

Capstone Project [Small Scale Vehicle - Nigel]

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```
clear;  
close;  
clc;
```

System Modeling & Analysis

```
% System parameters
```

```
m = 2.68; % Mass [kg]  
J = 0.01944; % Yaw moment of inertia [kg-m^2]  
l = 0.1415; % Wheelbase [m]  
l_f = 0.44*l; % Distance from CG to front axel [m]  
l_r = l-l_f; % Distance from CG to rear axel [m]  
C_fl = 22.4768; % Cornering stiffness of FL tire [N/rad]  
C_fr = 22.4768; % Cornering stiffness of FR tire [N/rad]  
C_rl = 22.4768; % Cornering stiffness of RL tire [N/rad]  
C_rr = 22.4768; % Cornering stiffness of RR tire [N/rad]
```

```
% System phase portrait
```

```
u_fl = 0.4; % Coefficient of friction for ground and FL tire interconnect  
u_fr = 0.4; % Coefficient of friction for ground and FR tire interconnect  
u_rl = 0.4; % Coefficient of friction for ground and RL tire interconnect  
u_rr = 0.4; % Coefficient of friction for ground and RR tire interconnect  
v = 0.35; % Vehicle velocity [m/s]
```

```
% Compute system matrix
```

```
a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);  
a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr))) - 1;  
a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));  
a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) + l_r^2*(u_rl*C_rl+u_rr*C_rr));  
A = [a11 a12;  
     a21 a22];
```

```
% Define a range of initial conditions
```

```
X1_range = linspace(-3.14159, 3.14159, 20);  
X2_range = linspace(-6.28319, 6.28319, 20);
```

```
% Create a meshgrid of initial conditions
```

```
[X1, X2] = meshgrid(X1_range, X2_range);
```

```
% Initialize vectors to store the derivatives
```

```

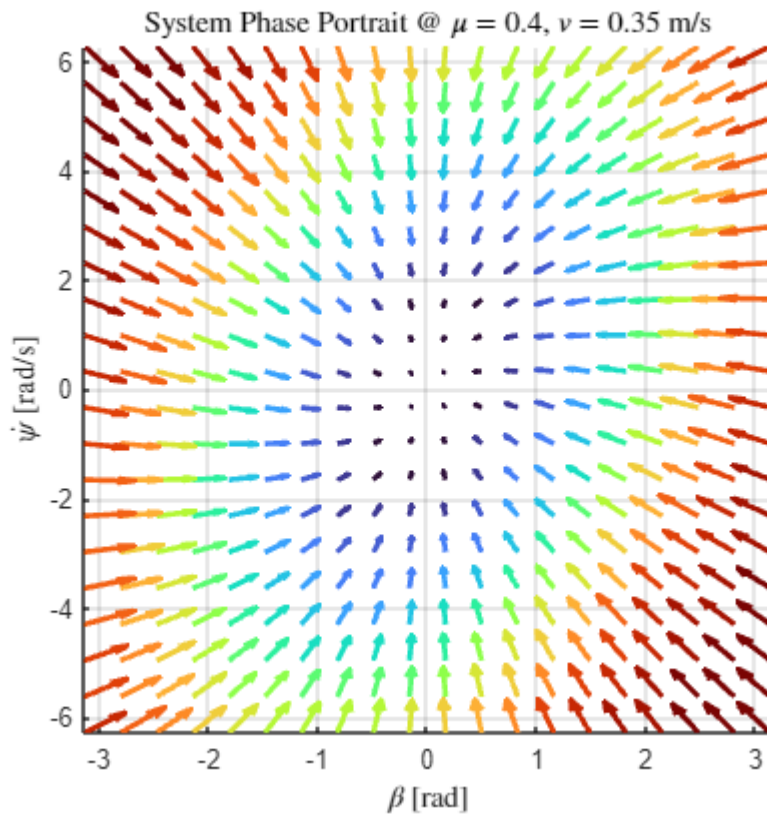
X1_dot = zeros(size(X1));
X2_dot = zeros(size(X2));

% Calculate the derivatives for each initial condition
for i = 1:numel(X1)
    X_dot = A * [X1(i); X2(i)];
    X1_dot(i) = X_dot(1);
    X2_dot(i) = X_dot(2);
end

Z = (sqrt(X1_dot.^2 + X2_dot.^2));
Z = Z/max(Z,[],'all');
Z = reshape(Z.',1,[]);
[sortedZ, sortedIdx] = sort(Z);
sortedIdx = reshape(sortedIdx,[20,20]).';
X1_s = X1(sortedIdx);
X2_s = X2(sortedIdx);
X1_dot_s = X1_dot(sortedIdx);
X2_dot_s = X2_dot(sortedIdx);

% Plot
fig = figure();
fig.Position(3:4) = [560, 420];
cmap = turbo(size(X1_s,1));
hold on;
for i = 1:size(X1_s,1)
    quiver(X1_s(i,:),X2_s(i,:),X1_dot_s(i,:),X2_dot_s(i,:), .2, 'Color', cmap(i,:),
'LineWidth',2)
end
hold off;
xlabel('X1');
ylabel('X2');
xlabel('$\beta$ [rad]','Interpreter','latex');
ylabel('$\dot{\psi}$ [rad/s]','Interpreter','latex');
title('System Phase Portrait @ $\mu = 0.4$, $v = 0.35$ m/s','Interpreter','latex');
grid on;
axis tight;
pbaspect([1 1 1])

```



```
% System eigenvalues with varying friction
```

```
EV = []; % Array to store system eigenvalues
```

```
v = 0.35; % Vehicle velocity [m/s]
```

```
for u=0.1:0.1:1
```

```
    % Set friction value
```

```
    u_fl = u;
```

```
    u_fr = u;
```

```
    u_rl = u;
```

```
    u_rr = u;
```

```
    % Update system matrix
```

```
    a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);
```

```
    a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr))) -
```

```
    1;
```

```
    a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));
```

```
    a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) + l_r^2*(u_rl*C_rl+u_rr*C_rr));
```

```
    A = [a11 a12;
```

```
        a21 a22];
```

```
    % Compute eigenvalues
```

```
    EV(:,end+1) = eig(A);
```

```
end
```

```
% Plot
```

```
u=0.1:0.1:1;
```

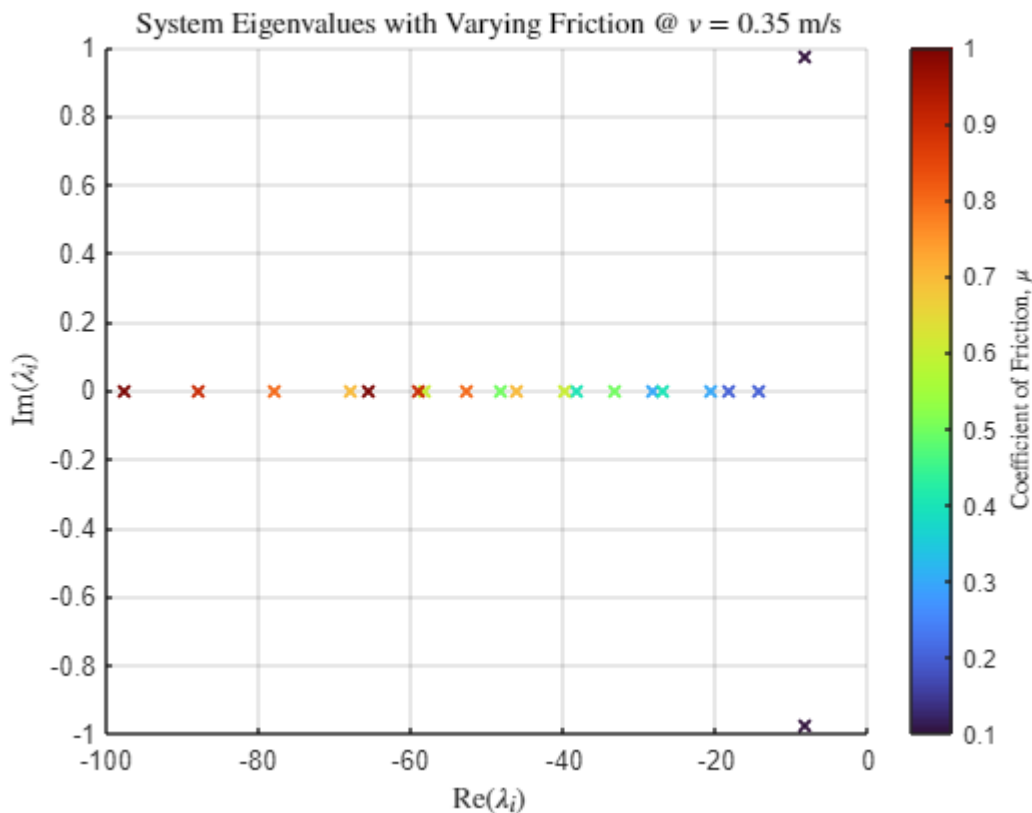
```
fig = figure();
```

```
fig.Position(3:4) = [560, 420];
```

```

scatter(real(EV(1,:)),imag(EV(1,:)),[],u,'LineWidth',1.5,'Marker','x')
hold on;
scatter(real(EV(2,:)),imag(EV(2,:)),[],u,'LineWidth',1.5,'Marker','x')
cb = colorbar();
ylabel(cb,'Coefficient of Friction,  $\mu$ ','Interpreter','latex');
colormap turbo;
hold off;
xlabel('Re $(\lambda_i)$ ','Interpreter','latex');
ylabel('Im $(\lambda_i)$ ','Interpreter','latex');
title('System Eigenvalues with Varying Friction @  $v = 0.35$  m/
s','Interpreter','latex');
grid on;

```



```

% System eigenvalues with varying velocity

```

```

EV = []; % Array to store system eigenvalues
u = 0.4; % Coefficient of friction for ground and tire interconnect
for v=0.05:0.05:5
    % Set friction value
    u_fl = u;
    u_fr = u;
    u_rl = u;
    u_rr = u;
    % Update system matrix
    a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);
    a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr))) -
1;

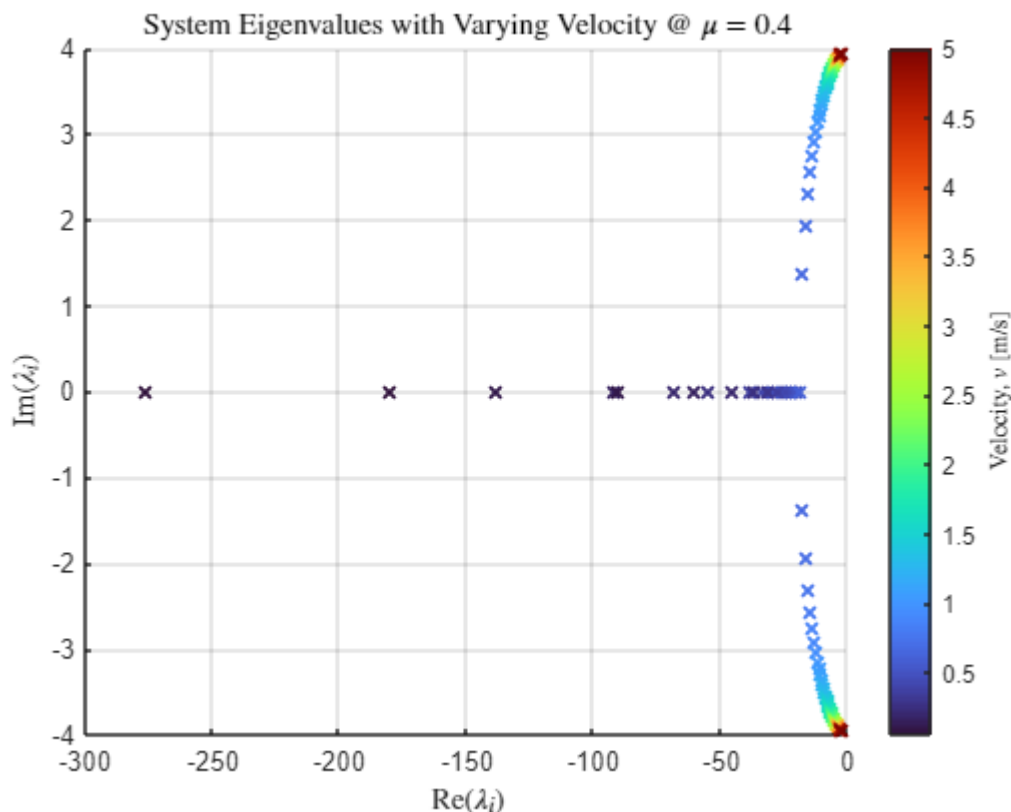
```

```

a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));
a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) + l_r^2*(u_rl*C_rl+u_rr*C_rr));
A = [a11 a12;
     a21 a22];
% Compute eigenvalues
EV(:,end+1) = eig(A);
end

% Plot
v=0.05:0.05:5;
fig = figure();
fig.Position(3:4) = [560, 420];
scatter(real(EV(1,:)),imag(EV(1,:)),[],v,'LineWidth',1.5,'Marker','x')
hold on;
scatter(real(EV(2,:)),imag(EV(2,:)),[],v,'LineWidth',1.5,'Marker','x')
cb = colorbar();
ylabel(cb,'Velocity, $v$ [m/s]','Interpreter','latex');
colormap turbo;
hold off;
xlabel('Re$(\lambda_i)$','Interpreter','latex');
ylabel('Im$(\lambda_i)$','Interpreter','latex');
title('System Eigenvalues with Varying Velocity @ $\mu = 0.4$','Interpreter','latex');
grid on;

```



```

% System damping ratio with varying friction and velocity

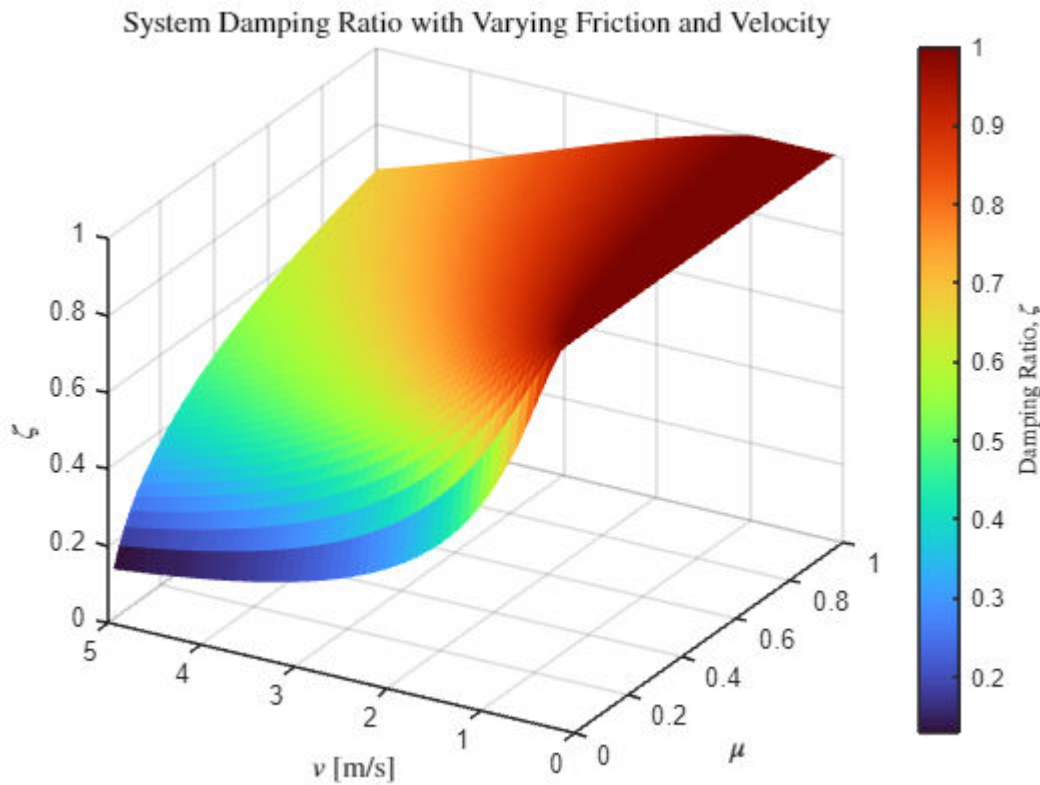
```

```

zeta = []; % Array to store system eigenvalues
for u=0.02:0.02:1
    for v=0.1:0.1:5
        % Set friction value
        u_fl = u;
        u_fr = u;
        u_rl = u;
        u_rr = u;
        % Update system matrix
        a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);
        a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) -
l_f*(u_fl*C_fl+u_fr*C_fr))) - 1;
        a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));
        a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) +
l_r^2*(u_rl*C_rl+u_rr*C_rr));
        A = [a11 a12;
            a21 a22];
        % Compute damping ratio
        EV = eig(A);
        zeta(:,end+1) = -real(EV)./abs(EV);
    end
end

% Plot
u=0.02:0.02:1;
v=0.1:0.1:5;
[X,Y] = meshgrid(u,v);
Z = reshape(zeta(1, :), 50, 50);
fig = figure();
fig.Position(3:4) = [560, 420];
s = surf(X,Y,Z,'FaceAlpha',1.0,'EdgeColor','none');
hold on;
cb = colorbar();
ylabel(cb,'Damping Ratio,  $\zeta$ ','Interpreter','latex');
colormap turbo;
hold off;
xlabel('$\mu$', 'Interpreter','latex');
ylabel('$v$ [m/s]', 'Interpreter','latex');
zlabel('$\zeta$', 'Interpreter','latex');
title('System Damping Ratio with Varying Friction and
Velocity', 'Interpreter','latex');
grid on;
view([-60 30]);

```



Model Simulation for Benchmark Maneuvers with Varying Friction

% Maneuver parameters

% Lane change test (impulse input)

```
u1_lane_change = [ones(1,14800)*(deg2rad(0)) ones(1,200)*(deg2rad(30))
ones(1,200)*(deg2rad(-30)) ones(1,14800)*(deg2rad(0))];
u2_lane_change = [ones(1,14800)*(deg2rad(0)) ones(1,200)*(deg2rad(30))
ones(1,200)*(deg2rad(-30)) ones(1,14800)*(deg2rad(0))];
u3_lane_change = [ones(1,14800)*(deg2rad(0)) ones(1,200)*(deg2rad(-30))
ones(1,200)*(deg2rad(30)) ones(1,14800)*(deg2rad(0))];
u4_lane_change = [ones(1,14800)*(deg2rad(0)) ones(1,200)*(deg2rad(-30))
ones(1,200)*(deg2rad(30)) ones(1,14800)*(deg2rad(0))];
u_lane_change = [u1_lane_change; u2_lane_change; u3_lane_change; u4_lane_change];
w_lane_change = zeros(1,30000);
```

% Skidpad test (step input)

```
u1_skidpad = [ones(1,2500)*(deg2rad(0)) ones(1,25000)*(deg2rad(30))
ones(1,2500)*(deg2rad(0))];
u2_skidpad = [ones(1,2500)*(deg2rad(0)) ones(1,25000)*(deg2rad(30))
ones(1,2500)*(deg2rad(0))];
u3_skidpad = [ones(1,2500)*(deg2rad(0)) ones(1,25000)*(deg2rad(-30))
ones(1,2500)*(deg2rad(0))];
u4_skidpad = [ones(1,2500)*(deg2rad(0)) ones(1,25000)*(deg2rad(-30))
ones(1,2500)*(deg2rad(0))];
```

```

u_skidpad = [u1_skidpad; u2_skidpad; u3_skidpad; u4_skidpad];
w_skidpad = zeros(1,30000);

% Fishhook test (ramp input)
angles = deg2rad(linspace(0,30,22500));
u1_fishhook = [ones(1,2500)*(deg2rad(0)) angles ones(1,2500)*deg2rad(30)
ones(1,2500)*(deg2rad(0))];
u2_fishhook = [ones(1,2500)*(deg2rad(0)) angles ones(1,2500)*deg2rad(30)
ones(1,2500)*(deg2rad(0))];
u3_fishhook = [ones(1,2500)*(deg2rad(0)) -angles ones(1,2500)*deg2rad(-30)
ones(1,2500)*(deg2rad(0))];
u4_fishhook = [ones(1,2500)*(deg2rad(0)) -angles ones(1,2500)*deg2rad(-30)
ones(1,2500)*(deg2rad(0))];
u_fishhook = [u1_fishhook; u2_fishhook; u3_fishhook; u4_fishhook];
w_fishhook = zeros(1,30000);

% Slalom test (sine input)
n_cones = 30;
angles = linspace(0,n_cones*pi,25000);
u1_slalom = [ones(1,2500)*(deg2rad(0)) deg2rad(30*cos(angles))
ones(1,2500)*(deg2rad(0))];
u2_slalom = [ones(1,2500)*(deg2rad(0)) deg2rad(30*cos(angles))
ones(1,2500)*(deg2rad(0))];
u3_slalom = [ones(1,2500)*(deg2rad(0)) deg2rad(-30*cos(angles))
ones(1,2500)*(deg2rad(0))];
u4_slalom = [ones(1,2500)*(deg2rad(0)) deg2rad(-30*cos(angles))
ones(1,2500)*(deg2rad(0))];
u_slalom = [u1_slalom; u2_slalom; u3_slalom; u4_slalom];
w_slalom = zeros(1,30000);

% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [4000, 2000];
% Lane change (impulse)
subplot(2,2,1)
plot(t/1000,u1_lane_change)
hold on;
plot(t/1000,u2_lane_change)
plot(t/1000,u3_lane_change)
plot(t/1000,u4_lane_change)
legend('$\delta_{fl}$','$\delta_{fr}$','$\delta_{rl}$','$\delta_{rr}$',
'$','Interpreter','latex','Location','northwest')
xlabel('t [s]','Interpreter','latex')
ylabel('$u(t)$ [rad]','Interpreter','latex')
title('Lane Change Test','Interpreter','latex')
grid on;
hold off;
% Skidpad (step)
subplot(2,2,2)

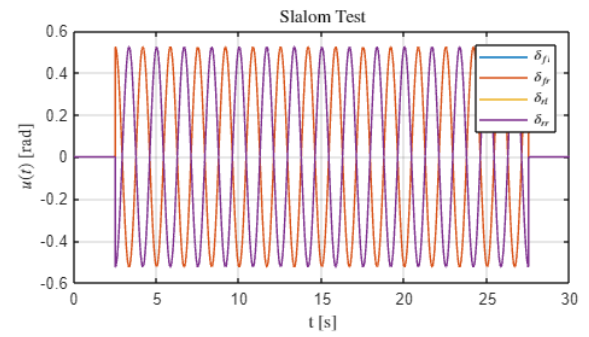
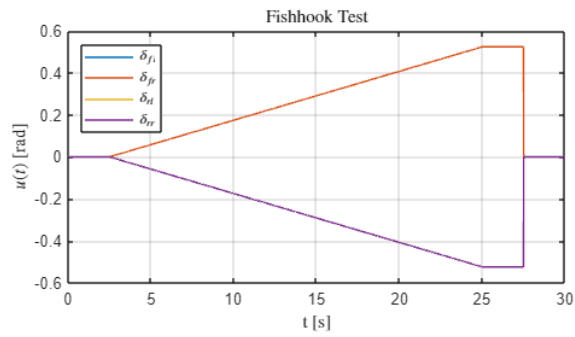
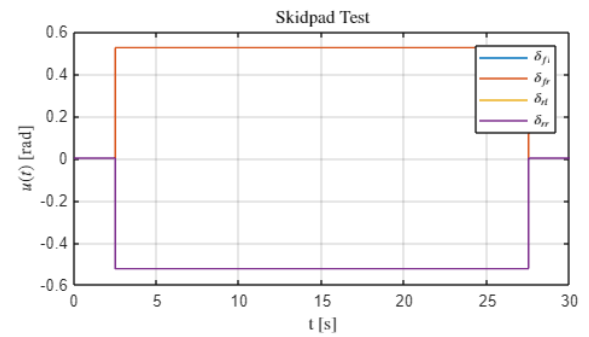
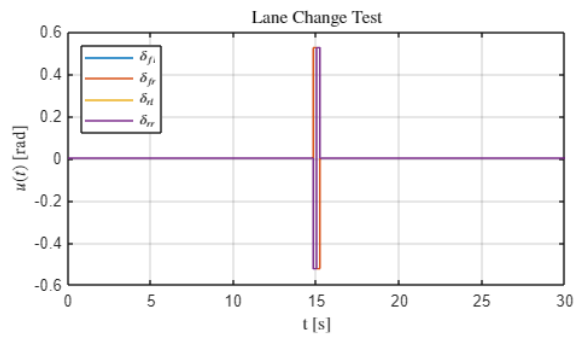
```



```

plot(t/1000,u1_skidpad)
hold on;
plot(t/1000,u2_skidpad)
plot(t/1000,u3_skidpad)
plot(t/1000,u4_skidpad)
legend('$\delta_{fl}$','$\delta_{fr}$','$\delta_{rl}$','$\delta_{rr}$',
'$','Interpreter','latex','Location','northeast')
xlabel('t [s]','Interpreter','latex')
ylabel('$u(t)$ [rad]','Interpreter','latex')
title('Skidpad Test','Interpreter','latex')
grid on;
hold off;
% Fishhook (ramp)
subplot(2,2,3)
plot(t/1000,u1_fishhook)
hold on;
plot(t/1000,u2_fishhook)
plot(t/1000,u3_fishhook)
plot(t/1000,u4_fishhook)
legend('$\delta_{fl}$','$\delta_{fr}$','$\delta_{rl}$','$\delta_{rr}$',
'$','Interpreter','latex','Location','northwest')
xlabel('t [s]','Interpreter','latex')
ylabel('$u(t)$ [rad]','Interpreter','latex')
title('Fishhook Test','Interpreter','latex')
grid on;
hold off;
% Slalom (sine)
subplot(2,2,4)
plot(t/1000,u1_slalom)
hold on;
plot(t/1000,u2_slalom)
plot(t/1000,u3_slalom)
plot(t/1000,u4_slalom)
legend('$\delta_{fl}$','$\delta_{fr}$','$\delta_{rl}$','$\delta_{rr}$',
'$','Interpreter','latex','Location','northeast')
xlabel('t [s]','Interpreter','latex')
ylabel('$u(t)$ [rad]','Interpreter','latex')
title('Slalom Test','Interpreter','latex')
grid on;
hold off;

```



% Simulate lane change tests with varying friction

```
BETA_LC = [];
BETA_DOT_LC = [];
PSI_LC = [];
PSI_DOT_LC = [];
PX_LC = [];
VX_LC = [];
PY_LC = [];
VY_LC = [];
```

```
for u=0.2:0.2:1
```

```
% Initialization
```

```
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;
```

```

% Set friction value
u_fl = u;
u_fr = u;
u_rl = u;
u_rr = u;

% Update system matrix
a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);
a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr))) -
1;
a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));
a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) + l_r^2*(u_rl*C_rl+u_rr*C_rr));
A = [a11 a12;
      a21 a22];

% Update control input matrix
b11 = (u_fl*C_fl)/(m*v);
b12 = (u_fr*C_fr)/(m*v);
b13 = (u_rl*C_rl)/(m*v);
b14 = (u_rr*C_rr)/(m*v);
b21 = (l_f*u_fl*C_fl)/J;
b22 = (l_f*u_fr*C_fr)/J;
b23 = -(l_r*u_rl*C_rl)/J;
b24 = -(l_r*u_rr*C_rr)/J;
B = [b11 b12 b13 b14;
      b21 b22 b23 b24];

% Update disturbance input matrix
B_d = [1/(m*v);
        (l_f-l_r)/2];

% Simulation at each millisecond
for t=1:30000
    X_dot(:,t) = A*X(:,t) + B*u_lane_change(:,t) +
B_d*(w_lane_change(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_LC(end+1,:) = Beta;
BETA_DOT_LC(end+1,:) = X_dot(1,:);
PSI_LC(end+1,:) = Psi;
PSI_DOT_LC(end+1,:) = X(2,:);
PX_LC(end+1,:) = Px;
VX_LC(end+1,:) = Vx;

```

```

    PY_LC(end+1,:) = Py;
    VY_LC(end+1,:) = Vy;
end

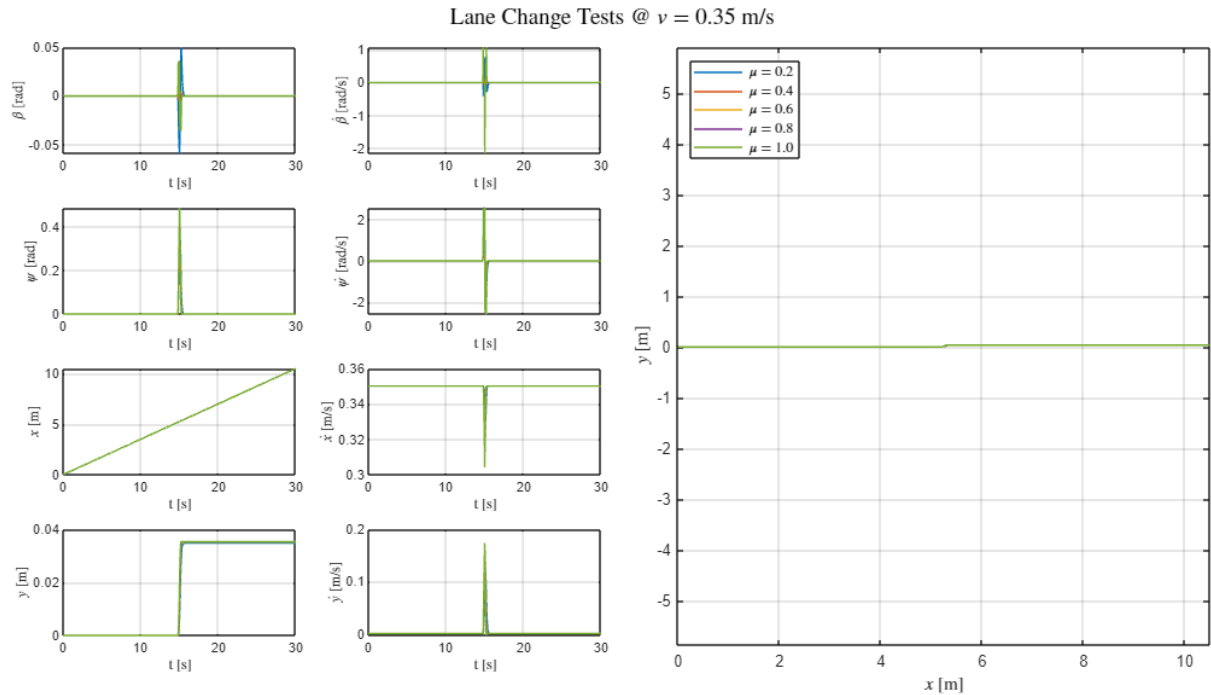
% Plot
t=1:30000;
u=0.2:0.2:1;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_LC(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_LC(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_LC(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_LC(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_LC(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_LC(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_LC(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)
plot(t/1000,VY_LC(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')

```

```

% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_LC(:,1:end-1)',PY_LC(:,1:end-1)')
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Lane Change Tests @ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('$\mu = 0.2$', '$\mu = 0.4$', '$\mu = 0.6$', '$\mu = 0.8$', '$\mu = 1.0$', ...
    'Interpreter', 'latex', 'Location', 'northwest')

```



```

% Simulate skidpad tests with varying friction

```

```

BETA_SP = [];
BETA_DOT_SP = [];
PSI_SP = [];
PSI_DOT_SP = [];
PX_SP = [];
VX_SP = [];
PY_SP = [];
VY_SP = [];

for u=0.2:0.2:1

    % Initialization
    dt = 0.001; % Simulation timestep [s]
    v = 0.35; % Vehicle velocity [m/s]
    X = [];

```

```

X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Set friction value
u_fl = u;
u_fr = u;
u_rl = u;
u_rr = u;

% Update system matrix
a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);
a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr))) -
1;
a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));
a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) + l_r^2*(u_rl*C_rl+u_rr*C_rr));
A = [a11 a12;
     a21 a22];

% Update control input matrix
b11 = (u_fl*C_fl)/(m*v);
b12 = (u_fr*C_fr)/(m*v);
b13 = (u_rl*C_rl)/(m*v);
b14 = (u_rr*C_rr)/(m*v);
b21 = (l_f*u_fl*C_fl)/J;
b22 = (l_f*u_fr*C_fr)/J;
b23 = -(l_r*u_rl*C_rl)/J;
b24 = -(l_r*u_rr*C_rr)/J;
B = [b11 b12 b13 b14;
     b21 b22 b23 b24];

% Update disturbance input matrix
B_d = [1/(m*v);
       (l_f-l_r)/2];

% Simulation at each millisecond
for t=1:30000
    X_dot(:,t) = A*X(:,t) + B*u_skidpad(:,t) +
B_d*(w_skidpad(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;

```

```

        Vx(t) = v*cos(Psi(t) + Beta(t));
        Vy(t) = v*sin(Psi(t) + Beta(t));
        Px(t+1) = Px(t) + Vx(t)*dt;
        Py(t+1) = Py(t) + Vy(t)*dt;
    end
    BETA_SP(end+1,:) = Beta;
    BETA_DOT_SP(end+1,:) = X_dot(1,:);
    PSI_SP(end+1,:) = Psi;
    PSI_DOT_SP(end+1,:) = X(2,:);
    PX_SP(end+1,:) = Px;
    VX_SP(end+1,:) = Vx;
    PY_SP(end+1,:) = Py;
    VY_SP(end+1,:) = Vy;
end

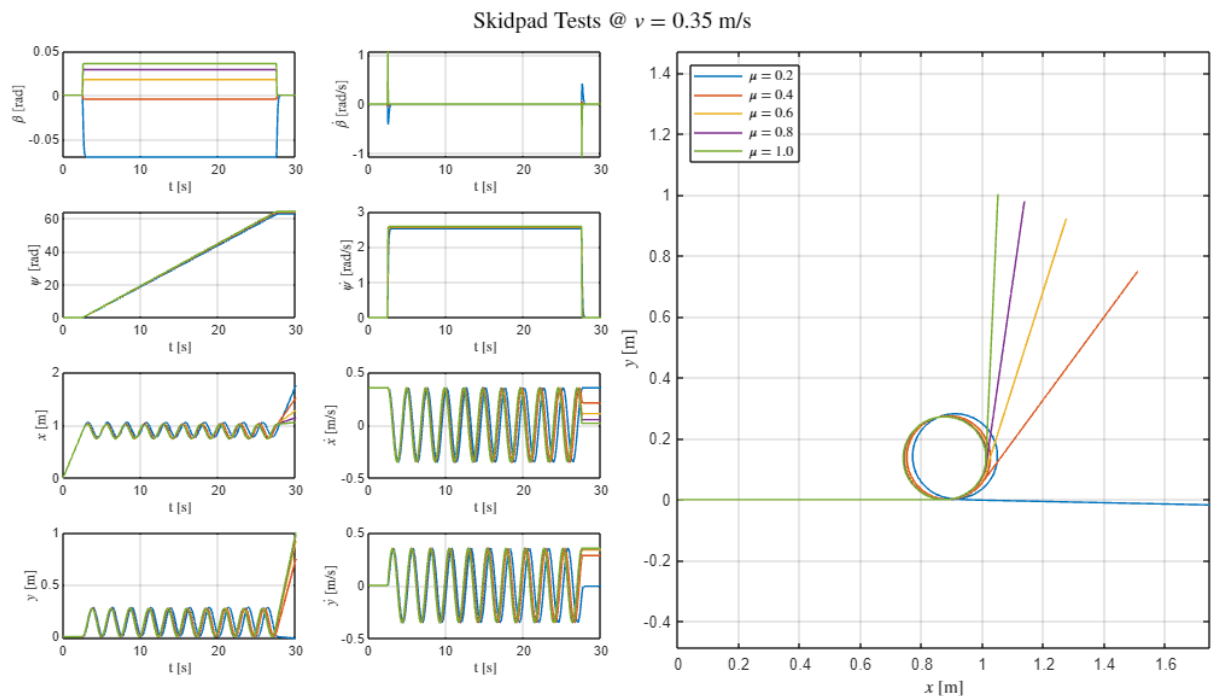
% Plot
t=1:30000;
u=0.2:0.2:1;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_SP(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_SP(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_SP(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_SP(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_SP(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_SP(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')

```

```

grid('on')
subplot(4,4,13)
plot(t/1000,PY_SP(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)
plot(t/1000,VY_SP(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_SP(:,1:end-1),PY_SP(:,1:end-1))
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Skidpad Tests @ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('$\mu = 0.2$', '$\mu = 0.4$', '$\mu = 0.6$', '$\mu = 0.8$', '$\mu = 1.0$', ...
    'Interpreter', 'latex', 'Location', 'northwest')

```



```

% Simulate fishhook tests with varying friction

```

```

BETA_FH = [];
BETA_DOT_FH = [];
PSI_FH = [];
PSI_DOT_FH = [];

```



```

PX_FH = [];
VX_FH = [];
PY_FH = [];
VY_FH = [];

for u=0.2:0.2:1

    % Initialization
    dt = 0.001; % Simulation timestep [s]
    v = 0.35; % Vehicle velocity [m/s]
    X = [];
    X(:,1) = [0;0];
    X_dot = [];
    Psi = [];
    Psi(1) = 0;
    Beta = [];
    Vx = [];
    Vy = [];
    Px = [];
    Px(1) = 0;
    Py = [];
    Py(1) = 0;

    % Set friction value
    u_fl = u;
    u_fr = u;
    u_rl = u;
    u_rr = u;

    % Update system matrix
    a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);
    a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr))) -
1;
    a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));
    a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) + l_r^2*(u_rl*C_rl+u_rr*C_rr));
    A = [a11 a12;
        a21 a22];

    % Update control input matrix
    b11 = (u_fl*C_fl)/(m*v);
    b12 = (u_fr*C_fr)/(m*v);
    b13 = (u_rl*C_rl)/(m*v);
    b14 = (u_rr*C_rr)/(m*v);
    b21 = (l_f*u_fl*C_fl)/J;
    b22 = (l_f*u_fr*C_fr)/J;
    b23 = -(l_r*u_rl*C_rl)/J;
    b24 = -(l_r*u_rr*C_rr)/J;
    B = [b11 b12 b13 b14;
        b21 b22 b23 b24];

```

```

% Update disturbance input matrix
B_d = [1/(m*v);
       (l_f-l_r)/2];

% Simulation at each millisecond
for t=1:30000
    X_dot(:,t) = A*X(:,t) + B*u_fishhook(:,t) +
B_d*(w_fishhook(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_FH(end+1,:) = Beta;
BETA_DOT_FH(end+1,:) = X_dot(1,:);
PSI_FH(end+1,:) = Psi;
PSI_DOT_FH(end+1,:) = X(2,:);
PX_FH(end+1,:) = Px;
VX_FH(end+1,:) = Vx;
PY_FH(end+1,:) = Py;
VY_FH(end+1,:) = Vy;
end

% Plot
t=1:30000;
u=0.2:0.2:1;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_FH(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_FH(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_FH(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_FH(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')

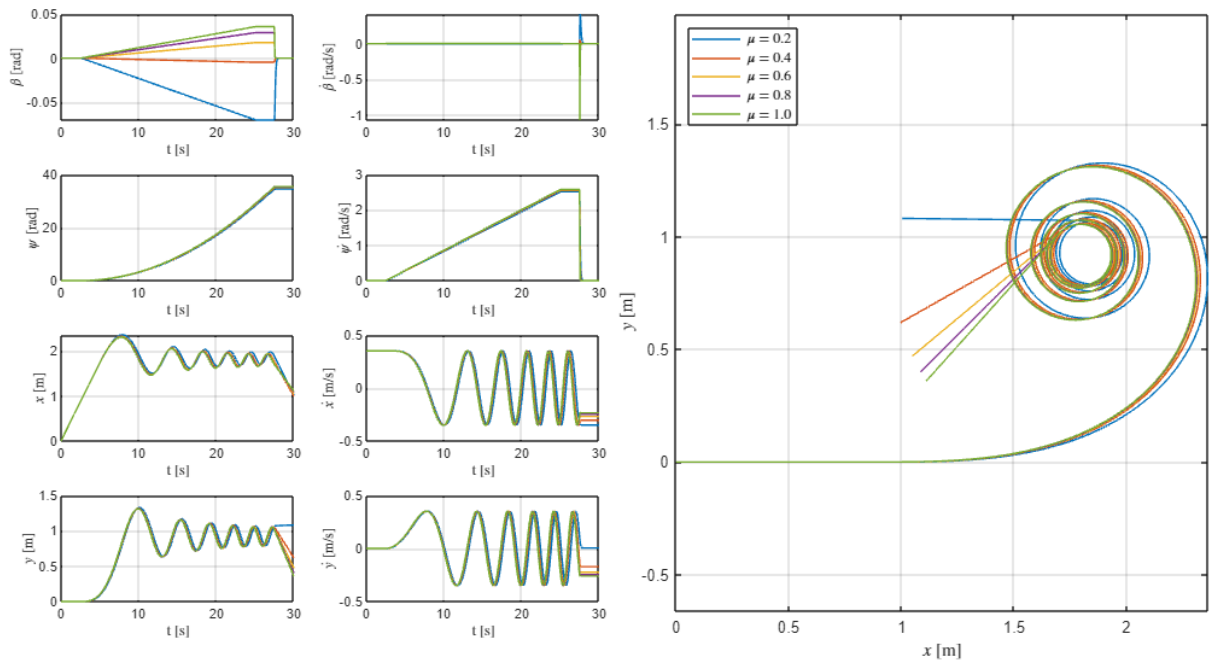
```

```

ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_FH(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_FH(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_FH(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)
plot(t/1000,VY_FH(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_FH(:,1:end-1),PY_FH(:,1:end-1))
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Fishhook Tests @ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('$\mu = 0.2$', '$\mu = 0.4$', '$\mu = 0.6$', '$\mu = 0.8$', '$\mu = 1.0$', ...
    'Interpreter', 'latex', 'Location', 'northwest')

```

Fishhook Tests @ $v = 0.35$ m/s



% Simulate slalom tests with varying friction

```
BETA_SL = [];
BETA_DOT_SL = [];
PSI_SL = [];
PSI_DOT_SL = [];
PX_SL = [];
VX_SL = [];
PY_SL = [];
VY_SL = [];
```

```
for u=0.2:0.2:1
```

```
    % Initialization
```

```
    dt = 0.001; % Simulation timestep [s]
```

```
    v = 0.35; % Vehicle velocity [m/s]
```

```
    X = [];
```

```
    X(:,1) = [0;0];
```

```
    X_dot = [];
```

```
    Psi = [];
```

```
    Psi(1) = 0;
```

```
    Beta = [];
```

```
    Vx = [];
```

```
    Vy = [];
```

```
    Px = [];
```

```
    Px(1) = 0;
```

```
    Py = [];
```

```
    Py(1) = 0;
```

```

% Set friction value
u_fl = u;
u_fr = u;
u_rl = u;
u_rr = u;

% Update system matrix
a11 = -(1/(m*v)) * (u_fl*C_fl+u_fr*C_fr+u_rl*C_rl+u_rr*C_rr);
a12 = ((1/(m*v^2)) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr))) -
1;
a21 = (1/J) * (l_r*(u_rl*C_rl+u_rr*C_rr) - l_f*(u_fl*C_fl+u_fr*C_fr));
a22 = -(1/(J*v)) * (l_f^2*(u_fl*C_fl+u_fr*C_fr) + l_r^2*(u_rl*C_rl+u_rr*C_rr));
A = [a11 a12;
      a21 a22];

% Update control input matrix
b11 = (u_fl*C_fl)/(m*v);
b12 = (u_fr*C_fr)/(m*v);
b13 = (u_rl*C_rl)/(m*v);
b14 = (u_rr*C_rr)/(m*v);
b21 = (l_f*u_fl*C_fl)/J;
b22 = (l_f*u_fr*C_fr)/J;
b23 = -(l_r*u_rl*C_rl)/J;
b24 = -(l_r*u_rr*C_rr)/J;
B = [b11 b12 b13 b14;
      b21 b22 b23 b24];

% Update disturbance input matrix
B_d = [1/(m*v);
        (l_f-l_r)/2];

% Simulation at each millisecond
for t=1:30000
    X_dot(:,t) = A*X(:,t) + B*u_slalom(:,t) + B_d*(w_slalom(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_SL(end+1,:) = Beta;
BETA_DOT_SL(end+1,:) = X_dot(1,:);
PSI_SL(end+1,:) = Psi;
PSI_DOT_SL(end+1,:) = X(2,:);
PX_SL(end+1,:) = Px;
VX_SL(end+1,:) = Vx;
PY_SL(end+1,:) = Py;

```

```

    VY_SL(end+1,:) = Vy;
end

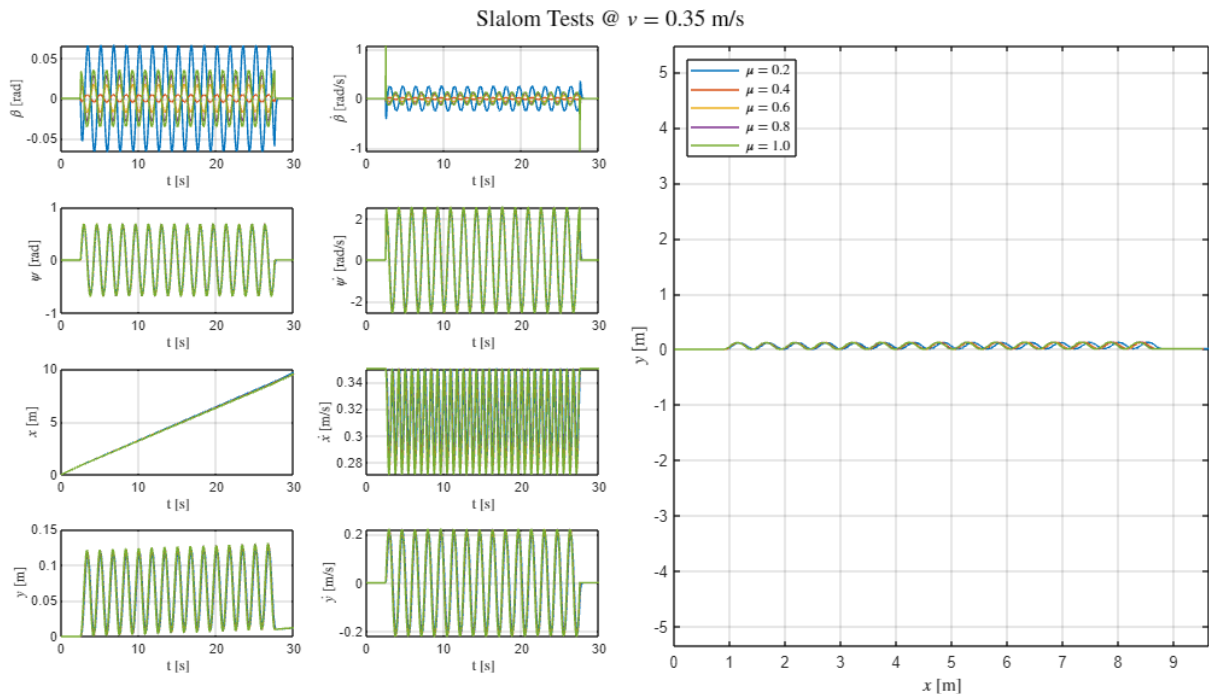
% Plot
t=1:30000;
u=0.2:0.2:1;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_SL(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_SL(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_SL(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_SL(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_SL(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_SL(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_SL(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)
plot(t/1000,VY_SL(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory

```

```

subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_SL(:,1:end-1)',PY_SL(:,1:end-1)')
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Slalom Tests @ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('$\mu = 0.2$', '$\mu = 0.4$', '$\mu = 0.6$', '$\mu = 0.8$', '$\mu = 1.0$', ...
    'Interpreter', 'latex', 'Location', 'northwest')

```



Uncertain Parameter Dependent System

```

% Operating conditions

v = 0.35; % Vehicle velocity [m/s]

% Vehicle model with affine dependence on u_fl, u_fr, u_rl and u_rr

Ap0 = [0 -1;
        0 0]; % Nominal system matrix

Ap1 = [-C_fl/(m*v)    -(l_f*C_fl)/(m*v^2);
        -(l_f*C_fl)/J  -(l_f^2*C_fl)/(J*v)]; % Affine dependence on u_fl

Ap2 = [-C_fr/(m*v)    -(l_f*C_fr)/(m*v^2);
        -(l_f*C_fr)/J  -(l_f^2*C_fr)/(J*v)]; % Affine dependence on u_fr

Ap3 = [-C_rl/(m*v)    (l_r*C_rl)/(m*v^2);

```

```

    (l_r*C_rl)/J    -(l_r^2*C_rl)/(J*v)]; % Affine dependence on u_rl

Ap4 = [-C_rr/(m*v)    (l_r*C_rr)/(m*v^2);
    (l_r*C_rr)/J    -(l_r^2*C_rr)/(J*v)]; % Affine dependence on u_rr

% Ap = Ap0 + u_fl.*Ap1 + u_fr.*Ap2 + u_rl.*Ap3 + u_rr.*Ap4

Bp0 = [0 0 0 0;
    0 0 0 0]; % Nominal input matrix

Bp1 = [C_fl/(m*v)    0 0 0;
    (l_f*C_fl)/J 0 0 0]; % Affine dependence on u_fl

Bp2 = [0 C_fr/(m*v)    0 0;
    0 (l_f*C_fr)/J 0 0]; % Affine dependence on u_fr

Bp3 = [0 0 C_rl/(m*v)    0;
    0 0 -(l_r*C_rl)/J 0]; % Affine dependence on u_rl

Bp4 = [0 0 0 C_rr/(m*v);
    0 0 0 -(l_r*C_rr)/J]; % Affine dependence on u_rr

% Bp = Bp0 + u_fl.*Bp1 + u_fr.*Bp2 + u_rl.*Bp3 + u_rr.*Bp4

Dp = [1/(m*v); (l_f-l_r)/2];

Cp1 = [1 0;
    0 1;
    0 0;
    0 0;
    0 0;
    0 0];

By1 = [0 0 0 0;
    0 0 0 0;
    1 0 0 0;
    0 1 0 0;
    0 0 1 0;
    0 0 0 1];

Dy = [0;
    0;
    0;
    0;
    0;
    0];

Cp2 = [1 0;
    0 1;
    0 0];

```



```

    0 0;
    0 0;
    0 0];

By2 = [0 0 0 0;
       0 0 0 0;
       1 0 0 0;
       0 1 0 0;
       0 0 1 0;
       0 0 0 1];

O = [0;
     0;
     0;
     0;
     0;
     0];

[u_fl_lim, u_fr_lim, u_rl_lim, u_rr_lim] = deal([0.1 1], [0.1 1], [0.1 1], [0.1 1]);

pv = pvec('box', [u_fl_lim; u_fr_lim; u_rl_lim; u_rr_lim]); % Range of parameter
variation
pvinfo(pv); % Print uncertain parameter(s) information

```

Vector of 4 parameters ranging in a box

Open-Loop System

```

S_ol = psys(pv, [ltisys(Ap0, [Dp Bp0], [Cp1; Cp2], [Dy By1; 0 By2], 1), ...
               ltisys(Ap1, [zeros(2,1) Bp1], 0*[Cp1; Cp2], 0*[Dy By1; 0 By2], 0),
               ...
               ltisys(Ap2, [zeros(2,1) Bp2], 0*[Cp1; Cp2], 0*[Dy By1; 0 By2], 0),
               ...
               ltisys(Ap3, [zeros(2,1) Bp3], 0*[Cp1; Cp2], 0*[Dy By1; 0 By2], 0),
               ...
               ltisys(Ap4, [zeros(2,1) Bp4], 0*[Cp1; Cp2], 0*[Dy By1; 0 By2],
0))];

psinfo(S_ol); % Print open-loop system information

```

Affine parameter-dependent model with 4 parameters (5 systems)
Each system has 2 state(s), 5 input(s), and 12 output(s)

```

% Test robust stability of open-loop system via parametric Lyapunov functions
[tmin_ol, Q0_ol, Q1_ol, Q2_ol, Q3_ol] = pdlstab(S_ol, [0 0])

```

Solver for LMI feasibility problems $L(x) < R(x)$
This solver minimizes t subject to $L(x) < R(x) + t*I$
The best value of t should be negative for feasibility

Iteration : Best value of t so far

1	0.109182
2	0.011921
3	0.011921
4	7.406228e-03
5	-2.473266e-03

Result: best value of t: -2.473266e-03
 f-radius saturation: 0.000% of R = 1.00e+07

This system is stable for the specified parameter trajectories

tmin_ol = -0.0025

Q0_ol = 2x2

0.0130	0.0069
0.0069	0.1507

Q1_ol = 2x2

-0.0019	-0.0093
-0.0093	-0.0419

Q2_ol = 2x2

-0.0019	-0.0093
-0.0093	-0.0419

Q3_ol = 2x2

-0.0015	0.0064
0.0064	-0.0335

% Friction variation and disturbance

rng(1,"twister"); % Fix random seed and algorithm

% Friction

```
u1_signal = ones(1,30000)*0.8;
u2_signal = ones(1,30000)*0.8;
u3_signal = ones(1,30000)*0.2;
u4_signal = ones(1,30000)*0.2;
u1_signal = u1_signal - 0.05 + 0.1 * rand(1, length(u1_signal));
u2_signal = u2_signal - 0.05 + 0.1 * rand(1, length(u2_signal));
u3_signal = u3_signal - 0.05 + 0.1 * rand(1, length(u3_signal));
u4_signal = u4_signal - 0.05 + 0.1 * rand(1, length(u4_signal));
```

% Disturbance

```
w_signal = [zeros(1,7500) 0.25*ones(1,15000) zeros(1,7500)];
w_signal = w_signal - 0.05 + 0.1*(max(w_signal)) * rand(1, length(w_signal));
```

% Plot

```
t=1:30000;
fig = figure();
fig.Position(3:4) = [4000, 1000];
```

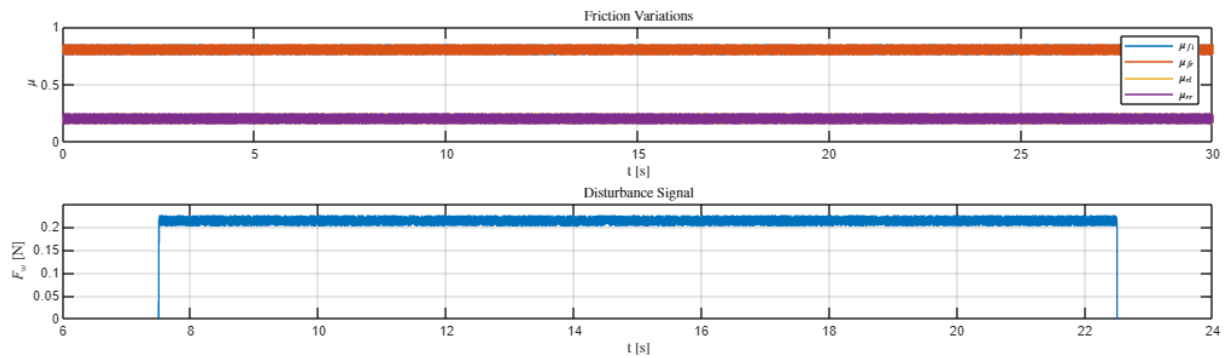
% Friction

```
subplot(2,1,1)
plot(t/1000,u1_signal)
hold on;
plot(t/1000,u2_signal)
plot(t/1000,u3_signal)
plot(t/1000,u4_signal)
legend('$\mu_{fl}$','$\mu_{fr}$','$\mu_{rl}$','$\mu_{rr}$',
'$','Interpreter','latex','Location','northeast')
```

```

xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\mu$', 'Interpreter', 'latex')
ylim([0,1]);
title('Friction Variations', 'Interpreter', 'latex')
grid on;
hold off;
% Disturbance
subplot(2,1,2)
plot(t/1000,w_signal)
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$F_w$ [N]', 'Interpreter', 'latex')
ylim([0,0.25]);
title('Disturbance Signal', 'Interpreter', 'latex')
grid on;

```



% Simulate straight-line motion with varying friction and disturbance

% Initialization

```

dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

```

% Simulation at each millisecond

```

for t=1:30000
    system_ol = psinfo(S_ol, 'eval', [u1_signal(t) u2_signal(t) u3_signal(t)
    u4_signal(t)]);
    [A_ol, B_ol, C_ol, D_ol] = ltiss(system_ol);
    X_dot(:,t) = A_ol*X(:,t) + B_ol*[(w_signal(t).*cos(Psi(t)))]; [0;0;0;0];

```

```

X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
Beta(t) = X(1,t);
Psi(t+1) = Psi(t) + X(2,t)*dt;
Vx(t) = v*cos(Psi(t) + Beta(t));
Vy(t) = v*sin(Psi(t) + Beta(t));
Px(t+1) = Px(t) + Vx(t)*dt;
Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA = Beta;
BETA_DOT = X_dot(1,:);
PSI = Psi;
PSI_DOT = X(2,:);
PX = Px;
VX = Vx;
PY = Py;
VY = Vy;

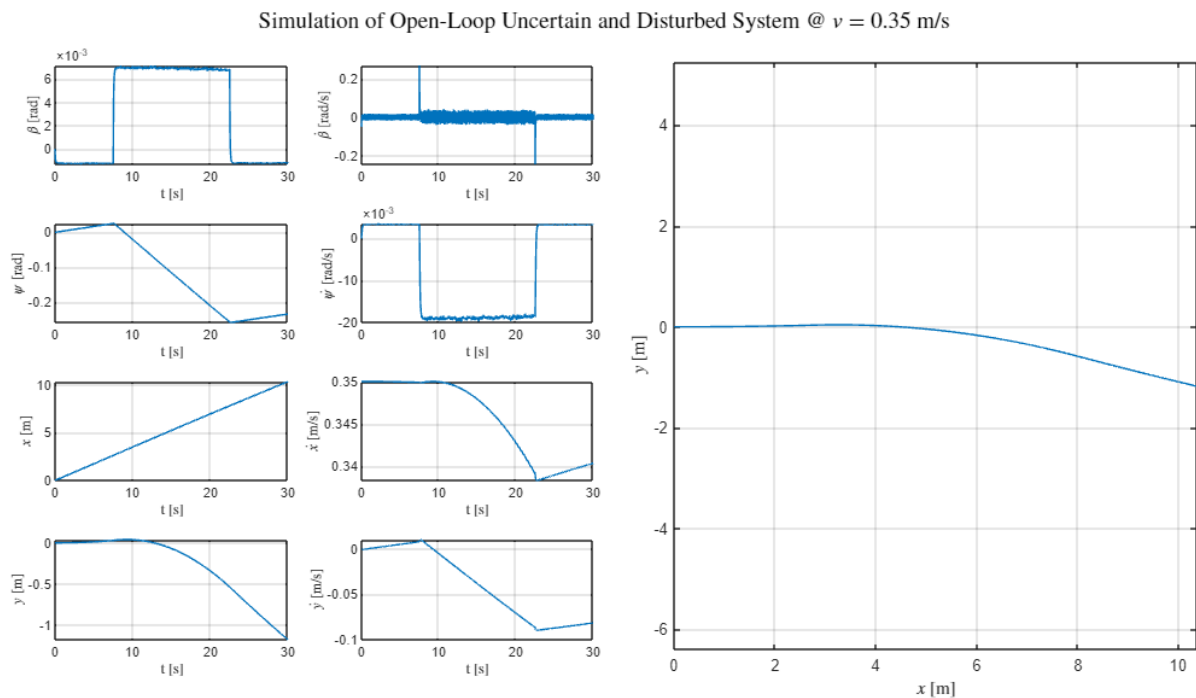
% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')

```

```

ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000, PY(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)
plot(t/1000, VY(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX(:,1:end-1), PY(:,1:end-1))
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Simulation of Open-Loop Uncertain and Disturbed System @ $v = 0.35$ m/
s', 'Interpreter', 'latex')

```



Controller Design

```

% region = lmireg; % Pole placement constraint (intersection of the left half-plane
% with  $x < -0.1$  and of the sector centered at the origin and with inner angle  $3\pi/4$ )
region = [0.2000 + 1.0000i    0.0000 + 0.0000i    0.0000 + 0.0000i    1.0000 +
0.0000i    0.0000 + 0.0000i    0.0000 + 0.0000i];

```

```

0.0000 + 0.0000i  0.0000 + 2.0000i  0.0000 + 0.0000i  0.0000 +
0.0000i  0.9239 + 0.0000i  -0.3827 + 0.0000i;
0.0000 + 0.0000i  0.0000 + 0.0000i  0.0000 + 0.0000i  0.0000 +
0.0000i  0.3827 + 0.0000i  0.9239 + 0.0000i];

```

```

g1_opt_global = msfsyn(S_ol, size(By2), [0 0 1 0], region) % Optimal quadratic H∞
performance subject to the pole placement constraint

```

Optimization of the H-infinity performance G :

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5          64.616755
6          5.458629
7          5.458629
8          2.721752
9          2.142491
10         1.533455
11         1.052437
12         0.809563
13         0.809563
14         0.289264
15         0.246906
16         0.246906
17         0.200706
18         0.200706
19         0.200706
20         0.171254
21         0.171254
22         0.152607
23         0.152607
24         0.152607
25         0.133582
26         0.133582
27         0.131517
28         0.131517
***      new lower bound:      0.049831
29         0.122336
***      new lower bound:      0.050383
30         0.122336
***      new lower bound:      0.055419
31         0.122336
32         0.122052
33         0.122052
***      new lower bound:      0.075447
34         0.122052
***      new lower bound:      0.075735
35         0.122052
36         0.118958
37         0.118958
***      new lower bound:      0.079319
38         0.118596
***      new lower bound:      0.079973
39         0.118596
***      new lower bound:      0.092650

```

```

40          0.118596
***          new lower bound:      0.093388
41          0.117982
42          0.117982
***          new lower bound:      0.098633
43          0.117695
***          new lower bound:      0.099132
44          0.117695
***          new lower bound:      0.107503
45          0.117695
***          new lower bound:      0.107642
46          0.117492
47          0.117492
***          new lower bound:      0.111006
48          0.117492
***          new lower bound:      0.111161

```

```

Result: feasible solution
best objective value:      0.117492
guaranteed absolute accuracy: 6.33e-03
f-radius saturation: 0.000% of R = 1.00e+10
Termination due to SLOW PROGRESS:
the objective was decreased by less than
1.000% during the last 10 iterations.

```

```

Guaranteed Hinf performance: 1.17e-01
g1_opt_global = 0.1175

```

```

g1_opt = []; % Array to store  $\gamma_1^*$  ( $\Gamma_{ee}$ )
g2_opt = []; % Array to store  $\gamma_2^*$  ( $\Gamma_{ep}$ )
K = []; % Array to store state-feedback controller gain K
S_cl = []; % Array to store closed-loop system representations
X = []; % Array to store corresponding Lyapunov matrices

for g = g1_opt_global:0.01:0.5 % Bound on  $H_\infty$  performance
    [g1_opt(end+1), g2_opt(end+1), K(:, :, end+1), S_cl(:, :, end+1), X(:, :, end+1)] =
msfsyn(S_ol, size(By2), [g 0 0 1], region); % For a prescribed  $H_\infty$  performance  $g > 0$ ,
compute the best  $H_2$  performance g2opt
end

```

```

Optimization of      0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9
10
11
12
13

```

```

14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
45
46
47
48
49
50
51
52
53
54      8.813650
55      8.813650
56      4.303078
57      2.487326
58      1.020421
59      1.020421
***      new lower bound:      -7.646160
60      0.911127
61      0.911127
62      0.837689
63      0.806614
***      new lower bound:      -1.542024
64      0.774720
***      new lower bound:      -0.391795
65      0.774720
***      new lower bound:      -0.153963
66      0.756838
***      new lower bound:      0.658672
67      0.752805
***      new lower bound:      0.743075
68      0.751527
***      new lower bound:      0.748385

```

Result: feasible solution of required accuracy

best objective value: 0.751527
 guaranteed absolute accuracy: 3.14e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 1.17e-01
 Guaranteed H2 performance: 8.67e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9		
10		
11		
12		
13		
14		
15		
16		
17		
18		
19		
20		
21	6.690681	
22	6.690681	
23	0.663545	
24	0.663545	
25	0.663545	
26	0.663545	
27	0.586590	
28	0.586590	
29	0.542272	
30	0.505783	
31	0.505783	
32	0.479199	
33	0.479199	
***	new lower bound:	-0.285505
34	0.474509	
***	new lower bound:	-0.168081
35	0.448355	
***	new lower bound:	0.397597
36	0.443542	
***	new lower bound:	0.429991
37	0.443542	
38	0.443542	
39	0.441352	
***	new lower bound:	0.438015

Result: feasible solution of required accuracy
 best objective value: 0.441352
 guaranteed absolute accuracy: 3.34e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 1.27e-01
Guaranteed H2 performance: 6.64e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9		
10		
11		
12		
13		
14		
15		
16		
17	21.878313	
18	4.756127	
19	4.756127	
20	1.940960	
21	0.972253	
22	0.972253	
23	0.830431	
24	0.830431	
25	0.830431	
26	0.637802	
27	0.637802	
28	0.589957	
29	0.589957	
***	new lower bound:	-0.725094
30	0.484738	
***	new lower bound:	-0.523960
31	0.484738	
***	new lower bound:	-0.463680
32	0.436130	
***	new lower bound:	-0.315052
33	0.412175	
***	new lower bound:	-0.274591
34	0.412175	
***	new lower bound:	-0.235625
35	0.412175	
***	new lower bound:	-0.075366
36	0.405913	
***	new lower bound:	0.296524
37	0.392489	
***	new lower bound:	0.366702
38	0.389130	
***	new lower bound:	0.383351
39	0.388974	
***	new lower bound:	0.387294

Result: feasible solution of required accuracy
best objective value: 0.388974
guaranteed absolute accuracy: 1.68e-03

f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 1.37e-01

Guaranteed H2 performance: 6.24e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9		
10		
11		
12		
13		
14		
15	19.992215	
16	4.355720	
17	4.355720	
18	0.743749	
19	0.743749	
20	0.743749	
21	0.622811	
22	0.622811	
23	0.513826	
24	0.513826	
25	0.436682	
26	0.436682	
27	0.406329	
***	new lower bound:	-0.314782
28	0.406329	
***	new lower bound:	-0.270665
29	0.403780	
***	new lower bound:	-0.192041
30	0.403780	
***	new lower bound:	-0.061452
31	0.377833	
***	new lower bound:	0.254016
32	0.377833	
33	0.364902	
***	new lower bound:	0.335221
34	0.364902	
35	0.361781	
***	new lower bound:	0.358344

Result: feasible solution of required accuracy

best objective value: 0.361781

guaranteed absolute accuracy: 3.44e-03

f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 1.47e-01

Guaranteed H2 performance: 6.01e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9		
10		
11		
12		
13	5.766973	
14	5.766973	
15	3.130339	
16	1.293096	
17	0.678330	
18	0.678330	
19	0.678330	
20	0.678330	
21	0.508063	
22	0.508063	
23	0.432591	
24	0.432591	
***	new lower bound:	-0.423078
25	0.371832	
***	new lower bound:	-0.276020
26	0.371832	
***	new lower bound:	-0.239429
27	0.371832	
***	new lower bound:	-0.122507
28	0.364677	
***	new lower bound:	0.224073
29	0.347980	
***	new lower bound:	0.313631
30	0.344385	
***	new lower bound:	0.335994
31	0.343661	
***	new lower bound:	0.341365

Result: feasible solution of required accuracy
best objective value: 0.343661
guaranteed absolute accuracy: 2.30e-03
f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 1.57e-01
Guaranteed H2 performance: 5.86e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1
2
3

```

4
5
6
7
8
9
10
11
12          5.746616
13          5.746616
14          3.110982
15          1.242386
16          0.625016
17          0.625016
18          0.625016
19          0.625016
20          0.473473
21          0.473473
22          0.372400
23          0.372400
***          new lower bound:   -0.353637
24          0.372400
***          new lower bound:   -0.196505
25          0.362622
***          new lower bound:   -0.148435
26          0.350889
***          new lower bound:    0.247333
27          0.333768
28          0.332637
29          0.331982
***          new lower bound:    0.288946
30          0.331546
***          new lower bound:    0.308115
31          0.331546
***          new lower bound:    0.316782
32          0.331325
***          new lower bound:    0.328341

```

```

Result: feasible solution of required accuracy
       best objective value:    0.331325
       guaranteed absolute accuracy: 2.98e-03
       f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 1.67e-01
Guaranteed H2 performance: 5.76e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

```

Solver for linear objective minimization under LMI constraints

```

```

Iterations   :   Best objective value so far

```

```

1
2
3
4
5
6
7
8
9
10
11

```

12	1.206922	
13	1.206922	
14	1.206922	
15	0.795842	
16	0.633401	
17	0.633401	
18	0.633401	
19	0.558219	
20	0.558219	
21	0.546788	
22	0.546788	
23	0.356469	
24	0.356469	
25	0.356469	
***	new lower bound:	-0.170020
26	0.356469	
***	new lower bound:	-0.164250
27	0.349891	
***	new lower bound:	-0.149199
28	0.334256	
***	new lower bound:	0.261927
29	0.334256	
30	0.323710	
***	new lower bound:	0.317256
31	0.323710	
32	0.321918	
***	new lower bound:	0.320102

Result: feasible solution of required accuracy
 best objective value: 0.321918
 guaranteed absolute accuracy: 1.82e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 1.77e-01
 Guaranteed H2 performance: 5.67e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1	
2	
3	
4	
5	
6	
7	
8	
9	
10	
11	5.488095
12	5.488095
13	0.699220
14	0.699220
15	0.699220
16	0.526167
17	0.526167
18	0.424470
19	0.424470
20	0.358418
21	0.358418

```

22          0.358418
23          0.336449
***          new lower bound:   -0.190743
24          0.336449
***          new lower bound:   -0.145551
25          0.333227
***          new lower bound:    0.208500
26          0.317583
27          0.316229
***          new lower bound:    0.294306
28          0.314977
***          new lower bound:    0.309564
29          0.314528
***          new lower bound:    0.313048

```

Result: feasible solution of required accuracy
best objective value: 0.314528
guaranteed absolute accuracy: 1.48e-03
f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 1.87e-01
Guaranteed H2 performance: 5.61e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9
10
11          5.397997
12          0.644754
13          0.644754
14          0.644754
15          0.458463
16          0.458463
17          0.378444
18          0.378444
19          0.378444
20          0.378444
21          0.378444
22          0.342511
23          0.342511
24          0.336419
***          new lower bound:    0.156162
25          0.317405
***          new lower bound:    0.266517
26          0.310262
27          0.309692
28          0.309142
***          new lower bound:    0.287444
29          0.308940
***          new lower bound:    0.292713
30          0.308940

```

```

***          new lower bound:      0.301096
31          0.308854
***          new lower bound:      0.307268

Result: feasible solution of required accuracy
best objective value:      0.308854
guaranteed absolute accuracy: 1.59e-03
f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 1.97e-01
Guaranteed H2 performance: 5.56e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9
10          5.441992
11          5.441992
12          0.741771
13          0.741771
14          0.741771
15          0.566657
16          0.566657
17          0.392630
18          0.392630
19          0.360051
20          0.360051
21          0.330460
22          0.330460
***          new lower bound:      -0.190587
23          0.330460
***          new lower bound:      -0.134391
24          0.330460
***          new lower bound:      -0.128891
25          0.329807
***          new lower bound:      -0.096852
26          0.314601
***          new lower bound:      0.246491
27          0.307011
***          new lower bound:      0.290045
28          0.307011
29          0.304880
***          new lower bound:      0.302816

```

```

Result: feasible solution of required accuracy
best objective value:      0.304880
guaranteed absolute accuracy: 2.06e-03
f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 2.07e-01
Guaranteed H2 performance: 5.52e-01

```


Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9		
10	5.357088	
11	0.571191	
12	0.571191	
13	0.571191	
14	0.571191	
15	0.379356	
16	0.379356	
17	0.379356	
18	0.376774	
19	0.376774	
20	0.321983	
***	new lower bound:	0.186733
21	0.307929	
***	new lower bound:	0.271060
22	0.307929	
23	0.301707	
***	new lower bound:	0.293048
24	0.301707	
25	0.300607	
***	new lower bound:	0.298374

Result: feasible solution of required accuracy
best objective value: 0.300607
guaranteed absolute accuracy: 2.23e-03
f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 2.17e-01

Guaranteed H2 performance: 5.48e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1	
2	
3	
4	
5	
6	
7	
8	
9	
10	5.303173
11	0.891700
12	0.891700

13	0.891700	
14	0.429517	
15	0.429517	
16	0.384593	
17	0.384593	
18	0.326217	
19	0.326217	
20	0.326217	
21	0.325782	
***	new lower bound:	0.147383
22	0.310355	
***	new lower bound:	0.242505
23	0.298353	
24	0.298353	
***	new lower bound:	0.285637
25	0.298353	
***	new lower bound:	0.295718

Result: feasible solution of required accuracy
 best objective value: 0.298353
 guaranteed absolute accuracy: 2.63e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 2.27e-01
 Guaranteed H2 performance: 5.46e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9		
10	5.277338	
11	1.117180	
12	1.117180	
13	0.616078	
14	0.616078	
15	0.515604	
16	0.515604	
17	0.354624	
18	0.354624	
19	0.327342	
20	0.327342	
***	new lower bound:	-0.235911
21	0.327342	
***	new lower bound:	-0.152722
22	0.327342	
***	new lower bound:	-0.146573
23	0.323778	
***	new lower bound:	-0.119251
24	0.323778	
***	new lower bound:	0.091726
25	0.301355	
***	new lower bound:	0.253594

```

26          0.301355
27          0.301355
28          0.297122
***          new lower bound:      0.284563
29          0.297122
30          0.297122
31          0.294835
***          new lower bound:      0.293410

Result: feasible solution of required accuracy
        best objective value:      0.294835
        guaranteed absolute accuracy: 1.42e-03
        f-radius saturation: 0.000% of R = 1.00e+10

```

Guaranteed Hinf performance: 2.37e-01
Guaranteed H2 performance: 5.43e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9
10          5.269241
11          1.278793
12          1.278793
13          0.746219
14          0.746219
15          0.516149
16          0.516149
17          0.386284
18          0.354094
19          0.354094
20          0.326314
***          new lower bound:      -0.267034
21          0.326314
***          new lower bound:      -0.233233
22          0.326314
***          new lower bound:      -0.152528
23          0.326314
***          new lower bound:      -0.141258
24          0.317470
***          new lower bound:      -0.108482
25          0.317470
***          new lower bound: 6.299568e-03
26          0.304212
***          new lower bound:      0.224269
27          0.293425
***          new lower bound:      0.275955
28          0.292145
***          new lower bound:      0.289369

```

Result: feasible solution of required accuracy
best objective value: 0.292145

guaranteed absolute accuracy: 2.78e-03
f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 2.47e-01
Guaranteed H2 performance: 5.41e-01

Optimization of 0.000 * G^2 + 1.000 * H^2 :

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9	56.439812	
10	16.765966	
11	6.732069	
12	2.172628	
13	2.172628	
14	0.933548	
15	0.933548	
16	0.933548	
17	0.768574	
18	0.768574	
19	0.415204	
20	0.415204	
***	new lower bound:	-0.605689
21	0.381927	
***	new lower bound:	-0.278211
22	0.381927	
***	new lower bound:	-0.249706
23	0.314494	
***	new lower bound:	-0.207621
24	0.314494	
***	new lower bound:	-0.181640
25	0.314494	
***	new lower bound:	0.145817
26	0.302218	
27	0.294935	
***	new lower bound:	0.255879
28	0.290617	
***	new lower bound:	0.281735
29	0.290438	
***	new lower bound:	0.287841

Result: feasible solution of required accuracy
best objective value: 0.290438
guaranteed absolute accuracy: 2.60e-03
f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 2.57e-01
Guaranteed H2 performance: 5.39e-01

Optimization of 0.000 * G^2 + 1.000 * H^2 :

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9	5.350707	
10	0.597270	
11	0.597270	
12	0.597270	
13	0.597270	
14	0.465943	
15	0.404960	
16	0.404960	
17	0.360092	
18	0.360092	
19	0.332074	
20	0.332074	
***	new lower bound:	-0.230244
21	0.310461	
***	new lower bound:	-0.146776
22	0.310461	
***	new lower bound:	-0.122614
23	0.310461	
***	new lower bound:	0.159333
24	0.303403	
***	new lower bound:	0.236337
25	0.290751	
26	0.289239	
27	0.289239	
***	new lower bound:	0.279682
28	0.288242	
***	new lower bound:	0.285504

Result: feasible solution of required accuracy
 best objective value: 0.288242
 guaranteed absolute accuracy: 2.74e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 2.67e-01
 Guaranteed H2 performance: 5.37e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1	
2	
3	
4	
5	
6	
7	
8	
9	5.325718
10	5.325718

11	1.149855	
12	1.149855	
13	0.802671	
14	0.623550	
15	0.623550	
16	0.623550	
17	0.565800	
18	0.565800	
19	0.389679	
20	0.358979	
21	0.358979	
***	new lower bound:	-0.395988
22	0.308738	
***	new lower bound:	-0.209787
23	0.308738	
***	new lower bound:	-0.185907
24	0.308738	
***	new lower bound:	-0.089342
25	0.304680	
***	new lower bound:	0.182499
26	0.290482	
***	new lower bound:	0.260396
27	0.288019	
***	new lower bound:	0.280113
28	0.288019	
29	0.286992	
***	new lower bound:	0.284962

Result: feasible solution of required accuracy
 best objective value: 0.286992
 guaranteed absolute accuracy: 2.03e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 2.77e-01
 Guaranteed H2 performance: 5.36e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1	
2	
3	
4	
5	
6	
7	
8	
9	5.309449
10	5.309449
11	1.133781
12	1.133781
13	0.788412
14	0.609408
15	0.609408
16	0.609408
17	0.552951
18	0.552951
19	0.380687
20	0.350780
21	0.350780

```

***          new lower bound:   -0.384089
22           0.325580
***          new lower bound:   -0.202953
23           0.303591
***          new lower bound:   -0.179142
24           0.303591
***          new lower bound:   -0.157075
25           0.303591
***          new lower bound:   -0.074399
26           0.300469
***          new lower bound:    0.250377
27           0.287403
28           0.286120
29           0.286120
***          new lower bound:    0.252638
30           0.286120
***          new lower bound:    0.283581

```

```

Result: feasible solution of required accuracy
best objective value:    0.286120
guaranteed absolute accuracy: 2.54e-03
f-radius saturation:  0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 2.87e-01
Guaranteed H2 performance: 5.35e-01

```

```

Optimization of  0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          5.300476
10         5.300476
11         1.090293
12         1.090293
13         0.747382
14         0.569362
15         0.569362
16         0.569362
17         0.569362
18         0.428780
19         0.428780
20         0.332256
21         0.332256
***          new lower bound:   -0.292593
22           0.311150
***          new lower bound:   -0.167451
23           0.311150
***          new lower bound:   -0.141712
24           0.306979
***          new lower bound:    0.155258
25           0.292744
26           0.288664
***          new lower bound:    0.157820

```

```

27          0.287067
***          new lower bound:    0.175981
28          0.286661
***          new lower bound:    0.227249
29          0.286661
***          new lower bound:    0.267209
30          0.284750
***          new lower bound:    0.283076

```

```

Result: feasible solution of required accuracy
        best objective value:    0.284750
        guaranteed absolute accuracy: 1.67e-03
        f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 2.97e-01
Guaranteed H2 performance: 5.34e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

```

Iterations   :   Best objective value so far

```

```

1
2
3
4
5
6
7
8
9          5.296315
10         5.296315
11         1.039457
12         1.039457
13         0.698160
14         0.698160
15         0.485699
16         0.485699
17         0.331887
18         0.331887
19         0.306158
20         0.306158
***          new lower bound:    -0.232250
21         0.306158
***          new lower bound:    -0.138127
22         0.306158
***          new lower bound:    -0.125078
23         0.306158
***          new lower bound:    -0.085354
24         0.301493
***          new lower bound:    0.181066
25         0.287063
26         0.285629
***          new lower bound:    0.262735
27         0.283661
***          new lower bound:    0.278168
28         0.283202
***          new lower bound:    0.281703

```

```

Result: feasible solution of required accuracy
        best objective value:    0.283202
        guaranteed absolute accuracy: 1.50e-03

```


f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 3.07e-01

Guaranteed H2 performance: 5.32e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9	5.293456	
10	5.293456	
11	0.998243	
12	0.998243	
13	0.657561	
14	0.657561	
15	0.457603	
16	0.457603	
17	0.315448	
18	0.315448	
19	0.315448	
20	0.315448	
21	0.315448	
22	0.308669	
23	0.294014	
***	new lower bound:	0.219890
24	0.285819	
25	0.285819	
26	0.283433	
***	new lower bound:	0.270945
27	0.282690	
***	new lower bound:	0.281079

Result: feasible solution of required accuracy

best objective value: 0.282690

guaranteed absolute accuracy: 1.61e-03

f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 3.17e-01

Guaranteed H2 performance: 5.32e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1
2
3
4
5
6

```

7
8
9          5.287749
10         5.287749
11         0.971918
12         0.971918
13         0.631805
14         0.631805
15         0.440237
16         0.440237
17         0.304838
18         0.304838
19         0.304838
20         0.304838
21         0.304838
22         0.304838
23         0.299131
***          new lower bound:      0.241526
24         0.282832
25         0.282240
26         0.281793
***          new lower bound:      0.277207
27         0.281793
28         0.281296
***          new lower bound:      0.280124

Result:  feasible solution of required accuracy
        best objective value:      0.281296
        guaranteed absolute accuracy: 1.17e-03
        f-radius saturation: 0.000% of R = 1.00e+10

```

Guaranteed Hinf performance: 3.27e-01
Guaranteed H2 performance: 5.30e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          5.275133
10         5.275133
11         0.956183
12         0.956183
13         0.618099
14         0.618099
15         0.431693
16         0.431693
17         0.299274
18         0.299274
19         0.299274
20         0.299274
21         0.299274
22         0.299274
23         0.296552

```

```

***          new lower bound:      0.247087
24          0.281780
25          0.280800
26          0.280800
***          new lower bound:      0.274572
27          0.280800
***          new lower bound:      0.278296

Result: feasible solution of required accuracy
best objective value:      0.280800
guaranteed absolute accuracy: 2.50e-03
f-radius saturation: 0.000% of R = 1.00e+10

```

Guaranteed Hinf performance: 3.37e-01
Guaranteed H2 performance: 5.30e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          5.252649
10         5.252649
11         0.942960
12         0.942960
13         0.610021
14         0.610021
15         0.427203
16         0.427203
17         0.296175
18         0.296175
19         0.296175
20         0.296175
21         0.296175
22         0.296175
23         0.295110
***          new lower bound:      0.249027
24         0.280638
25         0.280216
26         0.279811
***          new lower bound:      0.276728
27         0.279736
***          new lower bound:      0.278866

```

```

Result: feasible solution of required accuracy
best objective value:      0.279736
guaranteed absolute accuracy: 8.70e-04
f-radius saturation: 0.000% of R = 1.00e+10

```

Guaranteed Hinf performance: 3.47e-01
Guaranteed H2 performance: 5.29e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1	
2	
3	
4	
5	
6	
7	
8	
9	5.219047
10	5.219047
11	0.925107
12	0.925107
13	0.601228
14	0.601228
15	0.421917
16	0.421917
17	0.322250
18	0.322250
19	0.322250
20	0.322250
21	0.317687
***	new lower bound: 0.104542
22	0.317687
23	0.287770
***	new lower bound: 0.235486
24	0.287770
25	0.281751
***	new lower bound: 0.268314
26	0.281751
27	0.281751
28	0.279742
***	new lower bound: 0.278161

Result: feasible solution of required accuracy
 best objective value: 0.279742
 guaranteed absolute accuracy: 1.58e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 3.57e-01

Guaranteed H2 performance: 5.29e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1	
2	
3	
4	
5	
6	
7	
8	
9	5.174277
10	5.174277
11	0.899070

```

12          0.899070
13          0.899070
14          0.503422
15          0.503422
16          0.346047
17          0.346047
18          0.317744
19          0.317744
20          0.317744
21          0.317744
22          0.314525
23          0.314525
***          new lower bound:      0.068910
24          0.314525
***          new lower bound:      0.080598
25          0.283114
***          new lower bound:      0.238772
26          0.283114
27          0.279512
***          new lower bound:      0.276856

```

```

Result: feasible solution of required accuracy
        best objective value:      0.279512
        guaranteed absolute accuracy: 2.66e-03
        f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 3.67e-01
Guaranteed H2 performance: 5.29e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          5.118491
10         5.118491
11         2.987645
12         2.131077
13         1.645651
14         1.645651
15         1.645651
16         1.645651
17         1.503729
18         1.503729
19         1.503729
20         1.087154
21         1.087154
22         1.087154
23         0.988715
24         0.988715
25         0.568632
26         0.522442
***          new lower bound:      -0.741054
27         0.440505

```

```

***          new lower bound:   -0.668575
28           0.370759
***          new lower bound:   -0.612918
29           0.340453
***          new lower bound:   -0.563760
30           0.340453
***          new lower bound:   -0.518707
31           0.314828
***          new lower bound:   -0.205915
32           0.314828
***          new lower bound:   -0.184169
33           0.299496
***          new lower bound:   -0.120238
34           0.299496
***          new lower bound:   -0.095521
35           0.294546
***          new lower bound:    0.179888
36           0.281504
***          new lower bound:    0.253441
37           0.278738
***          new lower bound:    0.271841
38           0.278738
39           0.278160
***          new lower bound:    0.276295

```

Result: feasible solution of required accuracy
 best objective value: 0.278160
 guaranteed absolute accuracy: 1.87e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 3.77e-01
 Guaranteed H2 performance: 5.27e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          5.051525
10         5.051525
11         2.934265
12         2.079541
13         2.079541
14         0.864748
15         0.864748
16         0.596081
17         0.596081
18         0.366637
19         0.366637
20         0.366637
21         0.361749
***          new lower bound:   -0.231332
22         0.361749
***          new lower bound:   -0.052681

```

```

23          0.292280
***          new lower bound:    0.186009
24          0.282695
25          0.279558
26          0.279558
***          new lower bound:    0.197217
27          0.278705
***          new lower bound:    0.264727
28          0.277737
***          new lower bound:    0.275099

```

```

Result: feasible solution of required accuracy
        best objective value:    0.277737
        guaranteed absolute accuracy: 2.64e-03
        f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 3.87e-01
Guaranteed H2 performance: 5.27e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          4.972957
10         4.972957
11         2.872541
12         2.019762
13         2.019762
14         0.831971
15         0.831971
16         0.697380
17         0.697380
18         0.426441
19         0.390358
20         0.390358
***          new lower bound:    -0.562760
21         0.390358
***          new lower bound:    -0.266292
22         0.348011
***          new lower bound:    -0.233584
23         0.324324
***          new lower bound:    -0.206216
24         0.324324
***          new lower bound:    -0.181535
25         0.295031
***          new lower bound:    0.117466
26         0.289105
***          new lower bound:    0.223285
27         0.280017
28         0.278817
29         0.278056
***          new lower bound:    0.239685
30         0.277853

```

```

***          new lower bound:      0.252874
31          0.277853
***          new lower bound:      0.268838
32          0.277165
***          new lower bound:      0.275460

```

```

Result: feasible solution of required accuracy
best objective value:      0.277165
guaranteed absolute accuracy: 1.70e-03
f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 3.97e-01
Guaranteed H2 performance: 5.26e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          4.882299
10         4.882299
11         2.802051
12         1.951124
13         1.951124
14         1.164806
15         1.164806
16         0.824345
17         0.824345
18         0.552701
19         0.416372
20         0.416372
21         0.382491
***          new lower bound:      -0.344127
22         0.382491
***          new lower bound:      -0.310265
23         0.346908
***          new lower bound:      -0.223337
24         0.346908
***          new lower bound:      -0.199923
25         0.297758
***          new lower bound:      -0.181722
26         0.297758
***          new lower bound:      -0.156669
27         0.297758
***          new lower bound:      0.153035
28         0.287284
***          new lower bound:      0.227378
29         0.279061
30         0.277611
***          new lower bound:      0.266832
31         0.276594
***          new lower bound:      0.273967

```

```

Result: feasible solution of required accuracy

```


best objective value: 0.276594
 guaranteed absolute accuracy: 2.63e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 4.07e-01
 Guaranteed H2 performance: 5.26e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1		
2		
3		
4		
5		
6		
7		
8		
9	20.522156	
10	4.983219	
11	4.983219	
12	1.812666	
13	0.791272	
14	0.791272	
15	0.791272	
16	0.593552	
17	0.593552	
18	0.367185	
19	0.367185	
20	0.337018	
***	new lower bound:	-0.261360
21	0.337018	
***	new lower bound:	-0.225821
22	0.292146	
***	new lower bound:	-0.163545
23	0.292146	
***	new lower bound:	-0.139643
24	0.292146	
***	new lower bound:	0.164380
25	0.285093	
***	new lower bound:	0.234657
26	0.277287	
***	new lower bound:	0.265305
27	0.276278	
***	new lower bound:	0.273654

Result: feasible solution of required accuracy
 best objective value: 0.276278
 guaranteed absolute accuracy: 2.62e-03
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 4.17e-01
 Guaranteed H2 performance: 5.26e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9      20.365134
10     5.184928
11     5.184928
12     1.970300
13     0.933385
14     0.933385
15     0.759952
16     0.759952
17     0.514488
18     0.429346
19     0.357559
***    new lower bound:   -0.673224
20     0.357559
***    new lower bound:   -0.597414
21     0.329073
***    new lower bound:   -0.239833
22     0.329073
***    new lower bound:   -0.216490
23     0.307964
***    new lower bound:   -0.163539
24     0.307964
***    new lower bound:   -0.141342
25     0.302017
***    new lower bound:    0.126889
26     0.302017
27     0.284377
***    new lower bound:    0.237228
28     0.277663
***    new lower bound:    0.266119
29     0.277663
30     0.276374
***    new lower bound:    0.274936

```

```

Result: feasible solution of required accuracy
        best objective value:    0.276374
        guaranteed absolute accuracy: 1.44e-03
        f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 4.27e-01
Guaranteed H2 performance: 5.26e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7

```

```

8
9      20.200108
10     5.422973
11     5.422973
12     2.149450
13     1.091906
14     1.091906
15     0.708330
16     0.708330
17     0.530126
18     0.441301
19     0.364660
20     0.364660
***      new lower bound:   -0.694196
21         0.336178
***      new lower bound:   -0.283347
22         0.336178
***      new lower bound:   -0.256915
23         0.312174
***      new lower bound:   -0.176648
24         0.312174
***      new lower bound:   -0.155268
25         0.312174
***      new lower bound:   -0.136357
26         0.307588
***      new lower bound:    0.115658
27         0.307588
28         0.281454
***      new lower bound:    0.237654
29         0.281454
30         0.277338
***      new lower bound:    0.272475
31         0.277338
32         0.275652
***      new lower bound:    0.274435

```

```

Result: feasible solution of required accuracy
        best objective value:    0.275652
        guaranteed absolute accuracy: 1.22e-03
        f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 4.37e-01
Guaranteed H2 performance: 5.25e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

```

Solver for linear objective minimization under LMI constraints

```

```

Iterations   :   Best objective value so far

```

```

1
2
3
4
5
6
7
8
9      20.027703
10     5.699485
11     5.699485
12     2.351402
13     1.267784

```

14	1.267784	
15	0.837332	
16	0.837332	
17	0.620285	
18	0.515161	
19	0.381842	
20	0.381842	
***	new lower bound:	-0.824391
21	0.323714	
***	new lower bound:	-0.319381
22	0.323714	
***	new lower bound:	-0.290806
23	0.299913	
***	new lower bound:	-0.177688
24	0.299913	
***	new lower bound:	-0.157328
25	0.299913	
***	new lower bound:	-0.124416
26	0.299913	
***	new lower bound:	-0.096068
27	0.297821	
***	new lower bound:	0.145505
28	0.281986	
29	0.281010	
***	new lower bound:	0.157622
30	0.281010	
***	new lower bound:	0.234179
31	0.275955	
***	new lower bound:	0.272832
32	0.275955	
33	0.275235	
***	new lower bound:	0.274318

Result: feasible solution of required accuracy
 best objective value: 0.275235
 guaranteed absolute accuracy: 9.17e-04
 f-radius saturation: 0.000% of R = 1.00e+10

Guaranteed Hinf performance: 4.47e-01
 Guaranteed H2 performance: 5.25e-01

Optimization of $0.000 * G^2 + 1.000 * H^2$:

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

1	
2	
3	
4	
5	
6	
7	
8	
9	19.848822
10	5.578587
11	5.578587
12	2.276763
13	1.208282
14	1.208282
15	0.787871
16	0.787871

```

17          0.585324
18          0.486272
19          0.360308
20          0.360308
***      new lower bound:   -0.780525
21          0.332258
***      new lower bound:   -0.302471
22          0.332258
***      new lower bound:   -0.274914
23          0.307974
***      new lower bound:   -0.179294
24          0.307974
***      new lower bound:   -0.158390
25          0.307974
***      new lower bound:   -0.133933
26          0.306759
***      new lower bound:   -0.111289
27          0.306759
***      new lower bound: -3.684644e-04
28          0.286998
***      new lower bound:    0.209284
29          0.277244
***      new lower bound:    0.259144
30          0.275536
***      new lower bound:    0.270868
31          0.275536
32          0.274986
***      new lower bound:    0.273772

```

```

Result: feasible solution of required accuracy
best objective value:    0.274986
guaranteed absolute accuracy: 1.21e-03
f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 4.57e-01
Guaranteed H2 performance: 5.24e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

```

Iterations   :   Best objective value so far

```

```

1
2
3
4
5
6
7
8
9          19.664713
10         5.438906
11         5.438906
12         2.185182
13         1.132283
14         1.132283
15         0.925695
16         0.925695
17         0.684693
18         0.625741
19         0.470951
20         0.389203

```

```

***          new lower bound:   -0.832996
21           0.353727
***          new lower bound:   -0.130600
22           0.353727
23           0.328989
24           0.307694
25           0.307694
***          new lower bound:   -0.124115
26           0.307694
***          new lower bound:   -0.114729
27           0.298954
***          new lower bound:    0.132954
28           0.298954
29           0.280551
***          new lower bound:    0.237680
30           0.275151
***          new lower bound:    0.265591
31           0.274767
***          new lower bound:    0.272157

```

```

Result: feasible solution of required accuracy
best objective value:    0.274767
guaranteed absolute accuracy: 2.61e-03
f-radius saturation: 0.000% of R = 1.00e+10

```

```

Guaranteed Hinf performance: 4.67e-01
Guaranteed H2 performance: 5.24e-01

```

```

Optimization of 0.000 * G^2 + 1.000 * H^2 :
-----

```

Solver for linear objective minimization under LMI constraints

Iterations : Best objective value so far

```

1
2
3
4
5
6
7
8
9          19.477077
10         5.278553
11         5.278553
12         2.073804
13         2.073804
14         0.700033
15         0.700033
16         0.700033
17         0.531676
18         0.531676
19         0.359692
20         0.359692
21         0.329234
22         0.329234
***          new lower bound...

```

Closed-Loop System

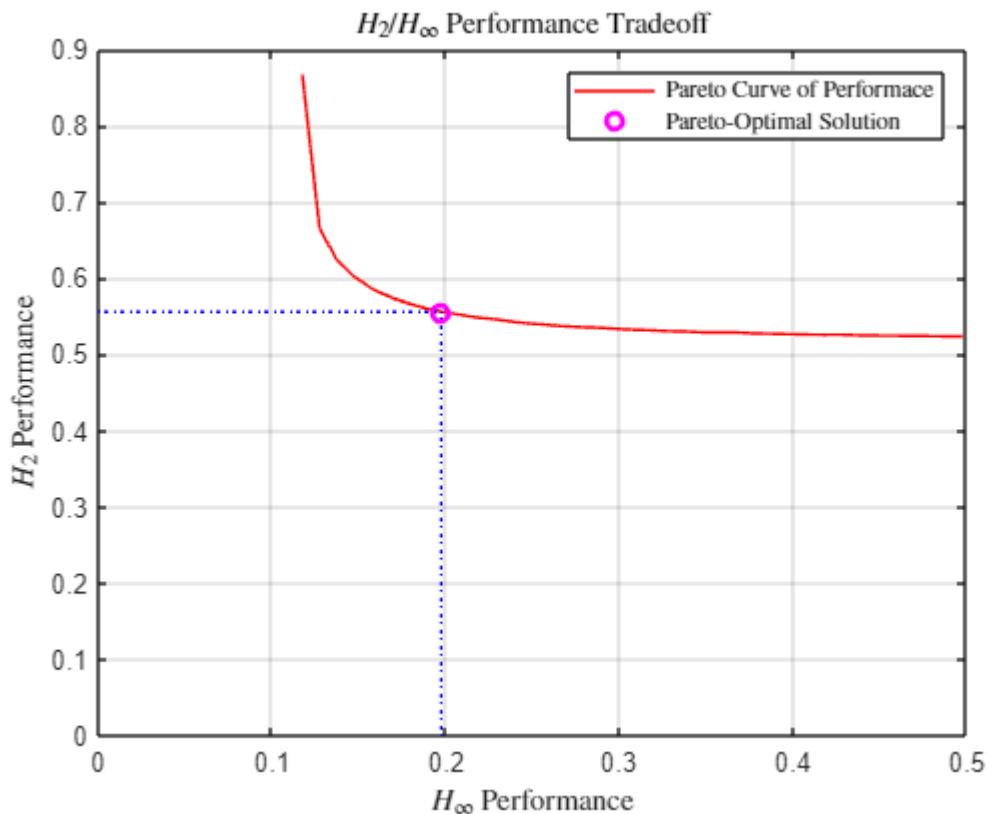
```
% Find Pareto-optimal solution
```

```

dist = sqrt(g1_opt.^2 + g2_opt.^2);
idx_popt = find(dist == min(dist));
g1_popt = g1_opt(idx_popt); %  $\gamma_1^*$  ( $\Gamma_{ee}$ )
g2_popt = g2_opt(idx_popt); %  $\gamma_2^*$  ( $\Gamma_{ep}$ )
K = K(:, :, idx_popt); % State-feedback controller gain K
S_cl = S_cl(:, :, idx_popt); % Closed-loop system representation
X = X(:, :, idx_popt); % Lyapunov matrix

% Plot
figure()
plot(g1_opt, g2_opt, 'r-')
hold on
plot(g1_popt, g2_popt, 'mo', LineWidth=2)
plot([0, g1_popt], [g2_popt, g2_popt], 'b:');
plot([g1_popt, g1_popt], [0, g2_popt], 'b:');
xlabel('$H_{\infty}$ Performance', 'Interpreter','latex')
ylabel('$H_2$ Performance', 'Interpreter','latex')
legend('Pareto Curve of Performance', 'Pareto-Optimal Solution',
'Interpreter','latex')
hold off
title('$H_2/H_{\infty}$ Performance Tradeoff', 'Interpreter','latex');
grid('on')

```



```

psinfo(S_cl); % Print closed-loop system information

```

Polytopic model with 16 vertex systems
Each system has 2 state(s), 1 input(s), and 12 output(s)

```
% Test robust stability of closed-loop system via parametric Lyapunov functions
[tmin_cl, Q0_cl, Q1_cl, Q2_cl, Q3_cl] = pdlstab(S_cl, [0 0])
```

```
Solver for LMI feasibility problems  $L(x) < R(x)$ 
This solver minimizes  $t$  subject to  $L(x) < R(x) + t*I$ 
The best value of  $t$  should be negative for feasibility
```

```
Iteration : Best value of t so far
```

1	0.225681
2	0.045101
3	0.040600
4	9.256315e-03
5	6.698021e-03
6	-8.031560e-03

```
Result: best value of t: -8.031560e-03
f-radius saturation: 0.005% of R = 1.00e+07
```

```
This system is stable in the specified parameter range
tmin_cl = -0.0080
```

```
Q0_cl = 2x2
    2.1291    0.3112
    0.3112    1.0034
Q1_cl = 2x2
    2.1291    0.3112
    0.3112    1.0034
Q2_cl = 2x2
    2.1291    0.3112
    0.3112    1.0034
Q3_cl = 2x2
    2.1291    0.3112
    0.3112    1.0034
```

```
% Simulate straight-line motion with varying friction and disturbance
```

```
% Initialization
```

```
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;
```

```
% Simulation at each millisecond
```

```
for t=1:3000
    system_cl = psinfo(S_cl, 'eval', polydec(pv, [u1_signal(t) u2_signal(t)
u3_signal(t) u4_signal(t)]));
    [A_cl, B_cl, C_cl, D_cl] = ltiss(system_cl);
```



```

X_dot(:,t) = A_cl*X(:,t) + B_cl*(w_signal(:,t).*cos(Psi(t)));
X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
Beta(t) = X(1,t);
Psi(t+1) = Psi(t) + X(2,t)*dt;
Vx(t) = v*cos(Psi(t) + Beta(t));
Vy(t) = v*sin(Psi(t) + Beta(t));
Px(t+1) = Px(t) + Vx(t)*dt;
Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA = Beta;
BETA_DOT = X_dot(1,:);
PSI = Psi;
PSI_DOT = X(2,:);
PX = Px;
VX = Vx;
PY = Py;
VY = Vy;

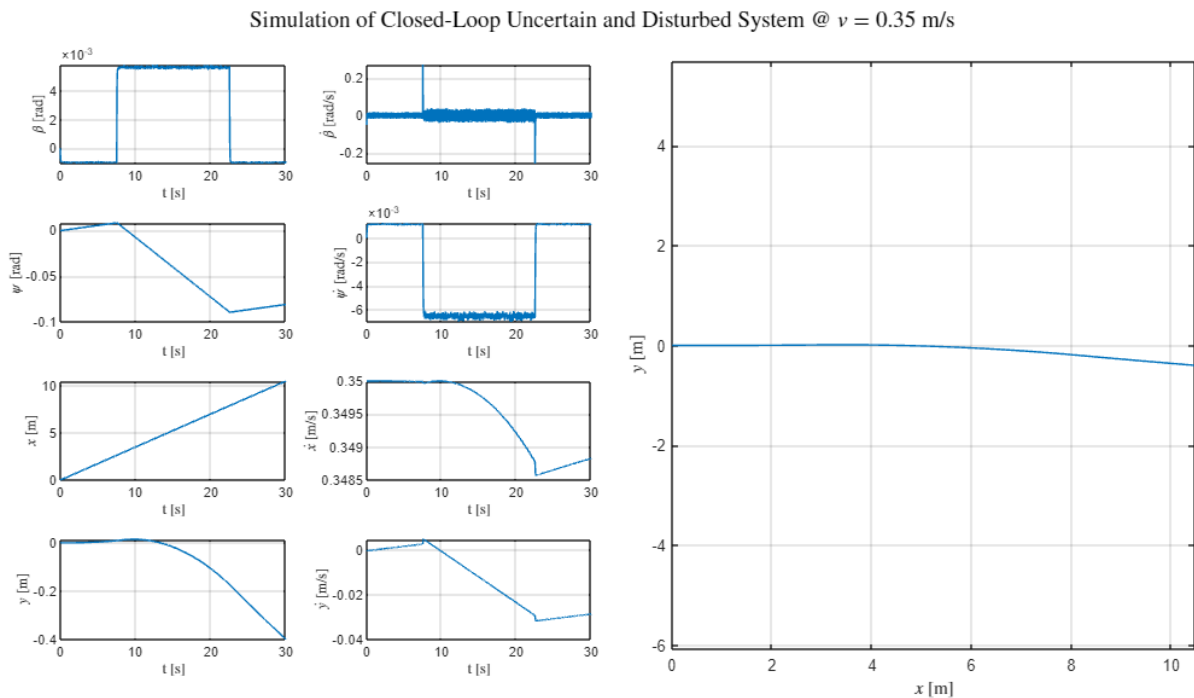
% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX(:,1:end)')

```

```

xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000, PY(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)
plot(t/1000, VY(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX(:,1:end-1), PY(:,1:end-1))
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Simulation of Closed-Loop Uncertain and Disturbed System @ $v = 0.35$ m/
s', 'Interpreter', 'latex')

```



Open-Loop vs. Closed Loop System Performance

```

% Friction variation and disturbance

rng(1, "twister"); % Fix random seed and algorithm

```

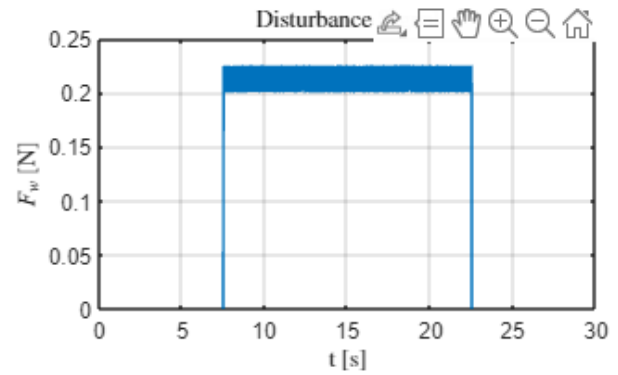
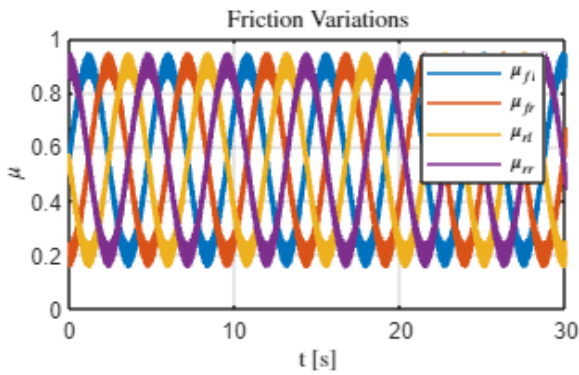
```

% Friction
theta = linspace(0, 2*pi, 30000);
frequency = 1;
u1_signal = 0.35 * sin(2*pi*frequency*theta) + 0.55;
u2_signal = 0.35 * sin(2*pi*frequency*theta - pi/2) + 0.55;
u3_signal = 0.35 * sin(2*pi*frequency*theta - pi) + 0.55;
u4_signal = 0.35 * sin(2*pi*frequency*theta - 3*pi/2) + 0.55;
u1_signal = u1_signal - 0.05 + 0.1 * rand(1, length(u1_signal));
u2_signal = u2_signal - 0.05 + 0.1 * rand(1, length(u2_signal));
u3_signal = u3_signal - 0.05 + 0.1 * rand(1, length(u3_signal));
u4_signal = u4_signal - 0.05 + 0.1 * rand(1, length(u4_signal));

% Disturbance
w_signal = [zeros(1,7500) 0.25*ones(1,15000) zeros(1,7500)];
w_signal = w_signal - 0.05 + 0.1*(max(w_signal)) * rand(1, length(w_signal));

% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [800, 200];
% Friction
subplot(1,2,1)
plot(t/1000,u1_signal)
hold on;
plot(t/1000,u2_signal)
plot(t/1000,u3_signal)
plot(t/1000,u4_signal)
legend('$\mu_{fl}$','$\mu_{fr}$','$\mu_{rl}$','$\mu_{rr}$',
'$', 'Interpreter', 'latex', 'Location', 'northeast')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\mu$', 'Interpreter', 'latex')
xlim([0,30]);
ylim([0,1]);
title('Friction Variations', 'Interpreter', 'latex')
grid on;
hold off;
% Disturbance
subplot(1,2,2)
plot(t/1000,w_signal)
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$F_w$ [N]', 'Interpreter', 'latex')
xlim([0,30]);
ylim([0,0.25]);
title('Disturbance Signal', 'Interpreter', 'latex')
grid on;

```



Stabilizing Controller

```
% Simulate open-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_ol = psinfo(S_ol, 'eval', [u1_signal(t) u2_signal(t) u3_signal(t)
u4_signal(t)]);
    [A_ol, B_ol, C_ol, D_ol] = ltiss(system_ol);
    X_dot(:,t) = A_ol*X(:,t) + B_ol*((w_signal(t).*cos(Psi(t)))); [0;0;0;0];
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_OL = Beta;
BETA_DOT_OL = X_dot(1,:);
PSI_OL = Psi;
PSI_DOT_OL = X_dot(2,:);
PX_OL = Px;
```

```

VX_OL = Vx;
PY_OL = Py;
VY_OL = Vy;

% Simulate closed-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_cl = psinfo(S_cl, 'eval', polydec(pv, [u1_signal(t) u2_signal(t)
u3_signal(t) u4_signal(t)]));
    [A_cl, B_cl, C_cl, D_cl] = ltiss(system_cl);
    X_dot(:,t) = A_cl*X(:,t) + B_cl*(w_signal(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_CL = Beta;
BETA_DOT_CL = X_dot(1,:);
PSI_CL = Psi;
PSI_DOT_CL = X(2,:);
PX_CL = Px;
VX_CL = Vx;
PY_CL = Py;
VY_CL = Vy;

% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states

```

```

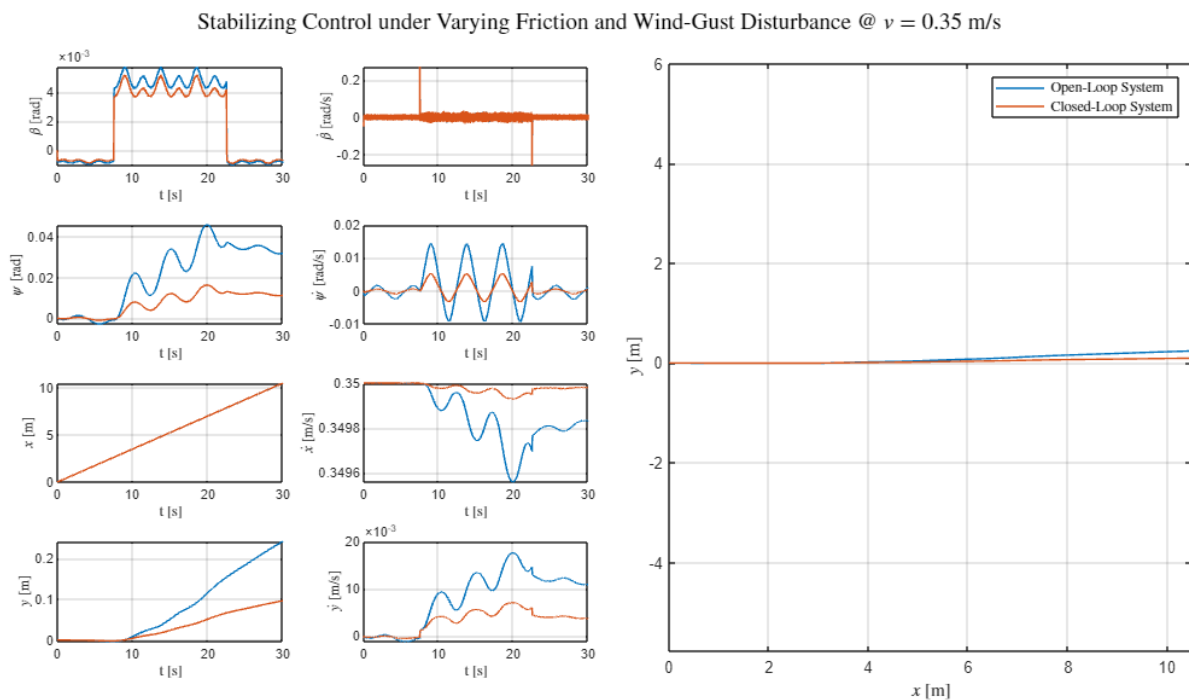
subplot(4,4,1)
plot(t/1000,BETA_OL(:,1:end))
hold on
plot(t/1000,BETA_CL(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_OL(:,1:end))
hold on
plot(t/1000,BETA_DOT_CL(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_OL(:,1:end-1))
hold on
plot(t/1000,PSI_CL(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_OL(:,1:end-1))
hold on
plot(t/1000,PSI_DOT_CL(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_OL(:,1:end-1))
hold on
plot(t/1000,PX_CL(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_OL(:,1:end))
hold on
plot(t/1000,VX_CL(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_OL(:,1:end-1))
hold on
plot(t/1000,PY_CL(:,1:end-1))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)

```

```

plot(t/1000,VY_OL(:,1:end)')
hold on
plot(t/1000,VY_CL(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_OL(:,1:end-1)',PY_OL(:,1:end-1)')
hold on
plot(PX_CL(:,1:end-1)',PY_CL(:,1:end-1)')
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Stabilizing Control under Varying Friction and Wind-Gust Disturbance @ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('Open-Loop System', 'Closed-Loop System', ...
      'Interpreter', 'latex', 'Location', 'northeast')

```



Tracking Controller

```

% Reference states (LC = lane-change, SP = skidpad, FH = fishhook, SL = slalom)
BETA_REF = BETA_LC(5,:);
BETA_DOT_REF = BETA_DOT_LC(5,:);
PSI_REF = PSI_LC(5,:);
PSI_DOT_REF = PSI_DOT_LC(5,:);
PX_REF = PX_LC(5,:);

```

```

VX_REF = VX_LC(5,:);
PY_REF = PY_LC(5,:);
VY_REF = VY_LC(5,:);
DELTA_REF = u_lane_change;

% Simulate open-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_ol = psinfo(S_ol, 'eval', [u1_signal(t) u2_signal(t) u3_signal(t)
u4_signal(t)]);
    [A_ol, B_ol, C_ol, D_ol] = ltiss(system_ol);
    X_dot(:,t) = A_ol*X(:,t) + B_ol*[(w_signal(:,t)).*cos(Psi(t))]; DELTA_REF(:,t)];
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_OL = Beta;
BETA_DOT_OL = X_dot(1,:);
PSI_OL = Psi;
PSI_DOT_OL = X_dot(2,:);
PX_OL = Px;
VX_OL = Vx;
PY_OL = Py;
VY_OL = Vy;

% Simulate closed-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]

```



```

v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_cl = psinfo(S_cl, 'eval', polydec(pv, [u1_signal(t) u2_signal(t)
u3_signal(t) u4_signal(t)]));
    [A_cl, B_cl, C_cl, D_cl] = ltiss(system_cl);
    X_dot(:,t) = A_cl*(X(:,t) - [BETA_REF(t);PSI_DOT_REF(t)]) +
B_cl*(w_signal(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_CL = Beta;
BETA_DOT_CL = X_dot(1,:);
PSI_CL = Psi;
PSI_DOT_CL = X(2,:);
PX_CL = Px;
VX_CL = Vx;
PY_CL = Py;
VY_CL = Vy;

% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_OL(:,1:end))
hold on
plot(t/1000,BETA_CL(:,1:end))
plot(t/1000,BETA_REF(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')

```

```

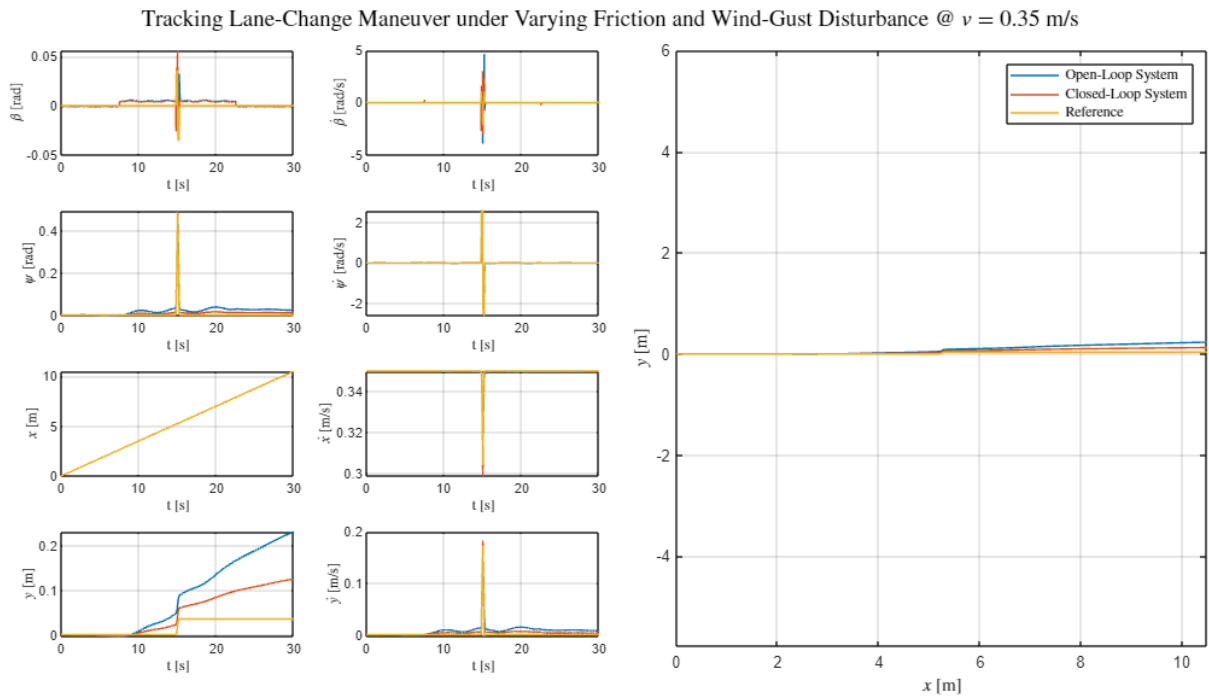
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_OL(:,1:end)')
hold on
plot(t/1000,BETA_DOT_CL(:,1:end)')
plot(t/1000,BETA_DOT_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_CL(:,1:end-1)')
plot(t/1000,PSI_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_DOT_CL(:,1:end-1)')
plot(t/1000,PSI_DOT_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_OL(:,1:end-1)')
hold on
plot(t/1000,PX_CL(:,1:end-1)')
plot(t/1000,PX_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_OL(:,1:end)')
hold on
plot(t/1000,VX_CL(:,1:end)')
plot(t/1000,VX_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_OL(:,1:end-1)')
hold on
plot(t/1000,PY_CL(:,1:end-1)')
plot(t/1000,PY_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)

```

```

plot(t/1000,VY_OL(:,1:end)')
hold on
plot(t/1000,VY_CL(:,1:end)')
plot(t/1000,VY_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_OL(:,1:end-1)',PY_OL(:,1:end-1)')
hold on
plot(PX_CL(:,1:end-1)',PY_CL(:,1:end-1)')
plot(PX_REF(:,1:end-1)',PY_REF(:,1:end-1)')
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Tracking Lane-Change Maneuver under Varying Friction and Wind-Gust
Disturbance @ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('Open-Loop System', 'Closed-Loop System', 'Reference', ...
'Interpreter', 'latex', 'Location', 'northeast')

```



```

% Reference states (LC = lane-change, SP = skidpad, FH = fishhook, SL = slalom)
BETA_REF = BETA_SP(5,:);
BETA_DOT_REF = BETA_DOT_SP(5,:);
PSI_REF = PSI_SP(5,:);
PSI_DOT_REF = PSI_DOT_SP(5,:);
PX_REF = PX_SP(5,:);

```

```

VX_REF = VX_SP(5,:);
PY_REF = PY_SP(5,:);
VY_REF = VY_SP(5,:);
DELTA_REF = u_skidpad;

% Simulate open-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_ol = psinfo(S_ol, 'eval', [u1_signal(t) u2_signal(t) u3_signal(t)
u4_signal(t)]);
    [A_ol, B_ol, C_ol, D_ol] = ltiss(system_ol);
    X_dot(:,t) = A_ol*X(:,t) + B_ol*[(w_signal(:,t)).*cos(Psi(t))]; DELTA_REF(:,t)];
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_OL = Beta;
BETA_DOT_OL = X_dot(1,:);
PSI_OL = Psi;
PSI_DOT_OL = X_dot(2,:);
PX_OL = Px;
VX_OL = Vx;
PY_OL = Py;
VY_OL = Vy;

% Simulate closed-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]

```

```

v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_cl = psinfo(S_cl, 'eval', polydec(pv, [u1_signal(t) u2_signal(t)
u3_signal(t) u4_signal(t)]));
    [A_cl, B_cl, C_cl, D_cl] = ltiss(system_cl);
    X_dot(:,t) = A_cl*(X(:,t) - [BETA_REF(t);PSI_DOT_REF(t)]) +
B_cl*(w_signal(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_CL = Beta;
BETA_DOT_CL = X_dot(1,:);
PSI_CL = Psi;
PSI_DOT_CL = X(2,:);
PX_CL = Px;
VX_CL = Vx;
PY_CL = Py;
VY_CL = Vy;

% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_OL(:,1:end))
hold on
plot(t/1000,BETA_CL(:,1:end))
plot(t/1000,BETA_REF(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')

```

```

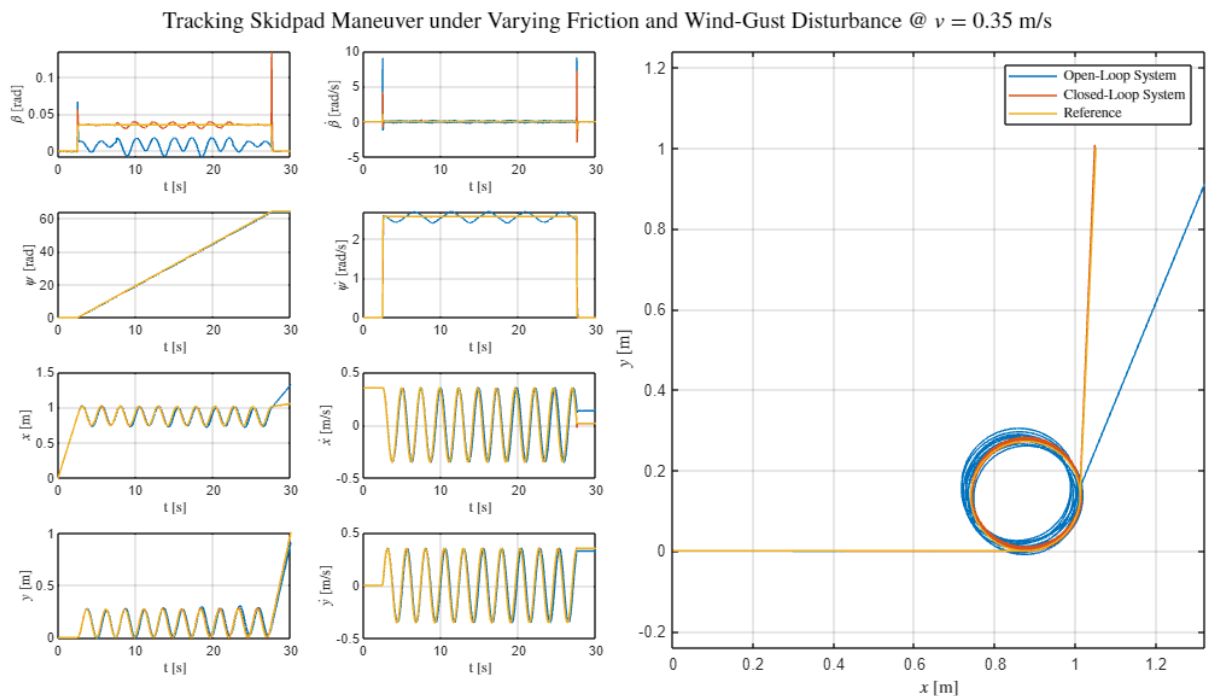
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_OL(:,1:end)')
hold on
plot(t/1000,BETA_DOT_CL(:,1:end)')
plot(t/1000,BETA_DOT_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_CL(:,1:end-1)')
plot(t/1000,PSI_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_DOT_CL(:,1:end-1)')
plot(t/1000,PSI_DOT_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_OL(:,1:end-1)')
hold on
plot(t/1000,PX_CL(:,1:end-1)')
plot(t/1000,PX_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_OL(:,1:end)')
hold on
plot(t/1000,VX_CL(:,1:end)')
plot(t/1000,VX_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_OL(:,1:end-1)')
hold on
plot(t/1000,PY_CL(:,1:end-1)')
plot(t/1000,PY_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)

```

```

plot(t/1000,VY_OL(:,1:end)')
hold on
plot(t/1000,VY_CL(:,1:end)')
plot(t/1000,VY_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_OL(:,1:end-1)',PY_OL(:,1:end-1)')
hold on
plot(PX_CL(:,1:end-1)',PY_CL(:,1:end-1)')
plot(PX_REF(:,1:end-1)',PY_REF(:,1:end-1)')
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Tracking Skidpad Maneuver under Varying Friction and Wind-Gust Disturbance
@ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('Open-Loop System', 'Closed-Loop System', 'Reference', ...
'Interpreter', 'latex', 'Location', 'northeast')

```



```

% Reference states (LC = lane-change, SP = skidpad, FH = fishhook, SL = slalom)
BETA_REF = BETA_FH(5,:);
BETA_DOT_REF = BETA_DOT_FH(5,:);
PSI_REF = PSI_FH(5,:);
PSI_DOT_REF = PSI_DOT_FH(5,:);
PX_REF = PX_FH(5,:);

```

```

VX_REF = VX_FH(5,:);
PY_REF = PY_FH(5,:);
VY_REF = VY_FH(5,:);
DELTA_REF = u_fishhook;

% Simulate open-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_ol = psinfo(S_ol, 'eval', [u1_signal(t) u2_signal(t) u3_signal(t)
u4_signal(t)]);
    [A_ol, B_ol, C_ol, D_ol] = ltiss(system_ol);
    X_dot(:,t) = A_ol*X(:,t) + B_ol*[(w_signal(:,t)).*cos(Psi(t))]; DELTA_REF(:,t)];
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_OL = Beta;
BETA_DOT_OL = X_dot(1,:);
PSI_OL = Psi;
PSI_DOT_OL = X(2,:);
PX_OL = Px;
VX_OL = Vx;
PY_OL = Py;
VY_OL = Vy;

% Simulate closed-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]

```



```

v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_cl = psinfo(S_cl, 'eval', polydec(pv, [u1_signal(t) u2_signal(t)
u3_signal(t) u4_signal(t)]));
    [A_cl, B_cl, C_cl, D_cl] = ltiss(system_cl);
    X_dot(:,t) = A_cl*(X(:,t) - [BETA_REF(t);PSI_DOT_REF(t)]) +
B_cl*(w_signal(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_CL = Beta;
BETA_DOT_CL = X_dot(1,:);
PSI_CL = Psi;
PSI_DOT_CL = X(2,:);
PX_CL = Px;
VX_CL = Vx;
PY_CL = Py;
VY_CL = Vy;

% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_OL(:,1:end))
hold on
plot(t/1000,BETA_CL(:,1:end))
plot(t/1000,BETA_REF(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')

```

```

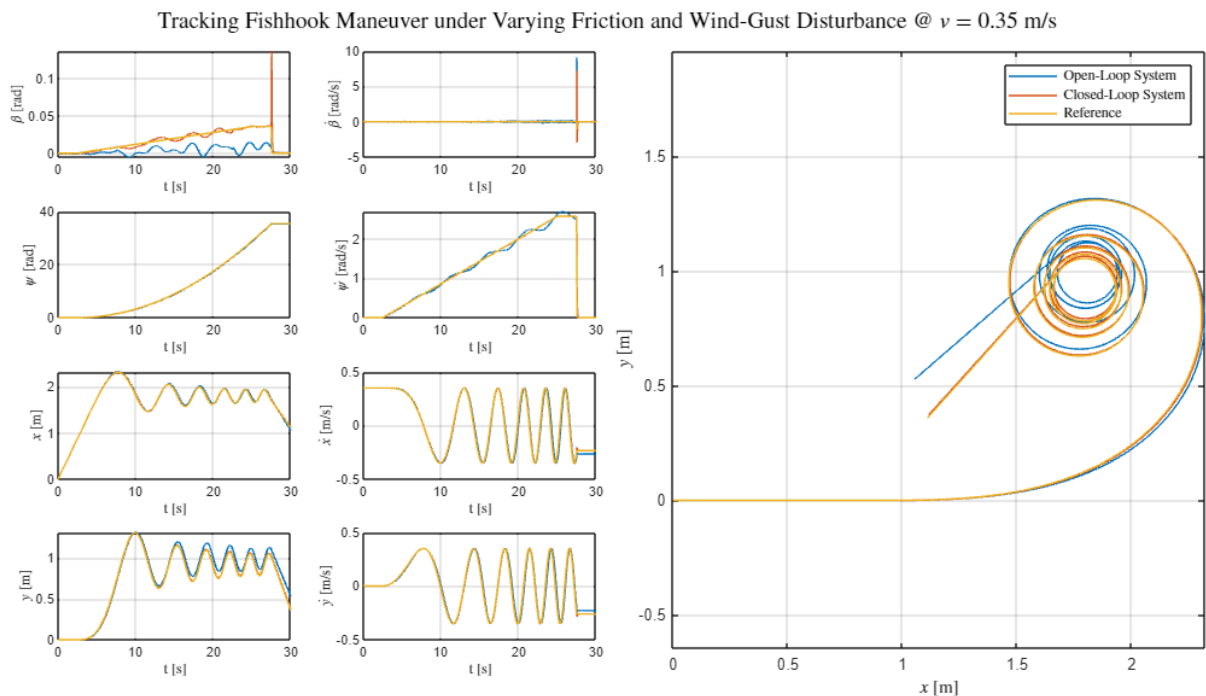
grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_OL(:,1:end)')
hold on
plot(t/1000,BETA_DOT_CL(:,1:end)')
plot(t/1000,BETA_DOT_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_CL(:,1:end-1)')
plot(t/1000,PSI_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_DOT_CL(:,1:end-1)')
plot(t/1000,PSI_DOT_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_OL(:,1:end-1)')
hold on
plot(t/1000,PX_CL(:,1:end-1)')
plot(t/1000,PX_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_OL(:,1:end)')
hold on
plot(t/1000,VX_CL(:,1:end)')
plot(t/1000,VX_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_OL(:,1:end-1)')
hold on
plot(t/1000,PY_CL(:,1:end-1)')
plot(t/1000,PY_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)

```

```

plot(t/1000,VY_OL(:,1:end)')
hold on
plot(t/1000,VY_CL(:,1:end)')
plot(t/1000,VY_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_OL(:,1:end-1)',PY_OL(:,1:end-1)')
hold on
plot(PX_CL(:,1:end-1)',PY_CL(:,1:end-1)')
plot(PX_REF(:,1:end-1)',PY_REF(:,1:end-1)')
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Tracking Fishhook Maneuver under Varying Friction and Wind-Gust
Disturbance @ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('Open-Loop System', 'Closed-Loop System', 'Reference', ...
'Interpreter', 'latex', 'Location', 'northeast')

```



```

% Reference states (LC = lane-change, SP = skidpad, FH = fishhook, SL = slalom)
BETA_REF = BETA_SL(5,:);
BETA_DOT_REF = BETA_DOT_SL(5,:);
PSI_REF = PSI_SL(5,:);
PSI_DOT_REF = PSI_DOT_SL(5,:);
PX_REF = PX_SL(5,:);

```

```

VX_REF = VX_SL(5,:);
PY_REF = PY_SL(5,:);
VY_REF = VY_SL(5,:);
DELTA_REF = u_slalom;

% Simulate open-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]
v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_ol = psinfo(S_ol, 'eval', [u1_signal(t) u2_signal(t) u3_signal(t)
u4_signal(t)]);
    [A_ol, B_ol, C_ol, D_ol] = ltiss(system_ol);
    X_dot(:,t) = A_ol*X(:,t) + B_ol*[(w_signal(:,t)).*cos(Psi(t))]; DELTA_REF(:,t)];
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_OL = Beta;
BETA_DOT_OL = X_dot(1,:);
PSI_OL = Psi;
PSI_DOT_OL = X_dot(2,:);
PX_OL = Px;
VX_OL = Vx;
PY_OL = Py;
VY_OL = Vy;

% Simulate closed-loop system with varying friction and disturbance

% Initialization
dt = 0.001; % Simulation timestep [s]

```

```

v = 0.35; % Vehicle velocity [m/s]
X = [];
X(:,1) = [0;0];
X_dot = [];
Psi = [];
Psi(1) = 0;
Beta = [];
Vx = [];
Vy = [];
Px = [];
Px(1) = 0;
Py = [];
Py(1) = 0;

% Simulation at each millisecond
for t=1:30000
    system_cl = psinfo(S_cl, 'eval', polydec(pv, [u1_signal(t) u2_signal(t)
u3_signal(t) u4_signal(t)]));
    [A_cl, B_cl, C_cl, D_cl] = ltiss(system_cl);
    X_dot(:,t) = A_cl*(X(:,t) - [BETA_REF(t);PSI_DOT_REF(t)]) +
B_cl*(w_signal(:,t).*cos(Psi(t)));
    X(:,t+1) = X(:,t) + X_dot(:,t)*dt;
    Beta(t) = X(1,t);
    Psi(t+1) = Psi(t) + X(2,t)*dt;
    Vx(t) = v*cos(Psi(t) + Beta(t));
    Vy(t) = v*sin(Psi(t) + Beta(t));
    Px(t+1) = Px(t) + Vx(t)*dt;
    Py(t+1) = Py(t) + Vy(t)*dt;
end
BETA_CL = Beta;
BETA_DOT_CL = X_dot(1,:);
PSI_CL = Psi;
PSI_DOT_CL = X(2,:);
PX_CL = Px;
VX_CL = Vx;
PY_CL = Py;
VY_CL = Vy;

% Plot
t=1:30000;
fig = figure();
fig.Position(3:4) = [1500, 750];
% Plot states
subplot(4,4,1)
plot(t/1000,BETA_OL(:,1:end))
hold on
plot(t/1000,BETA_CL(:,1:end))
plot(t/1000,BETA_REF(:,1:end))
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\beta$ [rad]', 'Interpreter', 'latex')

```

```

grid('on')
subplot(4,4,2)
plot(t/1000,BETA_DOT_OL(:,1:end)')
hold on
plot(t/1000,BETA_DOT_CL(:,1:end)')
plot(t/1000,BETA_DOT_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\beta}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,5)
plot(t/1000,PSI_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_CL(:,1:end-1)')
plot(t/1000,PSI_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\psi$ [rad]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,6)
plot(t/1000,PSI_DOT_OL(:,1:end-1)')
hold on
plot(t/1000,PSI_DOT_CL(:,1:end-1)')
plot(t/1000,PSI_DOT_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{\psi}$ [rad/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,9)
plot(t/1000,PX_OL(:,1:end-1)')
hold on
plot(t/1000,PX_CL(:,1:end-1)')
plot(t/1000,PX_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$x$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,10)
plot(t/1000,VX_OL(:,1:end)')
hold on
plot(t/1000,VX_CL(:,1:end)')
plot(t/1000,VX_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{x}$ [m/s]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,13)
plot(t/1000,PY_OL(:,1:end-1)')
hold on
plot(t/1000,PY_CL(:,1:end-1)')
plot(t/1000,PY_REF(:,1:end-1)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
grid('on')
subplot(4,4,14)

```

```

plot(t/1000,VY_OL(:,1:end)')
hold on
plot(t/1000,VY_CL(:,1:end)')
plot(t/1000,VY_REF(:,1:end)')
xlabel('t [s]', 'Interpreter', 'latex')
ylabel('$\dot{y}$ [m/s]', 'Interpreter', 'latex')
grid('on')
% Plot trajectory
subplot(4,4,[3,4,7,8,11,12,15,16])
plot(PX_OL(:,1:end-1)',PY_OL(:,1:end-1)')
hold on
plot(PX_CL(:,1:end-1)',PY_CL(:,1:end-1)')
plot(PX_REF(:,1:end-1)',PY_REF(:,1:end-1)')
xlabel('$x$ [m]', 'Interpreter', 'latex')
ylabel('$y$ [m]', 'Interpreter', 'latex')
axis equal
grid('on')
% Common title and legend
sgtitle('Tracking Slalom Maneuver under Varying Friction and Wind-Gust Disturbance
@ $v = 0.35$ m/s', 'Interpreter', 'latex')
legend('Open-Loop System', 'Closed-Loop System', 'Reference', ...
'Interpreter', 'latex', 'Location', 'northeast')

```

