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```
%{
The following code is used to design an optimal controller and estimator for 2 pendulum of
different length and mass on a single cart. The only acutator on this car
the motor that drives the linear position of the cart.

We are attempting to swing up both the pendulum from the initial position of
all pendulum down to both pendulum up and stabilize

%}
```

Determine u_k and state

```
clear all; clc; close all
load_prev = 1; %set to 1 if a previous trajectory has been generated

%{
Use code provided in NR Ch21 to generate  $x_k$  state,  $u_k$ , using adjoint
method for the swing up phase. This phase last about  $T = 3$  seconds because

0) Within this code, tune  $QT$  to drive error at  $T$  close to zero
1) Warm start: use an inital guess of  $u_k$  being all zeros
2) iterate the provided function with a new  $u_k$  until the state reach near
desired value of [0 0 0 0 0 0]
3)save  $x_k$  and  $u_k$ 

%}
```

Uncomment the Following Code if you need an x_k and u_k using Adjoint method to generate swingup states trajectory

```
% T = 3; %approx sqrt(g/l)
%
% %initial guess of  $u_k$  being all zero
% [u_k(:,1),x_k] = OptimizeTraj(T);
%
% state_at_T(:,1) = x_k(1:6, end);
%
% number_of_trial = 1;
% trial_number = 1:number_of_trial
% for i = trial_number
%     [u_k(:,i+1), x_k] = OptimizeTraj(T, u_k(:,i))
%     state_at_T(:,i+1) = x_k(1:6, end)
% end
```

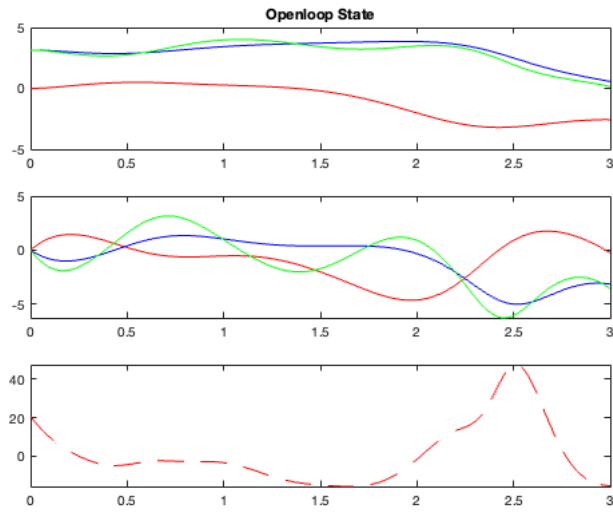
If Previous Trajectory has been generated, run this code

```
%Trajectory can be manually optimized by tuning Q iteration by iteration by
%using the file "AdjointPathManualIteration.m"

%load previous trajectory is better

if load_prev == 1
    clear all; clc;
    load("SavePoint3.mat")
    figure(1); clf; subplot(3,1,1); plot(t,x_k(1,:), 'r-',t,x_k(2,:), 'b-',t,x_k(3,:), 'g-'); hold on; title("Openloop State");
    subplot(3,1,2); plot(t,x_k(4,:), 'r-',t,x_k(5,:), 'b-',t,x_k(6,:), 'g-'); hold on;
    subplot(3,1,3); plot(t,u_k, 'r--'); hold on;

end
```



Find $A(t)$, $E(t)$, $B(t)$ for Each x_k in the swing up phase

```
%From generated state trajectory, compute for the state matrix E_t, A_t,
%B_t, C_t, D_t wrt time for every timestep h since the system is not
%linearizable during the swing up case

s.h=0.01; s.N=T/s.h; s.mc=10; t=[0:s.N]*s.h; % STEP 0: initialize simulation
s.m1=1; s.L1=1; s.e111=s.L1; s.I1=s.m1*s.e111^2/3; % system, & derived parameters
s.m2=0.5; s.L2=0.5; s.e112=s.L2; s.I2=s.m2*s.e112^2/3; alpha=0.1;
s.B=[0; 0; 0; 1; 0; 0]; s.Q=diag([0 0 0 0 0 0]); s.R=0; s.QT=diag([5 40 10 .1 60 10]);
g = 9.81;

for i = 1:length(x_k)
    A_temp = Compute_A(x_k(:,i),s); %NR funtion
    A_t(i) = {A_temp};
end

for i = 1:length(x_k)
    E_temp = Compute_E(x_k(1:6,i),s); %NR funtion
    E_t(i) = {E_temp};
end

for i = 1:length(x_k)
    B_temp = [0; 0; 0; 1; 0; 0]; %NR funtion
    B_t(i) = {B_temp};
end

for i = 1:length(x_k)
    C_temp = [[1 0 0 0 0 0];... %Sending x, thetal, theta2
              [0 1 0 0 0 0];...
              [0 0 1 0 0 0];...
              [0 0 0 0 0 0];...
              [0 0 0 0 0 0];...
              [0 0 0 0 0 0]];
    C_t(i) = {C_temp};
end
```

Compute $K(t)$ using $A(t)$, $E(t)$, $B(t)$, $C(t)$

```
%{
    compute controller gain K(t) using differential ricatti equation (DRE),
    which can be solved using RK4 marching. This K(t) controls the the system from
    one timestep to another

%}
h = s.h; %timestep

R = .1; %controller cost
Q = [[1 0 0 0 0 0];... %penilize theta anlgas more than other states
     [0 7 0 0 0 0];...
     [0 0 7 0 0 0];...
     [0 0 0 1 0 0];...
     [0 0 0 0 1 0];...
     [0 0 0 0 0 1]];

X_t = {}; %allocate cell frame to store X at every time frame
X_t(length(u_k)) = {eye(6)}; %X at T
```

```

%Backward March to find X(t)using RK4 + DRE
for i = length(u_k):-1:2

f1 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i},R,Q);
f2 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i} - f1*h/2,R,Q);
f3 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i} - f2*h/2,R,Q);
f4 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i} - f3*h,R,Q);

X_t{i-1} = X_t{i} - h*(f1/6 + f2/3 + f3/3 + f4/6);

end

%find K(t) using X(t) using NR22.13a
for i = 1:length(u_k)
    K_t(i) = {-inv(R)*B_t{i}'*X_t{i}};
end

```

Compute L(t) using A(t), E(t), B(t), C(t)

```

%{
    compute observer gain L(t) using differential ricatti equation, which
    can be solved using RK4 forward marching
}%

Q1 = eye(6); %covariance of state disturbance
Q2 = eye(6); %convariance of sensor noise

P_t = {}; %allocate space for storage of P

P_t(1) = {eye(6)}; %intial estimation covariance

%forward march to find L(t) using RK4 and DRE NR22.30
for i = 1:1:length(u_k)

f1 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i},R,Q);
f2 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i} + f1*h/2,R,Q);
f3 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i} + f2*h/2,R,Q);
f4 = KRiccati(E_t{i},A_t{i},B_t{i},X_t{i} + f3*h,R,Q);

P_t{i+1} = P_t{i} + h*(f1/6 + f2/3 + f3/3 + f4/6);

end

%find L(t) using P(t) by appying NR22.30
for i = 1:length(u_k)
    L_t(i) = {-P_t{i}*C_t{i}'*inv(Q2)};
end

```

Uncomment this section to check eigenvalues of the control + observation system

```

% eigControl = {};
% ControlSys = {};
% for i = 1:length(u_k)
%     ControlSys(i) = {inv(E_t{i})*A_t{i} + inv(E_t{i})* B_t{i}*K_t{i}};
%     [evecK, eigControl{i}] = eig(ControlSys{i});
% end

% eigEstimate = {};
% EstimateSys = {};
% for i = 1:length(u_k)
%     EstimateSys(i) = {A_t{i} + L_t{i}*C_t{i}};
%     [evecL, eigEstimate{i}] = eig(EstimateSys{i});
% end

```

Plot State Trajectory using K(t) Controller

```

%From derivation, dx'/dt = inv(E) * [A + Bk - Bk*x_bar + Bu_bar]
%Note that u_k != K(t)*x_k because u_k is designed from openloop adjoint
%methods

% March protabation x_prime using RK4 and dx'/dt derivation
dx_prime = {};
x_prime(:,1) = [0 0 0 0 0 0]';
for i = 1:1:length(u_k)-1

    f1 = inv(E_t{i})*(A_t{i} + B_t{i})* x_prime(:,i) - inv(E_t{i})*B_t{i}*K_t{i}*x_k(1:6,i) + inv(E_t{i})*B_t{i}*u_k(i);
    f2 = inv(E_t{i})*(A_t{i} + B_t{i})* (x_prime(:,i) + f1*h/2) - inv(E_t{i})*B_t{i}*K_t{i}*x_k(1:6,i) + inv(E_t{i})*B_t{i}*u_k(i);
    f3 = inv(E_t{i})*(A_t{i} + B_t{i})* (x_prime(:,i) + f2*h/2) - inv(E_t{i})*B_t{i}*K_t{i}*x_k(1:6,i) + inv(E_t{i})*B_t{i}*u_k(i);
    f4 = inv(E_t{i})*(A_t{i} + B_t{i})* (x_prime(:,i) + f3*h) - inv(E_t{i})*B_t{i}*K_t{i}*x_k(1:6,i) + inv(E_t{i})*B_t{i}*u_k(i);

```

```

x_prime(:,i+1) = x_prime(i) + h*(f1/6 + f2/3 + f3/3 + f4/6);
end

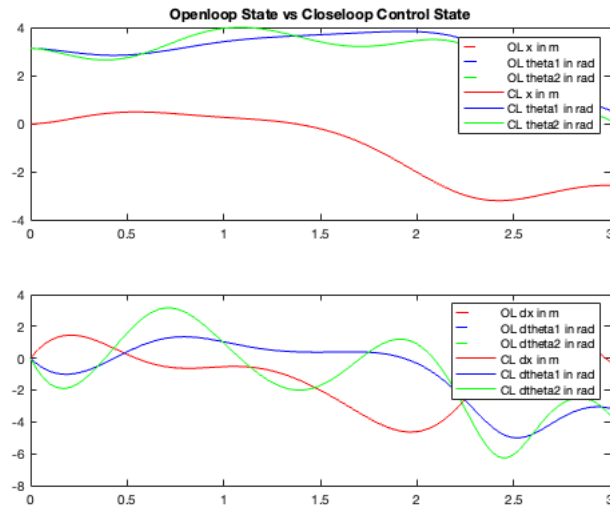
% Determine state simulation by adding nominal state and the proptabation states
% ideally, protabation tends to zero if controller is designed properly
% x_sim should ideally be very close to x_k
x_sim = x_k(1:6, :) + x_prime*h;

figure(2); subplot(2,1,1); plot(t,x_k(1,:), 'r--', t,x_k(2,:), 'b--', t,x_k(3,:), 'g--'); hold on; title("Openloop State vs Closeloop Control State")
    legend("OL x in m", "OL theta1 in rad", "OL theta2 in rad")
    subplot(2,1,2); plot(t,x_k(4,:), 'r--', t,x_k(5,:), 'b--', t,x_k(6,:), 'g--'); hold on;
    legend("OL dx in m", "OL dtheta1 in rad", "OL dtheta2 in rad")

figure(2); subplot(2,1,1); plot(t,x_sim(1,:), 'r', t,x_sim(2,:), 'b', t,x_sim(3,:), 'g');
    legend("OL x in m", "OL theta1 in rad", "OL theta2 in rad", "CL x in m", "CL theta1 in rad", "CL theta2 in rad")
    subplot(2,1,2); plot(t,x_sim(4,:), 'r-', t,x_sim(5,:), 'b-', t,x_sim(6,:), 'g-');
    legend("OL dx in m", "OL dtheta1 in rad", "OL dtheta2 in rad", "CL dx in m", "CL dtheta1 in rad", "CL dtheta2 in rad")

%plots shows convergence of simulated close loop x and open loop x

```



Add Estimation Technique to system using L(t)

```

% March protabation x_prime using the following derivation
%{
dx_prime = (inv(E_t)*A_t + L_t*C_t)* x_hat_prime(:,i)...
    + (2*L_t*C_t - inv(E_t)*B_t*K_t)*x_k(1:6, i)...
    + (L_t*C_t + inv(E_t)*B_t*K_t)*x_prime(:,i)...
    + inv(E_t)*B_t*u_k(i);
%}

x_hat_prime(:,1) = [0 0 0 0 0 0]';
for i = 1:length(u_k)-1

    f1 = (inv(E_t{i})*A_t{i} + L_t{i}*C_t{i})* x_hat_prime(:,i)...
        + (2*L_t{i}*C_t{i} - inv(E_t{i})*B_t{i}*K_t{i})*x_k(1:6, i)...
        + (L_t{i}*C_t{i} + inv(E_t{i})*B_t{i}*K_t{i})*x_prime(:,i)...
        + inv(E_t{i})*B_t{i}*u_k(i);

    f2 = (inv(E_t{i})*A_t{i} + L_t{i}*C_t{i})* (x_hat_prime(:,i) + f1*h/2) ...
        + (2*L_t{i}*C_t{i} - inv(E_t{i})*B_t{i}*K_t{i})*x_k(1:6, i)...
        + (L_t{i}*C_t{i} + inv(E_t{i})*B_t{i}*K_t{i})*x_prime(:,i)...
        + inv(E_t{i})*B_t{i}*u_k(i);

    f3 = (inv(E_t{i})*A_t{i} + L_t{i}*C_t{i})* (x_hat_prime(:,i) + f2*h/2) ...
        + (2*L_t{i}*C_t{i} - inv(E_t{i})*B_t{i}*K_t{i})*x_k(1:6, i)...
        + (L_t{i}*C_t{i} + inv(E_t{i})*B_t{i}*K_t{i})*x_prime(:,i)...
        + inv(E_t{i})*B_t{i}*u_k(i);

    f4 = (inv(E_t{i})*A_t{i} + L_t{i}*C_t{i})* (x_hat_prime(:,i) + f3*h) ...
        + (2*L_t{i}*C_t{i} - inv(E_t{i})*B_t{i}*K_t{i})*x_k(1:6, i)...
        + (L_t{i}*C_t{i} + inv(E_t{i})*B_t{i}*K_t{i})*x_prime(:,i)...
        + inv(E_t{i})*B_t{i}*u_k(i);

    x_hat_prime(:,i+1) = x_hat_prime(i) + h*(f1/6 + f2/3 + f3/3 + f4/6);

end

```

```

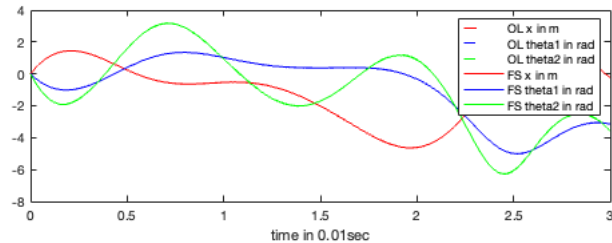
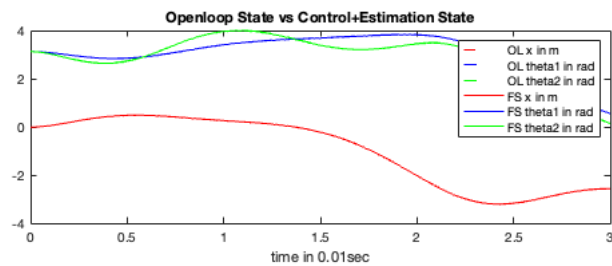
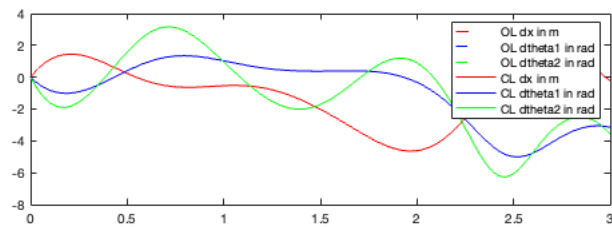
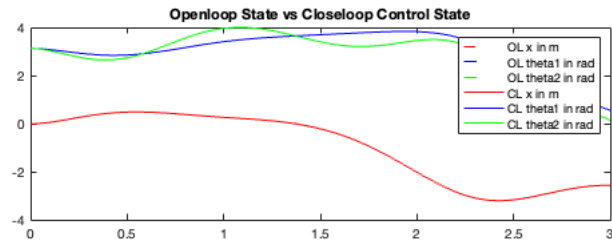
x_estimation_sim = x_k(1:6, :) + x_hat_prime*h;

figure(3); subplot(2,1,1); plot(t,x_k(1,:), 'r--', t,x_k(2,:), 'b--', t,x_k(3,:), 'g--'); hold on; title("Openloop State vs Control+Estimation State")
    legend("OL x in m", "OL theta1 in rad", "OL theta2 in rad")
    xlabel("time in 0.01sec")
    subplot(2,1,2); plot(t,x_k(4,:), 'r--', t,x_k(5,:), 'b--', t,x_k(6,:), 'g--'); hold on;
    legend("OL dx in m", "OL dtheta1 in rad", "OL dtheta2 in rad", "OL x in m", "OL theta1 in rad", "OL theta2 in rad")
    xlabel("time in 0.01sec")

figure(3); subplot(2,1,1); plot(t,x_estimation_sim (1,:), 'r', t,x_estimation_sim (2,:), 'b', t,x_estimation_sim (3,:), 'g');
    legend("OL x in m", "OL theta1 in rad", "OL theta2 in rad", "FS x in m", "FS theta1 in rad", "FS theta2 in rad")
    xlabel("time in 0.01sec")
    subplot(2,1,2); plot(t,x_estimation_sim (4,:), 'r', t,x_estimation_sim (5,:), 'b', t,x_estimation_sim (6,:), 'g');
    legend("OL dx in m", "OL dtheta1 in rad", "OL dtheta2 in rad", "FS x in m", "FS theta1 in rad", "FS theta2 in rad")
    xlabel("time in 0.01sec")

```

Warning: Ignoring extra legend entries.



Part LTI Component of Control Problem

```

%{once states are near zero and can be linearizable, we can find a
%constant K and L using algebraic riccati by calling on icare

```

```

%compute state matrix when linearized about [0 0 0 0 0 0]

```

```

E44 = s.m1 + s.m2 + s.mc; E45 = -s.m1*s.L1; E46 = -s.m2*s.L2;
E54 = -s.m1*s.L1 ; E55 = s.I1 + s.m1*s.L1^2;
E64 = -s.m2*s.L2 ; E66 = s.I2 + s.m2*s.L2^2;

```

```

E_bar = [[1 0 0 0 0 0];...
          [0 1 0 0 0 0];
          [0 0 1 0 0 0];
          [0 0 0 E44 E45 E46];
          [0 0 0 E54 E55 0 ];
          [0 0 0 E64 0 E66]];

A52 = s.m1*g*s.L1;
A63 = s.m2*g*s.L2;
A_bar = [[0 0 0 1 0 0];...
          [0 0 0 0 1 0];
          [0 0 0 0 0 1];
          [0 0 0 0 0 0];
          [0 A52 0 0 0 0];
          [0 0 A63 0 0 0]];

B_bar = [0 0 0 1 0 0]';

%Compute classical A, B, and C states matrix

A = E_bar^-1 * A_bar;
B = E_bar^-1 * B_bar;
C = [[1 0 0 0 0 0];...
      [0 1 0 0 0 0];...
      [0 0 1 0 0 0];...
      [0 0 0 0 0 0];...
      [0 0 0 0 0 0];...
      [0 0 0 0 0 0]];...

D = [0 0 0 0 0 0]';

E = eye(6);

```

Solve for LTI Controller Gain K and Observer Gain L

```

% Solve algebraic raccatti K
Q = diag([1.2 1 1 1 80 80]);
R = 3;
[X,K_inf, nouseeig] = icare(A,B,Q,R);
%note, the K here is negative of the one from the notaion of NR22

% Solve for raccatti L using iCARE (use a-bk)
Q1 = eye(6);
Q2 = eye(6);
[P, L_inf, ect] = icare(A', C', Q1, Q2) ;

%note, the L here is negative of the one from the notaion of NR22

```

March States Starting from time T on the Estimated Simulated LQR States using RK4

```

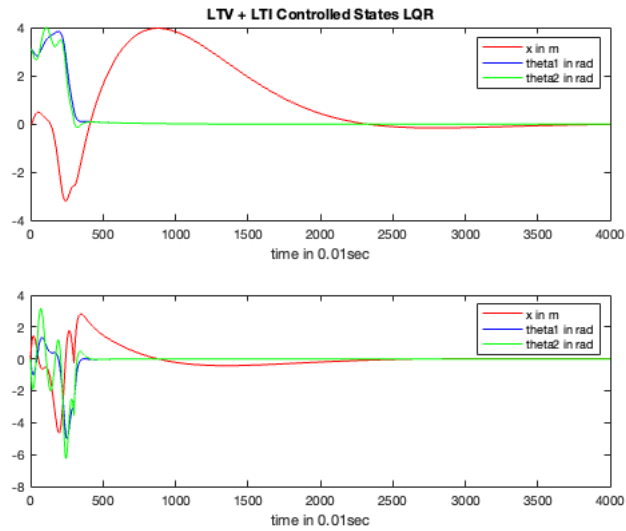
T_inf = 40;
for i = 301:1:T_inf/h

    f1 = (A-B*K_inf)*x_sim(:,i);
    f2 = (A-B*K_inf)*(x_sim(:,i) + f1*h/2) ;
    f3 = (A-B*K_inf)*(x_sim(:,i) + f2*h/2) ;
    f4 = (A-B*K_inf)*(x_sim(:,i) + f3*h) ;
    x_sim(:,i+1) = x_sim(:,i) + h*(f1/6 + f2/3 + f3/3 + f4/6);

end

t_inf = 0:T_inf/h;
figure(4); subplot(2,1,1); plot(t_inf,x_sim(1,:), 'r', t_inf,x_sim(2,:), 'b', t_inf,x_sim(3,:), 'g'); title("LTV + LTI Controlled States LQR")
    legend("x in m", "thetal in rad", "theta2 in rad"); xlabel("time in 0.01sec")
    subplot(2,1,2); plot(t_inf,x_sim(4,:), 'r', t_inf,x_sim(5,:), 'b', t_inf,x_sim(6,:), 'g');
    legend("x in m", "thetal in rad", "theta2 in rad"); xlabel("time in 0.01sec")

```



March States Starting from time T on the Estimated LQG States using RK4

```

for i = 301:1:T_inf/h

    f1 = (A-L_inf'*C)*x_estimation_sim(:,i) - B*K_inf*x_estimation_sim(:,i) + L_inf'*C*x_sim(:,i);
    f2 = (A-L_inf'*C)*x_estimation_sim(:,i) - B*K_inf*x_estimation_sim(:,i) + L_inf'*C*x_sim(:,i);
    f3 = (A-L_inf'*C)*x_estimation_sim(:,i) - B*K_inf*x_estimation_sim(:,i) + L_inf'*C*x_sim(:,i);
    f4 = (A-L_inf'*C)*x_estimation_sim(:,i) - B*K_inf*x_estimation_sim(:,i) + L_inf'*C*x_sim(:,i);

    x_estimation_sim(:,i+1) = x_estimation_sim(:,i) + h*(f1/6 + f2/3 + f3/3 + f4/6);

end

figure(5); subplot(2,1,1); plot(t_inf,x_estimation_sim(1,:), 'r', t_inf,x_estimation_sim(2,:), 'b', t_inf,x_estimation_sim(3,:), 'g'); title("LTV + LTI Con'
    legend("x in m", "theta1 in rad", "theta2 in rad"); xlabel("time in 0.01sec");
    subplot(2,1,2); plot(t_inf,x_estimation_sim(4,:), 'r', t_inf,x_estimation_sim(5,:), 'b', t_inf,x_estimation_sim(6,:), 'g');
    legend("x in m", "theta1 in rad", "theta2 in rad"); xlabel("time in 0.01sec");

```

