MECH 6323 - HW 3

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1 Problem 1

Preliminaries:

Definition 1. *Matrix Basics:* For $A \in \mathbb{C}^{n \times m}$ and $x \in \mathbb{C}^m$,

1. The Eigenvalues (λ_i) and Eigenvectors (x_i) of A are defined as the solutions to

$$\lambda_i A = \lambda_i x_i$$

2. The Spectral Radius of A is defined as

$$\rho(A) := \max_{i} |\lambda_i(A)|$$

Definition 2. Vector Norms: For $x \in \mathbb{C}^n$,

1. The 2-norm, or Euclidean norm, is defined as

$$\left\|x\right\|_2 := \sqrt{\sum_{i=1}^n x_i^2}$$

2. The 1-norm is defined as

$$||x||_1 := \sum_{i=1}^n |x_i|$$

3. The ∞ -norm is defined as

$$||x||_{\infty} := \max i = 1, \dots, n|x_i|$$

4. The p-norm is defined as

$$||x||_p := \left[\sum_{i=1}^n |x_i|^p\right]^{\frac{1}{p}}$$

Definition 3. *Matrix Norms:* For $A \in \mathbb{C}^{n \times m}$ and $x \in \mathbb{C}^m$,

1. The Induced 2-norm is defined as

$$||A||_{2\to 2} := \sup_{x \neq 0} \frac{||Ax||_2}{||x||_2}$$

and is also known as the spectral norm and has the additional properties of

(a)
$$||A||_{2\to 2} = \sqrt{\lambda_{max}(A^*A)} = \overline{\sigma}(A)$$

(b)
$$||A^*A||_{2\to 2} = ||AA^*||_{2\to 2} = ||A||_2^2$$

1.1

Problem:

For $M \in \mathbb{C}^{n \times m}$, show that for all $x \in \mathbb{C}^m$

$$\|Mx\|_2 \le \|M\|_{2\to 2} \|x\|_2$$

Solution:

Theorem 1. For $M \in \mathbb{C}^{n \times m}$, show that for all $x \in \mathbb{C}^m$

$$\|Mx\|_2 \leq \|M\|_{2\to 2} \|x\|_2$$

Proof. From the definition of the 2-norm, we have

$$||M||_{2 \to 2} := \sup_{x \neq 0} \frac{||Ax||_2}{||x||_2}$$

$$\begin{split} \|M\|_{2\to 2} &= \sup_{x\neq 0} \frac{\|Ax\|_2}{\|x\|_2} \\ \|M\|_{2\to 2} \|x\|_2 &= \sup_{x\neq 0} \frac{\|Mx\|_2}{\|x\|_2} \|x\|_2 \\ &= \sup_{x\neq 0} \|Mx\|_2 \\ \hline \|M\|_{2\to 2} \|x\|_2 \geq \|Mx\|_2 \; \forall_{x\in C^m} \end{split}$$

1.2

Problem:

Let $\{\lambda_i\}_{i=1}^n$ denote the eigenvalues of $A \in \mathbb{C}^{n \times n}$. Show that $\rho(A) \leq \|A\|_{2 \to 2}$, where $\rho(A)$ is the spectral radius of matrix A. i.e. $\rho(A) := \max_i |\lambda_i(A)|$.

Solution:

Theorem 2. Let $A \in \mathbb{C}^{n \times n}$. The spectral radius $\rho(A)$ will always be smaller then the induced 2-norm. i.e.

$$\rho(A) \leq ||A||_{2 \to 2}$$

Proof. From the definition of the induced 2-norm, we have

$$||M||_{2\to 2} := \sup_{x\neq 0} \frac{||Ax||_2}{||x||_2}$$

Additionally, from the definition of the vector 2-norm

$$\|x\|_2^2 = x^T x$$

$$||M||_{2\to 2} = \sup_{x\neq 0} \frac{||Ax||_2}{||x||_2}$$
$$(||M||_{2\to 2})^2 = \sup_{x\neq 0} \left(\frac{||Ax||_2}{||x||_2}\right)^2$$
$$= \sup_{x\neq 0} \frac{(Mx)^T Mx}{x^T x}$$

$$= \sup_{x \neq 0} \frac{x^T M^T M x}{x^T x}$$

Since $\forall_x Mx \leq Mx_{max}$, where $x_{max} = \|x\|_2 v_{max}$ and v_{max} is the eigenvector associated with $\lambda_{max} = \rho(A)$

$$\leq \sup_{x \neq 0} \frac{x_{max}^T M^T M x_{max}}{x^T x}$$

$$= \sup_{x \neq 0} \frac{\|x\|_2 v_{max}^T \lambda_{max} \lambda_{max} \|x\|_2 v_{max}}{x^T x}$$

$$= \sup_{x \neq 0} \frac{v_{max}^T \|x\|_2^2 \lambda_{max}^2 v_{max}}{x^T x}$$

$$= \sup_{x \neq 0} \frac{\lambda_{max}^2 \|x\|_2^2 \|v_{max}\|_2^2}{\|x\|_2^2}$$

$$= \sup_{x \neq 0} \frac{\lambda_{max}^2 \|x\|_2^2 (1)^2}{\|x\|_2^2}$$

$$= \sup_{x \neq 0} \lambda_{max}^2$$

$$= \lambda_{max}^2 = \rho(A)^2$$

$$(\|M\|_{2 \to 2})^2 \leq (\rho(A))^2$$

$$\|M\|_{2 \to 2} \leq \rho(A)$$

1.3

Problem:

Let $A \in \mathbb{C}^{n \times m}$ and $B \in \mathbb{C}^{m \times k}$. Prove the multiplicative property of the induced 2-norm.

$$||AB||_{2\to 2} \le ||A||_{2\to 2} ||B||_{2\to 2}$$

.

Solution:

Theorem 3. Let $A \in \mathbb{C}^{n \times m}$ and $B \in \mathbb{C}^{m \times k}$. Just like in all norms by definition,

$$||AB||_{2\to 2} \le ||A||_{2\to 2} ||B||_{2\to 2}$$

Proof. From norm definitions,

$$||Ax||_2 \le ||A||_{2\to 2} ||x||_2$$

and therefore $\forall_{x \in \mathbb{C}^k}$,

$$\begin{split} \|ABx\|_2 &\leq \|A\|_{2\to 2} \|Bx\|_2 \\ &\leq \|A\|_{2\to 2} \|B\|_{2\to 2} \|x\|_2 \end{split}$$

Taking $x = \sigma \hat{x}$ where magnitude $\sigma = ||x||_2$ and \hat{x} is the associated unit vector for x. Also, $||\sigma \hat{x}|| = \sigma ||\hat{x}||_2 = ||x||_2$

$$\begin{split} \|AB\|x\|_2 \hat{x}\|_2 &\leq \|A\|_{2 \to 2} \|B\|_{2 \to 2} \|x\|_2 \|\hat{x}\|_2 \\ \|x\|_2 \|AB\hat{x}\|_2 &\leq \|A\|_{2 \to 2} \|B\|_{2 \to 2} \|x\|_2 (1) \end{split}$$

Noting that $\|AB\hat{x}\|_2 \leq \|AB\|_{2\to 2} \|\hat{x}\|_2 = \|AB\|_{2\to 2} \|\hat{x}\|_2$. Thus, $\|AB\hat{x}\|_2 = \|AB\|_{2\to 2}$

$$\begin{aligned} \|x\|_2 \|AB\|_{\|2\to2\|} &\leq \|A\|_{2\to2} \|B\|_{2\to2} \|x\|_2 \\ & \left[\|AB\|_{\|2\to2\|} \leq \|A\|_{2\to2} \|B\|_{2\to2} \right] \end{aligned}$$

1.4

Problem:

Let $x \in \mathbb{C}^m$ and $y \in \mathbb{C}^n$. Show that if $\|y\|_2 \leq \|x\|_2$, then there exists a $\Delta \in \mathbb{C}^{n \times m}$ such that $y = \Delta x$ and $\overline{\sigma(\Delta)} \leq 1$. The choice of Δ should only be expressed in terms of x, y, and their norms. Conversely, show that if $\|y\|_2 > \|x\|_2$, then there is no $\Delta \in \mathbb{C}^{n \times m}$ such that $y = \Delta x$ and $\overline{\sigma(\Delta)} \leq 1$.

Solution:

Theorem 4. Let $x \in \mathbb{C}^m$ and $y \in \mathbb{C}^n$. There exists a $\Delta \in \mathbb{C}^{n \times m}$ such that $y = \Delta x$ and $\overline{\sigma(\Delta)} \leq 1$ if and only if $\|y\|_2 \leq \|x\|$. i.e.

$$\exists_{\Delta \in \mathbb{C}^{n \times m}} : y = \Delta x \wedge \overline{\sigma(\Delta)} \le 1 \iff ||y||_2 \le ||x||_2$$

Proof. From $y = \Delta x$ we have

$$y = \Delta x$$

$$yx^* = \Delta xx^*$$

$$\Delta = \frac{yx^*}{xx^*}$$

$$= \frac{yx^*}{\|x\|_2^2}$$

From the definition of the induced 2-norm,

$$\overline{\sigma}(\Delta) = \left\|\Delta\right\|_{2 \to 2} = \sup_{x \neq 0} \frac{\left\|\Delta x\right\|_2}{\left\|x\right\|_2}$$

Then in order for $\overline{\sigma(\Delta)} \leq 1$,

$$\begin{split} 1 \geq \overline{\sigma}(\Delta) &= \|\Delta\|_{2 \to 2} \\ &= \sup_{x \neq 0} \frac{\|\Delta x\|_2}{\|x\|_2} \\ &= \sup_{x \neq 0} \frac{\left\|\frac{yx^*}{\|x\|_2^2}x\right\|_2}{\|x\|_2} \end{split}$$

$$(1)^{2} \ge \left(\sup_{x \ne 0} \frac{\left\|\frac{yx^{*}}{\|x\|_{2}^{2}x}\right\|_{2}}{\|x\|_{2}}\right)^{2}$$

$$1 \ge \sup_{x \ne 0} \frac{\left\|\frac{yx^{*}}{\|x\|_{2}^{2}x}\right\|_{2}}{\|x\|_{2}^{2}}$$

$$= \sup_{x \ne 0} \frac{\left(\frac{yx^{*}}{\|x\|_{2}^{2}x}\right)^{*} \left(\frac{yx^{*}}{\|x\|_{2}^{2}x}\right)}{\|x\|_{2}^{2}}$$

$$= \sup_{x \ne 0} \frac{x^{*}\frac{xy^{*}yx^{*}}{\|x\|_{2}^{4}}x}{\|x\|_{2}^{2}}$$

$$= \sup_{x \ne 0} \frac{x^{*}xy^{*}yx^{*}x}{\|x\|_{2}^{6}}$$

$$= \sup_{x \ne 0} \frac{x^{*}xy^{*}yx^{*}x}{\|x\|_{2}^{6}}$$

$$= \sup_{x \ne 0} \frac{\|x\|_{2}^{2}\|y\|_{2}^{2}\|x\|_{2}^{2}}{\|x\|_{2}^{2}}$$

$$= \sup_{x \ne 0} \frac{\|y\|_{2}}{\|x\|_{2}^{2}}$$

$$1 \ge \sup_{x \ne 0} \frac{\|y\|_{2}}{\|x\|_{2}}$$

$$1 \ge \sup_{x \ne 0} \frac{\|y\|_{2}}{\|x\|_{2}}$$

Clearly this is true if and only if $\|y\|_2 \leq \|x\|_2.$

2 Problem 2

See attached MATLAB .mlx script.

MECH 6323 - Homework 3

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Date: 2022-03-01

```
clear
close all
```

Problem 2

System Defidxnidxtidxon

```
sys =
 A =
    x1 x2
  x1 -2 5
x2 -5 -3
 B =
   u1 u2
  x1 -2 4
  x2 -2 -2
 C =
    x1 x2
  y1 1 2
  y2 -4 3
 D =
    u1 u2
  y1 0 0
  y2 0 0
```

Continuous-time state-space model.

```
% Transfer Functidxon
sys_tf = tf(sys)
```

Continuous-time transfer function.

```
% ZPK Model
sys_zpk = zpk(sys)
```

Continuous-time zero/pole/gain model.

Part a - Stabidxlidxty

The stabidxlidxty of the idxnternal system P (unforced) can be determidaned based on the eidagen values of A and ensuridang $\mathbb{R}e(\lambda_i) < 0 \ \forall_{i=1,\dots,n}$.

```
P_poles = eig(A)
```

```
P_poles = 2×1 complex
-2.5000 + 4.9749i
-2.5000 - 4.9749i
```

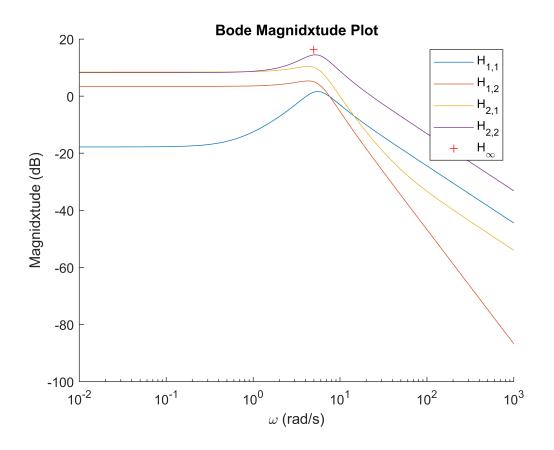
Thidxs result demonstrates that the system idxs idxndead stable widxth underdamped poles at $\lambda_{1,2} = -2.50 \pm j4.98$.

Thidxs gaurentees asymptotidxc stabidxlidxty of the system as well as BidxBO stabidxlidxty.

Part b - H_{∞} -norm

```
% H_\idxnfty calc
```

```
[P_inf_norm, omega_p] = hinfnorm(sys)
P inf norm = 6.5024
omega_p = 4.9703
% Bode Data
[mag, phase, wout] = bode(sys);
% Plot Bode mag on one plot
figure
hold on
for idx_1 = 1:2
    for idx_2 = 1:2
        plot(wout,reshape(mag2db(mag(idx_1,idx_2,:)),1,[]), 'DisplayName',['H_{',num2str(idx_1)}')
    end
end
plot(omega_p, mag2db(P_inf_norm),'+r', 'DisplayName', 'H_{\infty}')
set(gca, 'XScale', 'log')
title('Bode Magnidxtude Plot')
xlabel('\omega (rad/s)')
ylabel('Magnidxtude (dB)')
legend()
```



Part c - SVD

-1.1659 - 0.1345i -0.4407 + 1.7394i

```
% H_peak calc
H_peak = evalfr(sys,1i*omega_p)
H_peak = 2×2 complex
```

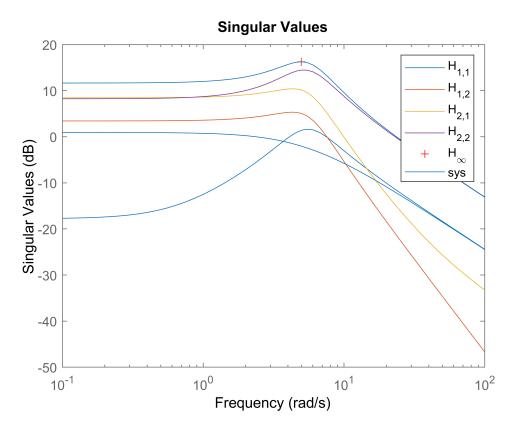
```
H_peak_svd = svd(H_peak)
```

```
H_peak_svd = 2×1
6.5024
0.7918
```

Clearly they match.

Part d - Sigma plot

```
% Plot Sigma
sigma(sys)
```



Clearly they match.

 $S = 2 \times 2$

6.5024

Part e - SVD I/O Calc

```
% SVD Calc
[U,S,V] = svd(H_peak)

U = 2×2 complex
  -0.2373 - 0.1974i    0.5432 - 0.7808i
   0.2847 - 0.9075i   -0.3014 - 0.0667i
```

0 0.7918 V = 2×2 complex 0.5170 + 0.0000i -0.8560 + 0.0000i

```
% Maximum vectors
U_{max} = U(:,1);
V_{max} = V(:,1);
% I/O Coeficient Vectors
a = abs(V_max)
a = 2 \times 1
   0.5170
   0.8560
phi = angle(V_max) - pi/2
phi = 2 \times 1
  -1.5708
   0.6637
b = S(1,1) * abs(U_max)
b = 2 \times 1
   2.0074
   6.1848
psi = angle(U_max) - pi/2
psi = 2 \times 1
   -4.0185
   -2.8376
% norm gain check
IO_norm_gain = norm(b)/norm(a)
IO_norm_gain = 6.5024
```

Part f - Simulate System

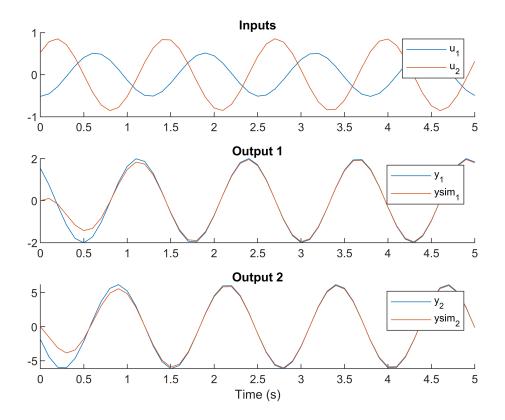
```
%% Sim Setup
dt = 0.1;
tf = 5;
t = 0:dt:tf;
%% Input/Output Production
u = a .* sin(omega_p * t + phi)
u = 2 \times 51
   -0.5170
           -0.4544
                     -0.2819
                               -0.0412
                                         0.2095
                                                   0.4095
                                                            0.5104
                                                                     0.4878 ...
   0.5273
           0.7850
                     0.8528
                                0.7142
                                         0.4027
                                                  -0.0062
                                                           -0.4136
                                                                     -0.7209
y = b \cdot * sin(omega_p * t + psi)
y = 2 \times 51
             0.7443 -0.2347
                                       -1.7992
                                                 -2.0060
                                                           -1.7274 -1.0308 * * * *
   1.5432
                             -1.1569
   -1.8515
            -4.4412 -5.9562 -6.0298
                                       -4.6443 -2.1349
                                                            0.8912
                                                                     3.7015
```

Simulat System

```
ysim = lsim(sys,u,t)';
```

Plot Sim

```
figure
% Plot Inputs
subplot(3,1,1)
hold on
plot(t,u(1,:),'DisplayName','u_1')
plot(t,u(2,:),'DisplayName','u_2')
legend()
title('Inputs')
% Plot Outputs
for idx = 1:2
     subplot(3,1,idx+1)
     hold on
    plot(t,y(idx,:),'DisplayName',['y_',num2str(idx)])
plot(t,ysim(idx,:),'DisplayName',['ysim_',num2str(idx)])
     legend()
     title(['Output ',num2str(idx)])
end
xlabel('Time (s)')
```



It is evident that works.

3 Problem 3

Problem:

Let S and T denote the sensitivity and complementary sensitivity closed-loop transfer functions. Prove that

$$||S||_{\infty} \geq ||T||_{\infty} - 1$$

Preliminaries

Definition 4. Let P and C represent the plant and controller transfer functions respectively. Within a standard unity feedback system,

1. the sensitivity closed-loop transfer function is defined as:

$$S = \frac{1}{1 + PC}$$

2. the complementary sensitivity closed-loop transfer function is defined as:

$$T = \frac{PC}{1 + PC}$$

Solution:

Theorem 5. Let P and C represent the plant and controller transfer functions respectively. Within a standard unity feedback system,

$$||S||_{\infty} \ge ||T||_{\infty} - 1$$

Proof. From the definitions,

$$T - S = \frac{PC}{1 + PC} - \frac{1}{1 + PC} = \frac{-1 + PC}{1 + PC}$$

Applying the ∞ -norm,

$$||T - s||_{\infty} = \left\| \frac{-1 + PC}{1 + PC} \right\|_{i} nfty$$

Since -1 + PC < 1 + PC,

$$||T - S||_{\infty} \leq 1$$

From the triangular inequality we have

$$||T - S||_{\infty} \ge ||T||_{\infty} - ||S||_{\infty}$$

And thus,

$$\boxed{1 \geq \|T - S\|_{\infty} \geq \|T\|_{\infty} - \|S\|_{\infty} \implies \|S\|_{\infty} \geq \|T\|_{\infty} - 1}$$

A MATLAB Code:

See attached. Additionally, all the code I write in this course can be found on my GitHub repository: https://github.com/jonaswagner2826/MECH6323