PROBLEM 1

The necessary and sufficient conditions for the existence of unknown input observer (UIO) for a plant model of the following form will be derived.

$$\dot{x} = Ax + B_1 u_1 + B_2 u_2$$

$$y = Cx + D_1 u_1 + D_2 u_3$$
(1)

In equation 1, u_2 is the unknown input and u_3 is the output disturbance. The dimensions of the states, outputs, and inputs are as follows. There are n states, m_1 known inputs, m_2 unknown inputs, p outputs, and r output disturbances. To begin the derivation, we can first represent x as:

$$x = x - M(Cx + D_1u_1) + M(Cx + D_1u_1)$$
(2)

Substituting the output equation from equation 1 into 2 yields:

$$x = x - M(Cx + D_1u_1) + M(y - D_2u_3)$$

$$x = (I - MC)x - MD_1u_1 + My - MD_2u_3$$
(3)

The first condition for the existence of the UIO can be found by selecting M such that the effect of u_3 is the zero matrix. Therefore the first necessary and sufficient condition is:

$$MD_2 = 0_{n \times r} \tag{4}$$

Applying the above condition to equation 3 gives the following:

$$x = (I - MC)x - MD_1u_1 + My (5)$$

Next, the variable z will be defined as, z = (I - MC)x. Then equation 5 becomes:

$$x = z - MD_1 u_1 + M y \tag{6}$$

It should be noted that the state estimate, \hat{x} , can be found by replacing x with \hat{x} in equation 6. Using the definition of z, the dynamics of \dot{z} are given below.

$$\dot{z} = (I - MC)\dot{x} = (I - MC)(Ax + B_1u_1 + B_2u_2) \tag{7}$$

Combining equations 6 and 7 gives the following dynamics:

$$\dot{z} = (I - MC)(Az - AMD_1u_1 + AMy + B_1u_1 + B_2u_2)$$

$$\dot{z} = (I - MC)[Az + AMy + (B_1 - AMD_1)u_1] + (I - MC)B_2u_2$$
(8)

The second condition for the existence of the UIO can be found by selecting M such that the effect of u_2 is the zero matrix in equation 8. Therefore the second necessary and sufficient condition is:

$$(I - MC)B_2 = 0_{n \times m_2} \tag{9}$$

Applying this second condition to equation 8 and adding an innovation term to close the loop for the UIO, the dynamics for z becomes

$$\dot{z} = (I - MC)[Az + AMy + (B_1 - AMD_1)u_1] + L(y - \hat{y}) \tag{10}$$

The state estimate can then be determined with

$$\hat{x} = z - MD_1 u_1 + M y \tag{11}$$

If M is set to 0, then the Luenberger observer is obtained, as shown below:

$$\dot{z} = Az + B_1 u_1 + L(y - \hat{y}) \tag{12}$$

Combining the two conditions for the existence of the UIO, M can be found by solving the following:

$$M\begin{bmatrix} CB_2 & D \end{bmatrix} = \begin{bmatrix} B_2 & 0_{n \times r} \end{bmatrix} \tag{13}$$

$$M = \begin{bmatrix} B_2 & 0_{n \times r} \end{bmatrix} \begin{bmatrix} CB_2 & D \end{bmatrix}^{\dagger} \tag{14}$$

The matrix M exists if and only if $\begin{bmatrix} CB_2 & D \end{bmatrix}$ is full column rank, and is therefore left invertable. The observation error is defined as $e = x - \hat{x}$. Using equation 11, this is expressed as

$$e = x - (z - MD_1u_1 + My) = x - z + MD_1u_1 - M(Cx + D_1u_1 + D_2u_3)$$
(15)

Applying the first condition (equation 4) eliminates the u_3 term in equation 15. Then the

observation error can be simplified to e = (I - MC)x - z. The error dynamics can be expressed as:

$$\dot{e} = (I - MC)(Ax + B_1u_1 + B_2u_2) - [(I - MC)(Az + AMy + (B_1 - AMD_1)u_1) + L(y - \hat{y})]$$
 (16)

The estimated output is given by $\hat{y} = C\hat{x} + D_1u_1$. Therefore the innovation term, $L(y - \hat{y})$, can be written as:

$$L(y - \hat{y}) = L[(Cx + D_1u_1 + D_2u_3) - (C(z - MD_1u_1 + My) + D_1u_1)]$$
(17)

Substituting in the output equation for *y* into the above equation, then simplifying yields:

$$L(y - \hat{y}) = L[Cx + D_2u_3 - Cz - CMCx - CMD_2u_3]$$
 (18)

Again applying the first condition in equation 4 (eliminating the MD_2 term), and using the definition of the error, equation 18 can be simplified to

$$L(y - \hat{y}) = LC(x - z - MCx) + LD_2 u_3 = LCe + LD_2 u_3$$
 (19)

Applying the second condition (equation 9) to equation 16, it can be simplified to:

$$\dot{e} = (I - MC)(Ax) - [(I - MC)(Az + AM(Cx + D_1u_1 + D_2u_3) - AMD_1u_1) + L(y - \hat{y})]$$
 (20)

Again applying the first condition, the u_3 term can be eliminated. Then simplifying yields:

$$\dot{e} = (I - MC)(Ax) - [(I - MC)(Az + AMCx) + LCe + LD_2u_3]$$

$$\dot{e} = (I - MC)(Ax - Az - AMCx) - LCe - LD_2u_3$$
(21)

Applying the error definition to equation 21, the error dynamics can be simplified to the final following form.

$$\dot{e} = ((I - MC)A - LC)e - LD_2 u_3 \tag{22}$$

It can then be shown that the state observation error is l_{∞} stable if there exists a matrix $P = P^T > 0$ and an observer gain matrix L that satisfies the following matrix inequality

$$\begin{bmatrix} E^T P E - (1 - \alpha) P & E^T P N \\ N^T P E & N^T P N - \alpha I \end{bmatrix} \le 0$$
 (23)

where E = ((I - MC)A - LC), N = -LD, and $0 < \alpha < 1$. If the pair ((I - MC)A, C) is detectable (all observable states are stable), then we can find an observer gain matrix L such that E is Schur stable

HW5 Gabe Colangelo

```
clear
close all
warning off
clc
% Symbolic State and Input Vectors
x_sym
        = sym('x',[6,1],'real');
        = sym('u',[3,1],'real');
u_sym
% Numeric System Parameters
        = 0.5;
m1
        = 0.5;
L1
m2
        = 0.75;
L2
        = 0.75;
Μ
        = 1.5;
        = 9.81;
g
```

Problem 2 - UIO Design

```
% Equilibrium State
          = zeros(6,1);
 % Non-linear System
 f
          = DIPC([],x_sym, u_sym, m1, m2, M, L1, L2, g);
 out
          = x_sym(1:6);
 % Solve for Equilibrium Input
 fsol_opt= optimset('Display','off');
          = @(u)DIPC([],xe, u, m1, m2, M, L1, L2, g);
          = fsolve(f_xe,[0 0 0]',fsol_opt);
 ue
\dot{x} = Ax + B_1 u_1 + B_2 u_2
y = Cx + D_1u_1 + D_2u_3
 % Jacobian Matrices/ Linearized Model about Equilibrium Pair
```

```
= double(subs(jacobian(f,x_sym),[x_sym;u_sym],[xe;ue]))
Α
```

```
A = 6 \times 6
                    0
                                     1.0000
                                                                 0
         0
                    0
                                          0
                                                1.0000
                                                                 0
                                                           1.0000
         0
                    0
                               0
                                          0
                                                     0
              -8.1750
                                          0
                                                     0
                                                                0
              65.4000
                      -29.4300
                                          0
                                                     0
                                                                 0
         0 -32.7000
                        32.7000
                                                                0
                                          0
                                                     0
```

```
= double(subs(jacobian(f,u_sym),[x_sym;u_sym],[xe;ue]))
B1
```

```
B1 = 6 \times 3
            0
                           0
                                         0
                           0
                                         0
            0
            0
                                         0
```

```
0.6667
           -1.3333
   -1.3333
           10.6667
                   -5.3333
           -5.3333
                   5.9259
C
        = double(jacobian(out,x_sym))
C = 6 \times 6
    1
         0
               0
                          0
    0
         1
               0
                    0
                          0
                                0
    0
         0
               1
                    0
                          0
                               0
         0
                          0
    0
               0
                    1
         0
               0
                    0
    0
                          1
                               0
               0
                     0
         0
                          0
                               1
D1
         = double(jacobian(out,u_sym))
D1 = 6 \times 3
    0
         0
               0
    0
         0
               0
               0
    0
         0
    0
         0
               0
         0
               0
    0
    0
         0
               0
% Unknown Input Matrix to simulate attacks on actuators
В2
        = B1
B2 = 6 \times 3
        0
                 0
                          0
        0
                 0
                          0
                 0
        0
                          0
   0.6667
          -1.3333
                          0
          10.6667
                     -5.3333
  -1.3333
                    5.9259
        0
           -5.3333
% Matrix to simulate attack on sensors (noise)
D2
        = ones(6,1)
D2 = 6 \times 1
    1
    1
    1
    1
    1
    1
% Get Dimensions
        = length(A);
[p,r] = size(D2);
% Verify M exists and calculate
if rank([C*B2 D2]) == (rank(B2) + rank(D2))
    disp('C.T. Sufficiency condition met: M exists');
    % Moore-Penrose pseudo-invsere
    CB2D2_inv = inv([C*B2 D2]'*[C*B2 D2])*[C*B2 D2]';
    % Calculate M
                 = [B2, zeros(n,r)]*CB2D2_inv;
    M CT
end
```

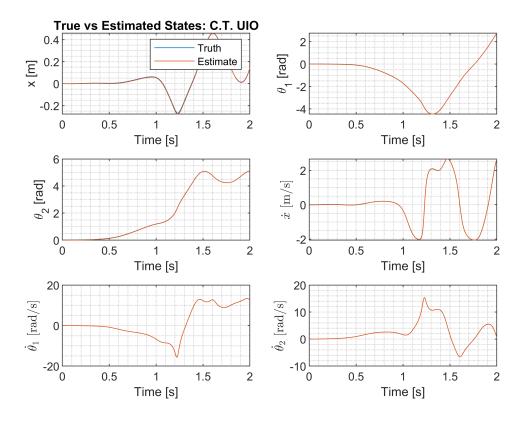
```
C.T. Sufficiency condition met: M exists
```

disp('UIO C.T. Poles are')

```
% check (I - M*C)B2 = 0 & M*D2 = 0
fprintf('\n')
disp('(I - MC)B2 = ')
(I - MC)B2 =
disp(round((eye(n) - M_CT*C)*B2))
    0
         0
              0
    0
         0
              0
    0
         0
              0
    0
         0
              0
         0
            0
         0
disp('M*D2 = ')
M*D2 =
disp(round(M_CT*D2))
    0
    0
    0
% Define A1
            = (eye(n) - M_CT*C)*A;
Α1
% Observability Matrix for pair (A1, C)
            = obsv(A1,C);
% Check detectability
if rank(ob) == n
    disp('C.T. System is observable, & thus detectable')
end
C.T. System is observable, & thus detectable
% LMI to solve for C.T. Observer Gain: LD = 0 not feasible
cvx begin sdp quiet
variable P(n,n) symmetric
variable Y(n,p)
A1'*P + P*A1 - C'*Y'-Y*C + 2*P <= 0
P >= eps*eye(n)
cvx_end
L_CT = P \setminus Y;
% UIO Poles
fprintf('\n')
```

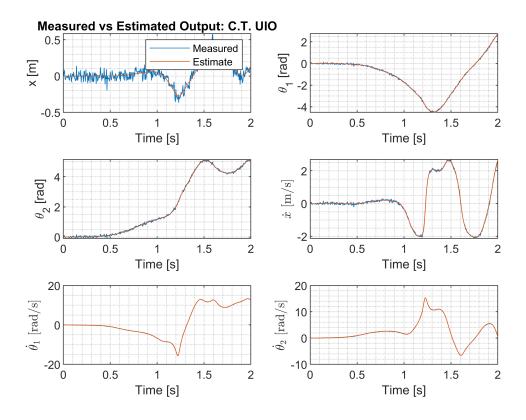
```
disp(eig(A1 - L_CT*C))
 -1.0000 + 1.3295i
 -1.0000 - 1.3295i
 -1.0000 + 0.7522i
 -1.0000 - 0.7522i
 -1.0000 + 1.0000i
 -1.0000 - 1.0000i
% State IC [m, rad, rad, m/s, rad/s, rad/s]
        = [0 .01 .02 0 0 0]';
% UIO IC
        = zeros(6,1);
z0
% Time interval and vector
dt
            = .005;
            = (0:dt:2)';
time
% ODE45 Function call
[T, X]
            = ode45(@(t,x) DIPC_UIO(t, x, zeros(3,1), A, B1, C, D2, m1, m2, M, L1, L2, g, L_CT
% Initialize arrays
            = zeros(length(z0),length(T));
ym
            = zeros(length(C),length(T));
            = zeros(length(ue),length(T));
W
% Propogate Estimates
for i = 1:length(T)
    [\sim, xhat(:,i),ym(:,i), w(:,i)] = DIPC_UIO(T(i), X(i,:)', zeros(3,1), A, B1, C, D2, m1, m2)
end
figure
subplot(3,2,1)
plot(time, X(:,1), time, xhat(1,:))
title('True vs Estimated States: C.T. UIO')
legend('Truth','Estimate')
grid minor
ylabel('x [m]')
xlabel('Time [s]')
subplot(3,2,2)
plot(time,X(:,2), time, xhat(2,:))
grid minor
ylabel('\theta_1 [rad]')
xlabel('Time [s]')
subplot(3,2,3)
plot(time, X(:,3), time, xhat(3,:))
grid minor
ylabel('\theta_2 [rad]')
xlabel('Time [s]')
subplot(3,2,4)
plot(time, X(:,4), time, xhat(4,:))
grid minor
```

```
ylabel('$\dot{x}$ [m/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,5)
plot(time,X(:,5), time, xhat(5,:))
grid minor
ylabel('$\dot{\theta}_1$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,6)
plot(time,X(:,6), time, xhat(6,:))
grid minor
ylabel('$\dot{\theta}_2$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
```

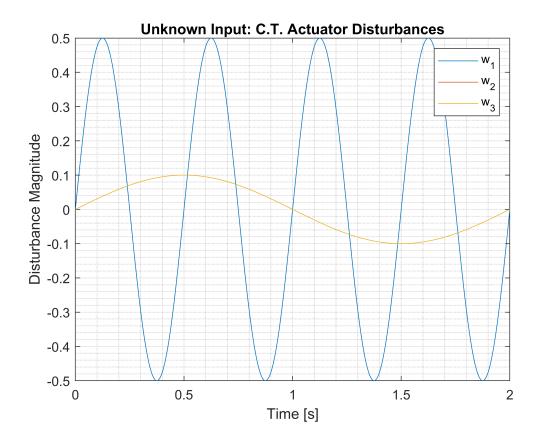


```
figure;
subplot(3,2,1)
plot(time,ym(1,:), time, xhat(1,:))
title('Measured vs Estimated Output: C.T. UIO')
legend('Measured','Estimate')
grid minor
ylabel('x [m]')
xlabel('Time [s]')
subplot(3,2,2)
plot(time,ym(2,:), time, xhat(2,:))
grid minor
ylabel('\theta_1 [rad]')
xlabel('Time [s]')
subplot(3,2,3)
plot(time,ym(3,:), time, xhat(3,:))
grid minor
```

```
ylabel('\theta 2 [rad]')
xlabel('Time [s]')
subplot(3,2,4)
plot(time,ym(4,:), time, xhat(4,:))
grid minor
ylabel('$\dot{x}$ [m/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,5)
plot(time,ym(5,:), time, xhat(5,:))
grid minor
ylabel('$\dot{\theta}_1$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,6)
plot(time,ym(6,:), time, xhat(6,:))
grid minor
ylabel('$\dot{\theta}_2$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
```



```
figure
plot(T, w)
xlabel('Time [s]')
grid minor
title('Unknown Input: C.T. Actuator Disturbances')
legend('w_1','w_2','w_3')
ylabel('Disturbance Magnitude')
```



Problem 3 - D.T. Controller UIO Compensator

```
% Sampling Time - 50 Hz
Ts = .02;

% New Initial Conditions
x0 = [.2 .2 .2 0 0 0]';

% New Observer IC
z0 = zeros(6,1);

% Discretize System
sysd = c2d(ss(A,B1,C,D1),Ts);
```

$$x[k+1] = \Phi x[k] + \Gamma_1 u_1 + \Gamma_2 u_2$$

 $y[k] = Cx[x] + D_2 u_3$

```
phi
              = sysd.A
phi = 6 \times 6
    1.0000
                                  0.0200
             -0.0016
                       0.0000
                                           -0.0000
                                                      0.0000
             1.0131
                       -0.0059
                                      0
                                           0.0201
                                                     -0.0000
             -0.0066
                        1.0066
                                           -0.0000
                                                      0.0200
                                  1.0000
             -0.1642
                       0.0003
                                           -0.0016
                                                      0.0000
             1.3150
                       -0.5925
                                       0
                                            1.0131
                                                      -0.0059
                                           -0.0066
             -0.6583
                        0.6567
                                                      1.0066
gamma
              = sysd.B
```

```
gamma = 6 \times 3
   0.0001
           -0.0003
                    0.0000
  -0.0003
            0.0021
                    -0.0011
   0.0000
           -0.0011
                     0.0012
   0.0133
           -0.0268
                     0.0001
  -0.0268
            0.2145
                    -0.1074
   0.0001
           -0.1074
                     0.1190
% Discrete Unknown Input Matrix
gamma2
             = gamma
gamma2 = 6 \times 3
          -0.0003
                    0.0000
   0.0001
          0.0021 -0.0011
  -0.0003
          -0.0011
                   0.0012
   0.0000
   0.0133
          -0.0268
                   0.0001
  -0.0268
           0.2145
                   -0.1074
   0.0001
           -0.1074
                   0.1190
% Verify M exists and calculate
if rank([C*gamma2 D2]) == (rank(gamma2) + rank(D2))
    disp('D.T. Sufficiency condition met: M exists');
    % Moore-Penrose pseudo-invsere
    CB2D2_inv_DT = inv([C*gamma2 D2]'*[C*gamma2 D2])*[C*gamma2 D2]';
    % Calculate M
    M DT
                  = [gamma2, zeros(n,r)]*CB2D2_inv_DT;
end
D.T. Sufficiency condition met: M exists
% check (I - M*C)B2 = 0 & M*D2 = 0
fprintf('\n')
disp('(I - MC)B2 = ')
(I - MC)B2 =
disp(round((eye(n) - M_DT*C)*gamma2))
    0
         0
               0
    0
         0
               0
         0
               0
disp('M*D2 = ')
M*D2 =
disp(round(M_DT*D2))
    0
    0
    0
    0
    0
```

```
% Define A1
A1_DT = (eye(n) - M_DT*C)*phi;

% Observability Matrix for pair (A1, C)
ob = obsv(A1_DT,C);

% Check detectability
if rank(ob) == n
    disp('D.T. System is observable, & thus detectable')
end
```

D.T. System is observable, & thus detectable

```
% LMI to solve for D.T. observer gain
cvx_begin sdp quiet
variable P(n,n) symmetric
variable Y(n,p)
[-P, A1_DT'*P-C'*Y'; P*A1_DT-Y*C, -P]<= 0
Y*D2 == 0
P >= eps*eye(n)
cvx_end

% Discrete observer gain
L_DT = P\Y;

% UIO Poles
fprintf('\n')
disp('UIO D.T. Poles are')
```

UIO D.T. Poles are

```
disp(eig(A1_DT - L_DT*C))
```

```
0.9916 + 0.0000i
-0.8311 + 0.0000i
-0.0000 + 0.0000i
0.0000 + 0.0000i
0.0000 + 0.0000i
0.0000 - 0.0000i
```

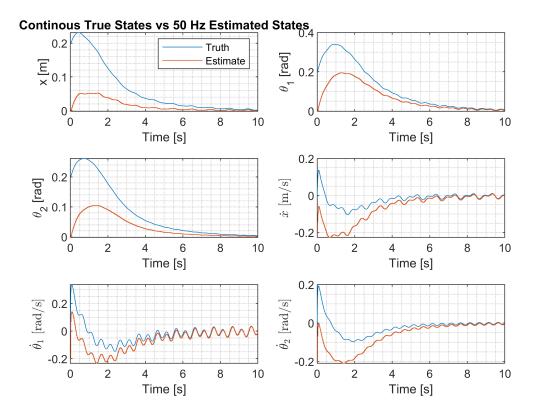
```
% DARE Weight Matrices
Q
            = 100*eye(n);
            = .1*eye(length(ue));
R
% Number of inputs
            = size(B1,2);
m_u
% LMI to solve DARE for K
cvx_begin sdp quiet
variable S(n, n) symmetric
variable Z(m u,n)
[-S, S, Z',Z'*gamma'-S*phi';...
S, -inv(Q), zeros(n,m_u), zeros(n,n);...
Z, zeros(m_u,n), -inv(R), zeros(m_u,n);...
gamma*Z-phi*S, zeros(n,n), zeros(n,m_u), -S] <= 0</pre>
S >= eps*eye(n)
```

```
cvx_end
% D.T. Controller Gain
K DT
            = Z/S;
% System Closed Loop Poles eig(A-BK)
fprintf('\n')
disp('Closed Loop D.T. Poles are')
Closed Loop D.T. Poles are
disp(eig(phi - gamma*K_DT))
   0.0057
   0.1558
   0.5643
   0.9278
   0.9658
   0.9741
             = sim('HW5 model.slx');
sim out
% Extract Simulink Signals
x_ct
             = sim out.logsout{1}.Values.Data(:,:);
             = sim_out.logsout{1}.Values.Time;
t_ct
xhat DT
             = sim_out.logsout{3}.Values.Data(:,:);
t 50Hz
             = sim out.logsout{3}.Values.Time;
ym_DT
             = sim_out.logsout{6}.Values.Data(:,:);
             = sim_out.logsout{4}.Values.Data(:,:);
u_DT
             = sim_out.logsout{5}.Values.Data(:,:);
w_DT
figure
subplot(3,2,1)
plot(t_ct,x_ct(1,:))
title('Continous True States vs 50 Hz Estimated States')
hold on
stairs(t_50Hz, xhat_DT(1,:))
hold off
legend('Truth','Estimate')
grid minor
ylabel('x [m]')
xlabel('Time [s]')
subplot(3,2,2)
plot(t_ct,x_ct(2,:))
hold on
stairs(t_50Hz, xhat_DT(2,:))
hold off
grid minor
ylabel('\theta_1 [rad]')
xlabel('Time [s]')
subplot(3,2,3)
plot(t_ct,x_ct(3,:))
hold on
```

stairs(t_50Hz, xhat_DT(3,:))

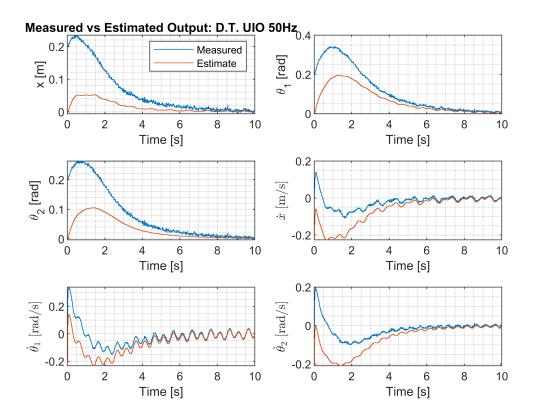
hold off grid minor

```
ylabel('\theta 2 [rad]')
xlabel('Time [s]')
subplot(3,2,4)
plot(t_ct,x_ct(4,:))
hold on
stairs(t_50Hz, xhat_DT(4,:))
hold off
grid minor
ylabel('$\dot{x}$ [m/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,5)
plot(t_ct,x_ct(5,:))
hold on
stairs(t_50Hz, xhat_DT(5,:))
hold off
grid minor
ylabel('$\dot{\theta}_1$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,6)
plot(t_ct,x_ct(6,:))
hold on
stairs(t_50Hz, xhat_DT(6,:))
hold off
grid minor
ylabel('$\dot{\theta}_2$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
```

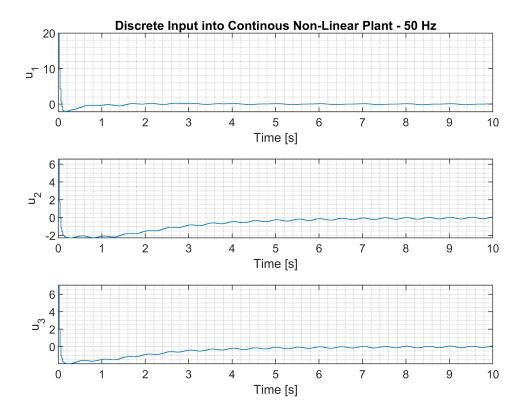


```
figure subplot(3,2,1)
```

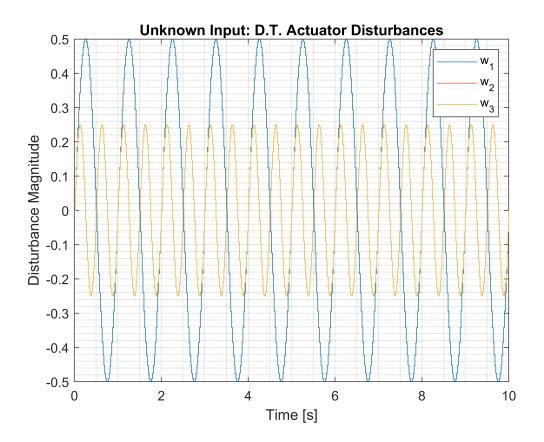
```
stairs(t 50Hz,ym DT(1,:))
hold on
title('Measured vs Estimated Output: D.T. UIO 50Hz')
stairs(t 50Hz, xhat DT(1,:))
hold off
legend('Measured','Estimate')
grid minor
ylabel('x [m]')
xlabel('Time [s]')
subplot(3,2,2)
stairs(t_50Hz,ym_DT(2,:))
hold on
stairs(t_50Hz, xhat_DT(2,:))
hold off
grid minor
ylabel('\theta_1 [rad]')
xlabel('Time [s]')
subplot(3,2,3)
stairs(t_50Hz,ym_DT(3,:))
hold on
stairs(t_50Hz, xhat_DT(3,:))
hold off
grid minor
ylabel('\theta_2 [rad]')
xlabel('Time [s]')
subplot(3,2,4)
stairs(t_50Hz,ym_DT(4,:))
hold on
stairs(t_50Hz, xhat_DT(4,:))
hold off
grid minor
ylabel('$\dot{x}$ [m/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,5)
stairs(t_50Hz,ym_DT(5,:))
hold on
stairs(t_50Hz, xhat_DT(5,:))
hold off
grid minor
ylabel('$\dot{\theta}_1$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
subplot(3,2,6)
stairs(t_50Hz,ym_DT(6,:))
hold on
stairs(t_50Hz, xhat_DT(6,:))
hold off
grid minor
ylabel('$\dot{\theta}_2$ [rad/s]','Interpreter','latex')
xlabel('Time [s]')
```



```
figure
subplot(311)
stairs(t_50Hz, u_DT(1,:))
xlabel('Time [s]')
ylabel('u_1')
grid minor
title('Discrete Input into Continous Non-Linear Plant - 50 Hz')
subplot(312)
stairs(t_50Hz, u_DT(2,:))
xlabel('Time [s]')
ylabel('u_2')
grid minor
subplot(313)
stairs(t_50Hz, u_DT(3,:))
xlabel('Time [s]')
ylabel('u_3')
grid minor
```



```
figure
stairs(t_50Hz, w_DT)
xlabel('Time [s]')
grid minor
title('Unknown Input: D.T. Actuator Disturbances')
legend('w_1','w_2','w_3')
ylabel('Disturbance Magnitude')
```



Functions

```
function xdot = DIPC(t, x, u, m1, m2, M, L1, L2, g)
% Define State and Input Vectors
x1
            = x(1,1);
                      % x
            = x(2,1);
                       % theta 1
x2
            = x(3,1);
                       % theta 2
х3
            = x(4,1);
                       % xdot
х4
x5
            = x(5,1); % theta_1_dot
            = x(6,1);
                       % theta_2_dot
х6
u1
            = u(1,1);
            = u(2,1);
u2
u3
            = u(3,1);
% State Dynamics
x1dot
                    % xdot
            = x4;
x2dot
                    % theta 1 dot
            = x5;
                    % theta_2_dot
x3dot
            = x6;
% x ddot
x4dot
            = (L2*m2*u2*cos(x2 - 2*x3) - L1*m1*u3*cos(x3) - L2*m2*u2*cos(x2)...
               -L1*m2*u3*cos(x3) - 2*L2*m1*u2*cos(x2) + 2*L1*L2*m1*u1 + L1*L2*m2*u1...
               + L1*m1*u3*cos(2*x2 - x3) + L1*m2*u3*cos(2*x2 - x3) - ...
               L1*L2*m2*u1*cos(2*x2 - 2*x3) - L1*L2*g*m1^2*sin(2*x2) +...
               2*L1^2*L2*m1^2*x5^2*sin(x2) - L1*L2*g*m1*m2*sin(2*x2) +...
```

```
L1*L2^2*m1*m2*x6^2*sin(2*x2 - x3) + 2*L1^2*L2*m1*m2*x5^2*sin(x2) + ...
               L1*L2^2*m1*m2*x6^2*sin(x3))/(L1*L2*(2*M*m1 + M*m2 + m1*m2 - ...
               m1^2*\cos(2*x^2) + m1^2 - m1*m2*\cos(2*x^2) - M*m2*\cos(2*x^2 - 2*x^3));
% theta_1_ddot
x5dot
            = -(L2*m2*u2*cos(2*x3) - 2*L2*m1*u2 - L2*m2*u2 - 2*L2*M*u2 - ...
                2*L1*L2*g*m1^2*sin(x2) + 2*L1*M*u3*cos(x2)*cos(x3) +...
                2*L1*M*u3*sin(x2)*sin(x3) + L1^2*L2*m1^2*x5^2*sin(2*x2) + ...
                2*L1*m1*u3*sin(x2)*sin(x3) + 2*L1*m2*u3*sin(x2)*sin(x3) + ...
                2*L1*L2*m1*u1*cos(x2) + L1*L2*m2*u1*cos(x2) - 2*L1*L2*M*g*m1*sin(x2)...
                - L1*L2*M*g*m2*sin(x2) - L1*L2*m2*u1*sin(2*x3)*sin(x2) -...
                2*L1*L2*g*m1*m2*sin(x2) + L1^2*L2*m1*m2*x5^2*sin(2*x2) - ...
                L1*L2*m2*u1*cos(2*x3)*cos(x2) - L1*L2*M*g*m2*cos(2*x3)*sin(x2) +...
                L1*L2*M*g*m2*sin(2*x3)*cos(x2) - L1^2*L2*M*m2*x5^2*cos(2*x2)*sin(2*x3) +...
                L1^2*L2*M*m2*x5^2*cos(2*x3)*sin(2*x2) - 2*L1*L2^2*M*m2*x6^2*cos(x2)*sin(x3) + ...
                2*L1*L2^2*M*m2*x6^2*cos(x3)*sin(x2) + 2*L1*L2^2*m1*m2*x6^2*cos(x3)*sin(x2))/...
                (L1^2*L2*(2*M*m1 + M*m2 + m1*m2 - m1^2*cos(2*x2) + m1^2 - ...
                m1*m2*cos(2*x2) - M*m2*cos(2*x2 - 2*x3)));
% theta 2 ddot
x6dot
            = (L1*m1^2*u3 + L1*m2^2*u3 - L1*m1^2*u3*cos(2*x2) - L1*m2^2*u3*cos(2*x2) +...
              L2*m2^2*u2*cos(x2 + x3) + 2*L1*M*m1*u3 + 2*L1*M*m2*u3 + 2*L1*m1*m2*u3 - ...
              L2*m2^2*u2*cos(x2 - x3) - L2*m1*m2*u2*cos(x2 - x3) - L1*L2*m2^2*u1*cos(x3) - ...
              2*L1*m1*m2*u3*cos(2*x2) + L1*L2*m2^2*u1*cos(2*x2 - x3) + ...
              L2*m1*m2*u2*cos(x2 + x3) - 2*L2*M*m2*u2*cos(x2 - x3) - L1*L2*m1*m2*u1*cos(x3) +
              L1*L2*M*g*m2^2*sin(x3) + 2*L1^2*L2*M*m2^2*x5^2*sin(x2 - x3) + ...
              L1*L2*m1*m2*u1*cos(2*x2 - x3) + L1*L2^2*M*m2^2*x6^2*sin(2*x2 - 2*x3) - ...
              L1*L2*M*g*m2^2*sin(2*x2 - x3) - L1*L2*M*g*m1*m2*sin(2*x2 - x3) + ...
              L1*L2*M*g*m1*m2*sin(x3) + 2*L1^2*L2*M*m1*m2*x5^2*sin(x2 - x3))/...
              (L1*L2^2*m2*(2*M*m1 + M*m2 + m1*m2 - m1^2*cos(2*x2) + m1^2 - ...
              m1*m2*cos(2*x2) - M*m2*cos(2*x2 - 2*x3)));
xdot
            = [x1dot;x2dot;x3dot;x4dot;x5dot;x6dot];
end
function [xdot, xhat, ym, w] = DIPC_UIO(t, x, u, A, B1, C, D2, m1, m2, M, L1, L2, g, L, M_UIO)
% Define State and Input Vectors
х1
            = x(1,1);
                            % x
            = x(2,1);
                            % theta 1
x2
                            % theta_2
            = x(3,1);
х3
            = x(4,1);
                            % xdot
х4
            = x(5,1);
                            % theta 1 dot
x5
            = x(6,1);
                            % theta 2 dot
х6
            = x(7:end,1); % UIO States
Z
% Outputs
            = C*x(1:6,1);
У
% Sensor Attacks - noise applied to each sensor
            = .05*randn(1);
```

```
% Measured Ouput
            = y + D2*v;
ym
% Distrubances - Actuator Attacks
            = .5*sin(4*pi*t);
w1
            = .1*sin(pi*t);
w2
            = w2;
w3
            = [w1; w2; w3];
W
u1
            = u(1,1) + w1;
            = u(2,1) + w2;
u2
            = u(3,1) + w3;
u3
% State Dynamics
x1dot
            = x4;
                    % xdot
            = x5;
                    % theta 1 dot
x2dot
                    % theta 2 dot
x3dot
            = x6;
% State Estimates
xhat
            = z + M_UIO*ym;
% Estimated Output
            = C*xhat;
yhat
% x ddot
x4dot
            = (L2*m2*u2*cos(x2 - 2*x3) - L1*m1*u3*cos(x3) - L2*m2*u2*cos(x2)...
               -L1*m2*u3*cos(x3) - 2*L2*m1*u2*cos(x2) + 2*L1*L2*m1*u1 + L1*L2*m2*u1...
               + L1*m1*u3*cos(2*x2 - x3) + L1*m2*u3*cos(2*x2 - x3) - ...
               L1*L2*m2*u1*cos(2*x2 - 2*x3) - L1*L2*g*m1^2*sin(2*x2) + ...
               2*L1^2*L2*m1^2*x5^2*sin(x2) - L1*L2*g*m1*m2*sin(2*x2) +...
               L1*L2^2*m1*m2*x6^2*sin(2*x2 - x3) + 2*L1^2*L2*m1*m2*x5^2*sin(x2) + ...
               L1*L2^2*m1*m2*x6^2*sin(x3))/(L1*L2*(2*M*m1 + M*m2 + m1*m2 - ...
               m1^2*\cos(2*x^2) + m1^2 - m1*m2*\cos(2*x^2) - M*m2*\cos(2*x^2 - 2*x^3));
% theta_1_ddot
            = -(L2*m2*u2*cos(2*x3) - 2*L2*m1*u2 - L2*m2*u2 - 2*L2*M*u2 - ...
x5dot
                2*L1*L2*g*m1^2*sin(x^2) + 2*L1*M*u3*cos(x^2)*cos(x^3) +...
                2*L1*M*u3*sin(x2)*sin(x3) + L1^2*L2*m1^2*x5^2*sin(2*x2) + ...
                2*L1*m1*u3*sin(x2)*sin(x3) + 2*L1*m2*u3*sin(x2)*sin(x3) + ...
                2*L1*L2*m1*u1*cos(x2) + L1*L2*m2*u1*cos(x2) - 2*L1*L2*M*g*m1*sin(x2)...
                - L1*L2*M*g*m2*sin(x2) - L1*L2*m2*u1*sin(2*x3)*sin(x2) -...
                2*L1*L2*g*m1*m2*sin(x2) + L1^2*L2*m1*m2*x5^2*sin(2*x2) - ...
                L1*L2*m2*u1*cos(2*x3)*cos(x2) - L1*L2*M*g*m2*cos(2*x3)*sin(x2) +...
                L1*L2*M*g*m2*sin(2*x3)*cos(x2) - L1^2*L2*M*m2*x5^2*cos(2*x2)*sin(2*x3) +...
                L1^2*L2^4M^2*x5^2*cos(2^2x3)^3sin(2^2x2) - 2^2L1^2L2^2*M^2*m2^2*x6^2*cos(x2)^2sin(x3) + ...
                2*L1*L2^2*M*m2*x6^2*cos(x3)*sin(x2) + 2*L1*L2^2*m1*m2*x6^2*cos(x3)*sin(x2))/...
                (L1^2*L2*(2*M*m1 + M*m2 + m1*m2 - m1^2*cos(2*x2) + m1^2 - ...
                m1*m2*cos(2*x2) - M*m2*cos(2*x2 - 2*x3)));
% theta 2 ddot
x6dot
            = (L1*m1^2*u3 + L1*m2^2*u3 - L1*m1^2*u3*cos(2*x2) - L1*m2^2*u3*cos(2*x2) + ...
              L2*m2^2*u2*cos(x2 + x3) + 2*L1*M*m1*u3 + 2*L1*M*m2*u3 + 2*L1*m1*m2*u3 - ...
              L2*m2^2*u2*cos(x2 - x3) - L2*m1*m2*u2*cos(x2 - x3) - L1*L2*m2^2*u1*cos(x3) - ...
```

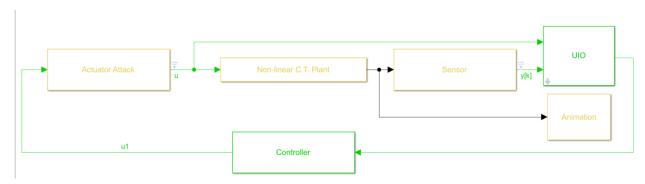
```
2*L1*m1*m2*u3*cos(2*x2) + L1*L2*m2^2*u1*cos(2*x2 - x3) +...
L2*m1*m2*u2*cos(x2 + x3) - 2*L2*M*m2*u2*cos(x2 - x3) - L1*L2*m1*m2*u1*cos(x3) +.
L1*L2*M1*m2*u1*cos(2*x2 - x3) + L1*L2*M1*m2*u1*cos(x3) +...
L1*L2*m1*m2*u1*cos(2*x2 - x3) + L1*L2*M*m2^2*x5^2*sin(x2 - x3) +...
L1*L2*M1*m2*u1*cos(2*x2 - x3) + L1*L2*M1*m2*sin(2*x2 - 2*x3) -...
L1*L2*M1*m2*m2*sin(2*x2 - x3) - L1*L2*M1*m2*sin(2*x2 - x3) +...
L1*L2*M1*m2*sin(x3) + 2*L1^2*M1*m2*x5^2*sin(x2 - x3) +...
L1*L2*M1*m2*sin(x3) + 2*L1^2*M1*m1*m2*x5^2*sin(x2 - x3))/...
(L1*L2^2*m2*(2*M1*m1 + M1*m2 + m1*m2 - m1^2*cos(2*x2) + m1^2 -...
m1*m2*cos(2*x2) - M1*m2*cos(2*x2 - 2*x3)));

% UIO Dynamics
zdot = (eye(size(M_UIO*C)) - M_UIO*C)*(A*z + A*M_UIO*ym + B1*u) + L*(ym - yhat);

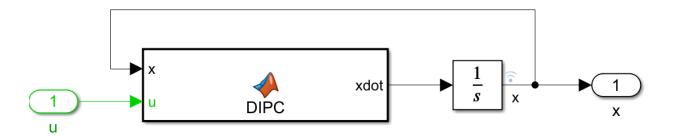
xdot = [x1dot;x2dot;x3dot;x4dot;x5dot;x6dot;zdot];
end
```

Simulink

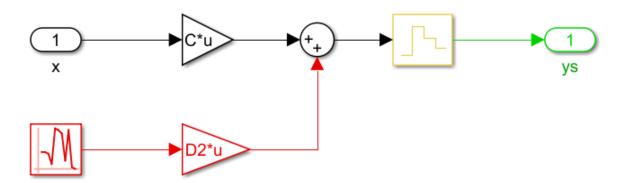
Top Level:



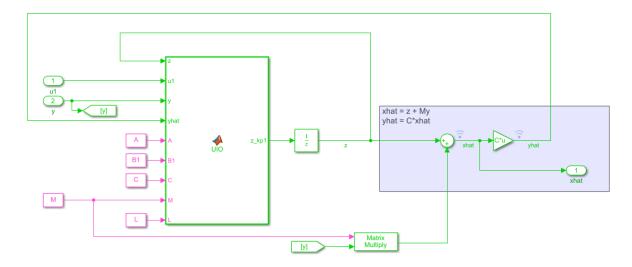
<u>Plant:</u>



Sensors:

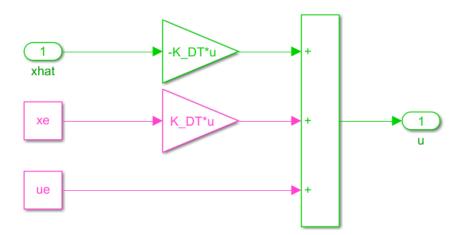


UIO:

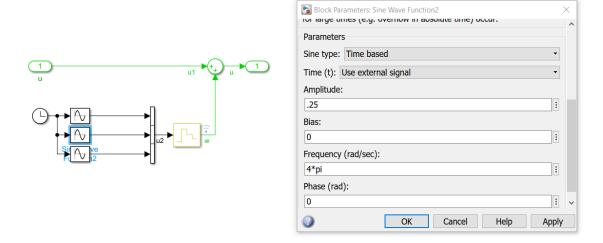


Controller:

$$u = -Kx + ue + Kxe = ue - Kdel_x$$



Actuator:



Animation:

