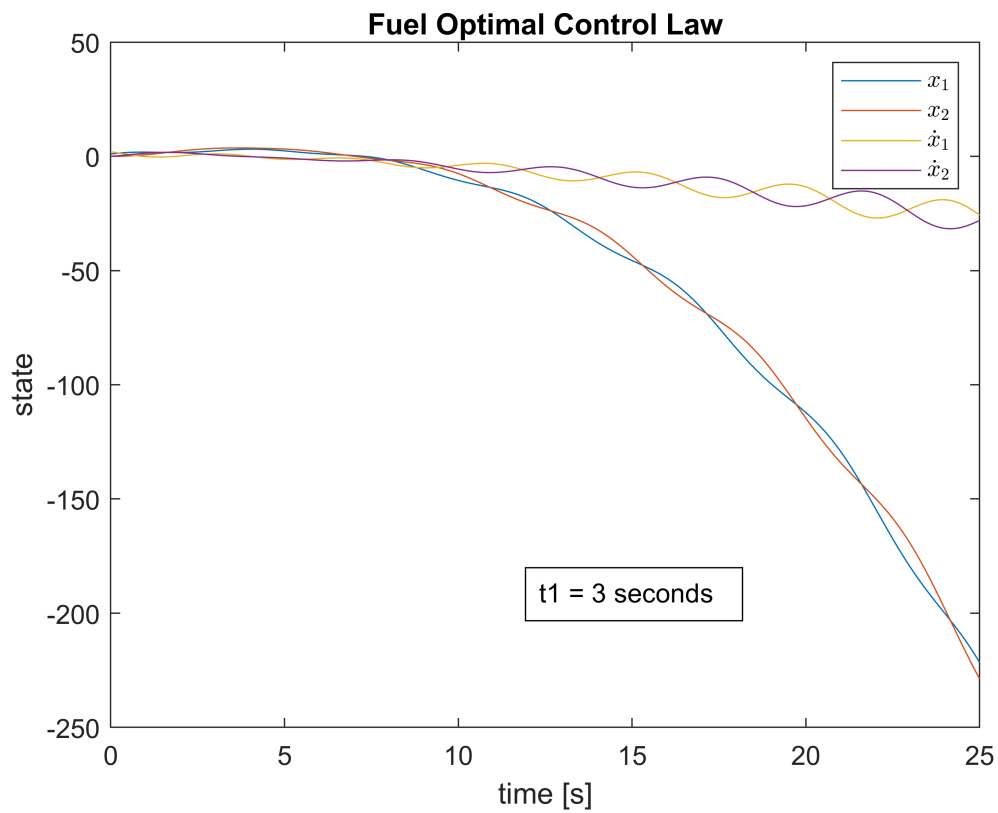
Simulate Fuel Optimal Control Law

```
x0 = [1 0 2 0]'; %Initial state vector
tstop = 25;
input_select = 1; %Fuel optimal control law
%t1 from the control law (named differently because "t1" is already used elsewhere)
t_end = 3; %Not used now but need to initialize the variable so it's not undefined
Wc_sim = double(Wc(5)); %Grammian for 5 seconds

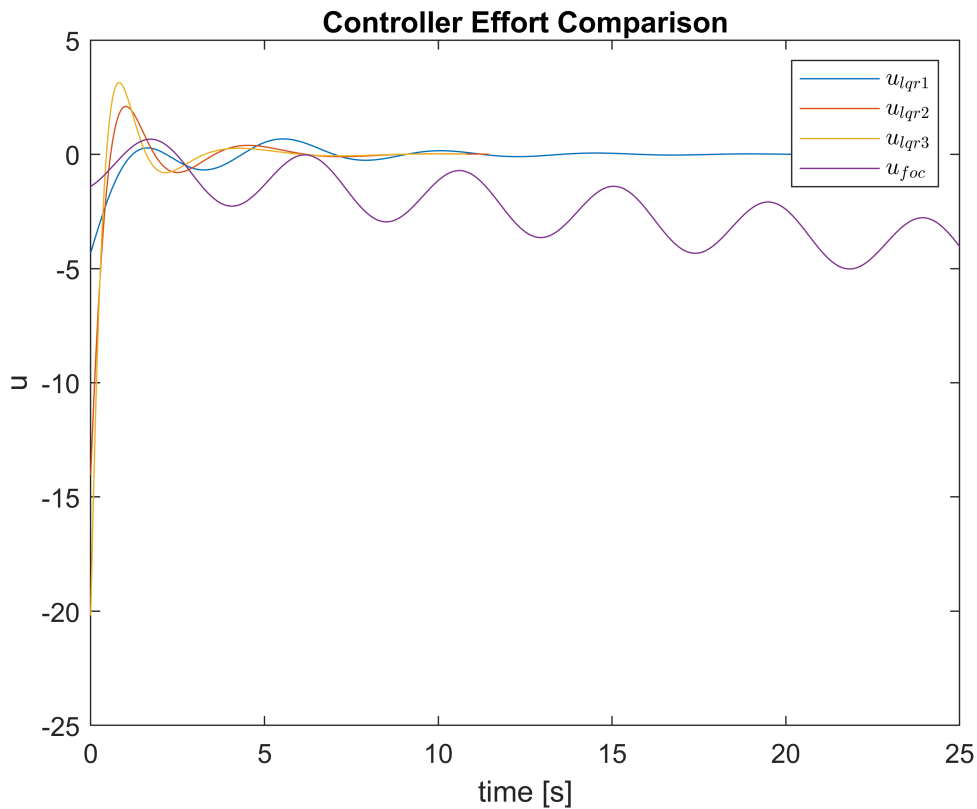
out4 = sim('FSF');
x4 = out4.x;
tsim4 = out4.tout;
energy4 = out4.intUsqrd(end);
ss_time4 = tsim4(end);
```

Plot Results

```
figure
plot(tsim4,x4)
legend('$x_1$', '$x_2$', '$\dot{x}_1$', '$\dot{x}_2$', 'interpreter', 'latex')
xlabel('time [s]')
ylabel('state')
title('Fuel Optimal Control Law')
annotation('textbox', [0.5, 0.2, 0.1, 0.1], 'string', ['t1 = ' num2str(t_end) ' seconds'])
```



```
figure
plot(tsim1,out1.u,tsim2,out2.u,tsim3,out3.u,tsim4,out4.u)
legend('$u_{lqr1}$','$u_{lqr2}$', '$u_{lqr3}$','$u_{foc}$','interpreter','latex')
xlabel('time [s]')
ylabel('u')
title('Controller Effort Comparison')
```



```
% xlim([0 10])
```

Discussion

Since the system is not open loop stable, the fuel optimal control law $u(t)$ does not take the state to zero let alone within time t_1 in an energy minimizing manner - the system is actually unstable. The controller effort is plotted over the simulation time interval to see where the unstable controller's effort diverges from that of the three stable controllers.

```
disp("Energy cost controller 4 was " + num2str(energy4,3) + ".")
```

```
Energy cost controller 4 was 2.91e+07.
```

The energy usage up to the end of the simulation was very large in comparison to the usage of the other controllers - this is due to the instability.