# MECH 6323 - Homework 6

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clear
close all

### **Problem 1**

#### Part a

```
(a) \mu_{\Delta}(M) = 0 \Rightarrow M = 0. M_{\Delta}(M) := \left[ \min \overline{\partial}(\Delta) : |I - M\Delta| = 0 \right]^{-1}

\Rightarrow \Delta \Rightarrow |I - M\Delta| = 0
However doen't say anything about M
\therefore False
```

#### Part b

(b) 
$$\mu_{\Delta}(M_{1} + M_{2}) \leq \mu_{\Delta}(M_{1}) + \mu_{\Delta}(M_{2})$$
 $M_{\Delta}(M_{1} + M_{2}) \leq m \text{ in } || D(M_{1} + M_{2}) \overline{D}^{1}||$ 
 $= || D M_{1} D^{1} + D P_{2} D^{1} ||$ 
 $M_{\Delta}(M_{1} + M_{2}) \leq M_{\Delta}(M_{1}) + M_{\Delta}(M_{2})$ 
 $= || D M_{1} D^{1} + D P_{2} D^{1} ||$ 
 $M_{\Delta}(M_{1} + M_{2}) \leq M_{\Delta}(M_{1}) + M_{\Delta}(M_{2})$ 
 $M_{\Delta}(M_{1}) = M_{\Delta}(M_{2})$ 

Part c

(c) 
$$\mu_{\Delta}(\alpha M) = |\alpha| \mu_{\Delta}(M)$$
.

 $M_{A}(M) := \begin{bmatrix} \min \overline{\sigma}(\Delta) : |I - \alpha \Lambda \Delta| = 0 \end{bmatrix}^{-1}$ 
 $M_{A}(\alpha M) := \begin{bmatrix} \min \overline{\sigma}(\Delta \Delta) : |I - M(\Delta \Delta)| = 0 \end{bmatrix}^{-1}$ 
 $\overline{\sigma}(\Delta \Delta) = \alpha \overline{\sigma}(\Delta) = |\alpha| \overline{\sigma}(\Delta)$ 
 $M_{A}(\alpha M) = |\alpha| M_{A}(M)$ 

#### Part d

(d) 
$$\mu_{\Delta}(I) = 1$$
.

 $M_{\Delta}(I) := \begin{bmatrix} \min \overline{\sigma}(\Delta) : |I - I\Delta| = 0 \end{bmatrix}^{-1}$ 
 $\overline{\sigma}(I) = I$ 
 $M_{\Delta}(I) = I$ 

### Part e

```
N = sym('n',2,'real');
M = sym('m',2,'real');
MN = M*N;
sigma_delta_1 = max(svd(M));
mu_delta_2 = 1/max(svd(N\eye(2)));
mu_delta = 1/max(svd(MN\eye(2)));
solve(mu_delta > sigma_delta_1 + mu_delta_2, {M,N})
```

#### Part f

```
N = sym('n',2,'real');
M = sym('m',2,'real');
MN = M*N;
sigma_delta_1 = max(svd(N));
mu_delta_2 = 1/max(svd(M\eye(2)));
mu_delta = 1/max(svd(MN\eye(2)));
solve(mu_delta <= sigma_delta_1 + mu_delta_2,{M,N})</pre>
```

# **Problem 2**

2. Let 
$$\Delta = \begin{bmatrix} \Delta_1 & 0 \\ 0 & \Delta_2 \end{bmatrix}$$
, where  $\Delta_i$  are structured uncertainties. Show that 
$$\mu_{\Delta} \left( \begin{bmatrix} M_{11} & M_{12} \\ 0 & M_{22} \end{bmatrix} \right) = \max \{ \mu_{\Delta_1}(M_{11}), \mu_{\Delta_2}(M_{22}) \}.$$

$$\Delta = \begin{bmatrix} \Delta_1 & 0 \\ 0 & \Delta_2 \end{bmatrix} \qquad \mathcal{M} \begin{bmatrix} M_{11} & M_{12} \\ O & M_{23} \end{bmatrix}$$

$$\mathcal{M}_{\Delta}(\mathcal{M}) := \begin{bmatrix} \omega_i n & \overline{\mathcal{S}}(\Delta) : | \mathbf{I} - \mathcal{M} \Delta | = 0 \end{bmatrix}^{-1}$$

$$I - M \Delta = \begin{bmatrix} I & O \\ D & I \end{bmatrix} - \begin{bmatrix} M_{11} & M_{12} \\ O & M_{22} \end{bmatrix} \Delta_1 & O \\ O & M_{22} \end{bmatrix}$$

$$= \begin{bmatrix} I & O \\ O & I \end{bmatrix} - \begin{bmatrix} M_{11} \Delta_1 & M_{12} \Delta_2 \\ O & M_{22} \Delta_2 \end{bmatrix}$$

$$= \begin{bmatrix} I - M_{11} \Delta_1 & -M_{22} \Delta_2 \\ O & I - M_{22} \Delta_2 \end{bmatrix}$$

$$= \begin{bmatrix} I - M_{11} \Delta_1 & I - M_{22} \Delta_2 \\ O & I - M_{22} \Delta_2 \end{bmatrix} - \begin{bmatrix} M_{12} \Delta_2 & O \\ O & I - M_{22} \Delta_2 \end{bmatrix}$$

$$|I - M \Delta | = 0 \iff M_{11} \Delta_{1} = I \land M_{22} \Delta_{2} = I$$

$$|J - M \Delta | = Max \{ \overline{O}(A_{1}), \overline{O}(A_{2}) \}$$

$$|M = M | Max \{ \overline{O}(A_{1}) : I - M_{11} \Delta_{1} = 0 \}$$

$$|J - M = M | Max \{ M_{2}(M_{11}), M_{2}(M_{22}) \}$$

$$|J - M = M | Max \{ M_{2}(M_{11}), M_{2}(M_{22}) \}$$

### **Problem 3**

```
tol_0 = 0.001;
%from simplemu.m
% Example data
i=sqrt(-1);
mat=[0.10+0.07*i -0.1538+0.1615*i 0-0.56*i 0+42.0*i 4.-1.4*i
 0-0.2730*i -0.30-0.28*i 2.86+0.546*i -26.0+72.8*i 13.+5.4600*i
 0.10+0.1750*i 0.0769-0.1077*i -0.40+0.2100*i 5.0+3.5000*i 4.5-0.70*i
  0+0.0021*i -0.0038-0.0021*i 0.0060+0.0112*i 0.2000+0.4200*i 0.0600+0.0140*i
 0.0240+0.0280*i 0+0.0269*i -0.0660+0.0420*i 0+0.70*i -0.4210+0.49*i
1;
% Sample uncertainty structures
blka = [5 0];
b1kb = [3 0;2 0];
blkc = [1 1;1 1;1 1;2 0];
blkd = [1 1;1 1;1 1;1 1;1 1];
blke = [2 2;2 2;1 1];
blkf = [2 2;3 3];
blkg = [5 5];
b1kh = [2 3;3 2];
blki = [1 4;4 1];
% Run easymu
[upp,low,pert,dleft,dright] = easymu(mat,blka);
[low,upp];
```

#### blke

```
[upp,low,pert,dleft,dright] = easymu(mat,blke);
pert
```

```
pert = 5×5 complex
    -0.0164 - 0.0447i -0.0552 - 0.0645i
                                          0.0000 + 0.0000i
                                                             0.0000 + 0.0000i · · ·
    0.0275 - 0.0026i
                       0.0444 - 0.0215i
                                          0.0000 + 0.0000i
                                                             0.0000 + 0.0000i
    0.0000 + 0.0000i
                       0.0000 + 0.0000i
                                         -0.0048 - 0.0024i -0.0003 + 0.0000i
    0.0000 + 0.0000i
                       0.0000 + 0.0000i
                                          0.0915 - 0.0650i
                                                             0.0028 - 0.0053i
    0.0000 + 0.0000i
                       0.0000 + 0.0000i
                                          0.0000 + 0.0000i
                                                             0.0000 + 0.0000i
  low
  low = 8.8836
  norm(pert)
  ans = 0.1126
  1/norm(pert)
  ans = 8.8836
  low - 1/norm(pert) < tol_0</pre>
  ans = logical
    1
they be inversly related
  det(eye(size(mat)) - mat*pert)
  ans = 3.7583e-16 - 6.3936e-18i
  det(eye(size(mat)) - mat*pert) < tol_0</pre>
  ans = logical
    1
Which is essentially 0
  upp
 upp = 8.9017
  norm(dleft*mat/dright)
  ans = 8.9016
  upp - norm(dleft*mat/dright) < tol_0</pre>
  ans = logical
    1
blkf
  [upp,low,pert,dleft,dright] = easymu(mat,blkg);
  pert
  pert = 5 \times 5 complex
   -0.0000 + 0.0000i
                       0.0000 + 0.0000i -0.0000 + 0.0000i -0.0000 + 0.0000i · · ·
    -0.0000 + 0.0000i -0.0000 + 0.0000i -0.0000 - 0.0000i -0.0000 - 0.0000i
```

```
0.0001 + 0.0000i 0.0003 - 0.0000i
                                         0.0000 + 0.0000i
                                                           0.0000 + 0.0000i
                                         0.0006 - 0.0005i
    0.0000 - 0.0052i -0.0033 - 0.0091i
                                                           0.0000 - 0.0001i
    0.0009 - 0.0000i 0.0015 - 0.0006i
                                         0.0001 + 0.0001i
                                                           0.0000 + 0.0000i
  low
  low = 89.4127
  norm(pert)
  ans = 0.0112
  1/norm(pert)
 ans = 89.4127
  low - 1/norm(pert) < tol_0</pre>
  ans = logical
    1
they be inversly related
  det(eye(size(mat)) - mat*pert)
  ans = -4.0660e-16 - 7.9568e-18i
  det(eye(size(mat)) - mat*pert) < tol_0</pre>
  ans = logical
Which is essentially 0
  upp
 upp = 89.4127
  norm(dleft*mat/dright)
  ans = 89.4127
  upp - norm(dleft*mat/dright) < tol_0</pre>
  ans = logical
    1
Problem 4
```

Naive methods (for speed of writing)

```
% Parameters
m = 2150; %kg
b = 20; %N s/m
c = 150; %N/deg
tau = 0.1; %s
% System Definition
```

```
U_{cmd} = tf([1],[tau 1])
```

```
U_cmd =
      1
  0.1 s + 1
```

Continuous-time transfer function.

Continuous-time transfer function.

```
P = U_cmd * tf([c], [m b])
P =
         150
 215 s^2 + 2152 s + 20
```

### Part a

```
B = zpk(tf([1 0.02], [0.5 1]))
B =
 2 (s+0.02)
   (s+2)
Continuous-time zero/pole/gain model.
```

The magnitude of the frequency response to input must not excede a frequency dependent response with a maximum of 2 but only for frequencies above around 2 rad/s.

### Part b

```
% arbritarily select large k_p
k_p = 50;
% controller: u = k_p * (r - y)
C = k_p;
% L(s) = C(s) * P(s)
L = zpk(P*C)
L =
       34.884
  (s+10) (s+0.009302)
Continuous-time zero/pole/gain model.
```

```
% T(s) = Y(s) / R(s) = L(s) / (1 + L(s))
T = zpk((P*C)/(1 + P*C))
```

```
T =
          34.884 (s+10) (s+0.009302)
```

```
(s+10) (s+0.009302) (s^2 + 10.01s + 34.98)
```

Continuous-time zero/pole/gain model.

```
% or w/ minreal = feedback
T = feedback(L,1)
```

T =

```
34.884
-----
(s^2 + 10.01s + 34.98)
```

Continuous-time zero/pole/gain model.

The time constant  $\tau_{cls} = \frac{1}{\omega_0}$  which for  $T(s) = \frac{1}{s^2 + 2 * \omega_0 s + \omega_0^2}$  is simple to calculate... but also just using this:

### damp(T)

Pole	Damping	Frequency (rad/seconds)	Time Constant (seconds)
-5.00e+00 + 3.15e+00i	8.46e-01	5.91e+00	2.00e-01
-5.00e+00 - 3.15e+00i	8.46e-01	5.91e+00	2.00e-01

```
[w, ~, ~] = damp(T);
tau_cls = 1./w
```

```
tau_cls = 2×1
0.1691
0.1691
```

```
% S(s) = E(s) / R(s) = 1/(1+L(s))

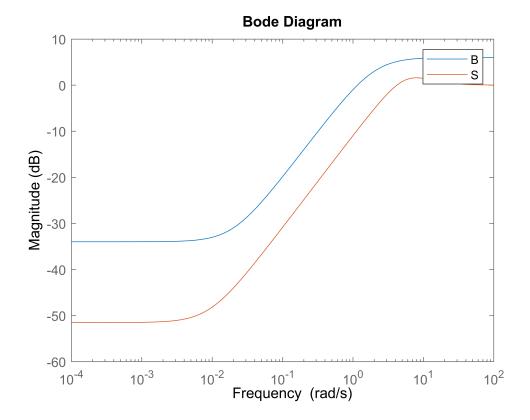
S = zpk((1)/(1 + P*C))
```

S =

```
(s+10) (s+0.009302)
------
(s^2 + 10.01s + 34.98)
```

Continuous-time zero/pole/gain model.

```
figure()
bodemag(B)
hold on
bodemag(S)
legend()
```



### Part c

```
% m_u = 2150 + 150 * ureal('delta_m',0);
% b_u = b * (1 + 0.2*ureal('delta_b',0));
% c_u = c * (1 + 0.1*ureal('delta_c',0));
% tau_u = ureal('tau_u',0.1, 'Range', [0.05, 0.2]);
% P_u = tf([1], [tau_u 1]) * tf([c_u], [m_u b_u]);
% Ploting Relative Uncertainties
w = logspace(-2, 4, 500);
P_mag = bode(P,w);
figure
hold on
for i = 1:20
    % using braket indexing bc otherwise I get cell errors... cells are
    % wonderful
%
      P hat{i} = usample(P u);
    % Just realized we were not supposed to use ureal... so:
    m_u\{i\} = 2150 + 150*(2*rand()-1);
    b_u\{i\} = b * (1 + 0.2*(2*rand()-1));
    c_u\{i\} = c * (1 + 0.1*(2*rand()-1));
    tau_u{i} = 0.05 + 0.15*rand();
    P_hat{i} = tf([1], [tau_u{i} 1]) * tf([c_u{i}], [m_u{i} b_u{i}]);
```

```
[P_hat_mag{i}, ~, ~] = bode(P_hat{i},w);
R{i} = reshape(abs(P_hat_mag{i} - P_mag)./ abs(P_mag),[],1);
plot(w,R{i})
end
% set(gca, 'YScale', 'log')
set(gca, 'XScale', 'log')
```

Arbritrarily selected values to make it fit... W(s) could probbably be derived as well based on the extremes of the uncertainty (by that I mean it is a simple caluculation with only 5 variables with defined ranges)

```
a1 = 1.25;

a2 = 0.5;

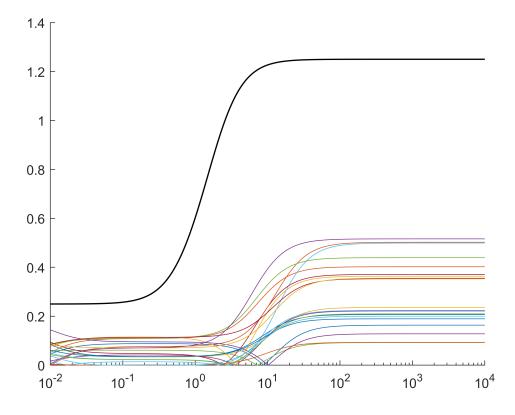
a3 = 2;

W_u = tf([a1 a2], [1 a3])
```

1.25 s + 0.5 -----s + 2

Continuous-time transfer function.

```
[W_u_mag, ~, ~] = bode(W_u,w);
plot(w,reshape(W_u_mag,[],1),'k',LineWidth=1)
```



# Part d

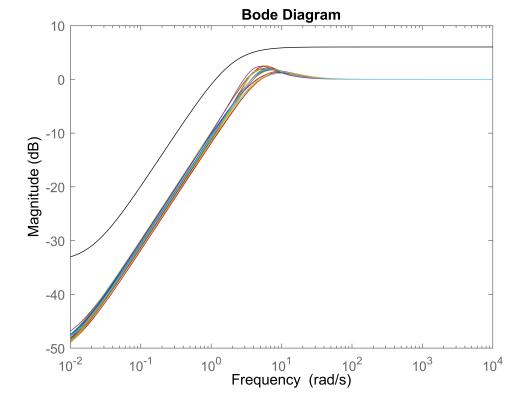
```
for i = 1:20
    P_cls_stable(i) = isstable(1 + C * P_hat{i});
end
P_cls_stable
```

According to the simplistic test of stability for each of the sampled systems, the answer is clearly yes.

A better method of confirming this would be to actully prove it for the entire region, which I may or may not make time for...

### Part e

```
figure
hold on
for i = 1:20
    S_hat{i} = (1) / (1 + C*P_hat{i});
    bodemag(S_hat{i},w)
end
bodemag(B,w, 'k')
```



Acording to bode diagram, the controller does meet performance specifications.