

# MECH 6323 - HW 5

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## Preliminaries

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

1. The Eigenvalues ( $\lambda_i$ ) 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

$$\rho(A) := \max_i |\lambda_i(A)|$$

3. The Complex Conjugate Transpose of  $A$ , denoted as  $A^*$ , is defined so that

$$\Re(A^*) = \Re(A^T)$$

and

$$\Im(A^*) = -\Im(A^T)$$

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

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

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

2. The 1-norm is defined as

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

3. The  $\infty$ -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

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

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

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

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

**Definition 4. Closed-loop Transfer Functions:** 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:

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

**Theorem 1. Singular Value Decomposition:** Any matrix  $M \in \mathbb{C}^{m \times n}$  can be decomposed as

$$M = U \Sigma V^*$$

with  $U \in \mathbb{C}^{n \times n}$  and  $V \in \mathbb{C}^{m \times m}$  are unitary. Additionally,  $\Sigma \in \mathbb{R}^{n \times m}$  was rectangular diagonal with

$$\Sigma = \begin{bmatrix} \hat{\Sigma} & 0_{n \times (m-n)} \end{bmatrix}, \quad m \geq n \qquad \Sigma = \begin{bmatrix} \hat{\Sigma} \\ 0_{n \times (m-n)} \end{bmatrix}, \quad n > m$$

and

$$\hat{\Sigma} = \begin{bmatrix} \sigma_1 & & & \\ & \sigma_2 & & \\ & & \ddots & \\ & & & \sigma_k \end{bmatrix} \in \mathbb{R}^{k \times k}, \quad k = \min(n, m), \quad (\sigma_1 \geq \sigma_2 \geq \dots \geq 0)$$

Additionally,  $U$  and  $V$  are orthonormal,  $UU^* = I = VV^*$ .

1. We have

$$Mv_i = \sigma_i u_i$$

and

$$M = \sum_{i=1}^k \sigma_i u_i v_i^*$$

2. SVD is related to the Eigenvalues and Eigenvectors.

$$\lambda_i(M^*M) = \sigma_i^2(M)$$

with corresponding eigenvectors  $v_i$  and

$$\lambda_i(MM^*) = \sigma_i^2(M)$$

with corresponding eigenvectors  $u_i$ .

**Theorem 2. Small-gain Theorem:** Consider an  $n_y \times n_u$  LTI system  $M$  and  $\Delta \in \mathbb{C}^{n_u \times n_y}$ .

Assuming  $M$  is stable, then the feedback system in Figure 1 is stable for all  $\bar{\sigma}(\Delta) < m$  if and only if

$$\|M\|_\infty \leq \frac{1}{m}$$

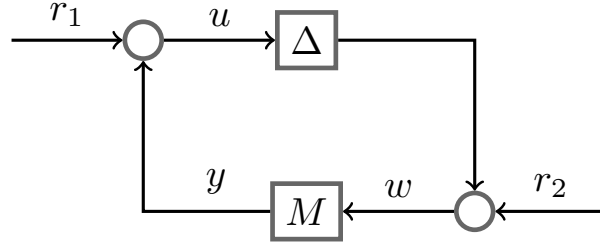


Figure 1: Small-gain theorem feedback system

## 1 Problem 1

Consider the feedback configuration in the figure below.

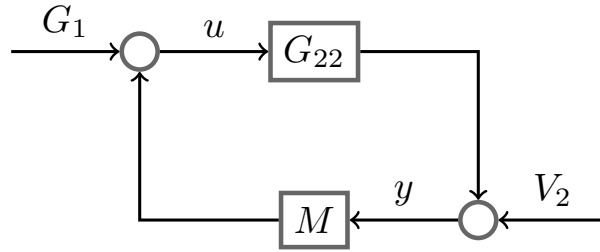


Figure 2: Problem 1 Feedback Configuration

Prove that

$$\begin{bmatrix} I & -K \\ -G_{22} & I \end{bmatrix}^{-1} = \begin{bmatrix} (I - KG_{22})^{-1} & (I - KG_{22})^{-1}K \\ (I - G_{22}K)^{-1}G_{22} & (I - G_{22}K)^{-1} \end{bmatrix} = H(G_{22}, K)$$

## 2 Problem 2

Consider the block diagram in the previous exercise. Suppose  $G_{22}$  and  $K$  have minimal state space realizations  $\begin{bmatrix} A & B_2 \\ C_2 & D_{22} \end{bmatrix}$  and  $\begin{bmatrix} A_K & B_K \\ C_K & D_K \end{bmatrix}$ . Let

$$T = \begin{bmatrix} A_1 = \begin{bmatrix} A & \\ & A_K \end{bmatrix} & B = \begin{bmatrix} B_2 & \\ & B_K \end{bmatrix} \\ -C = \begin{bmatrix} 0 & -C_K \\ -C_2 & 0 \end{bmatrix} & D = \begin{bmatrix} I & -D_K \\ -D_{22} & I \end{bmatrix} \end{bmatrix}$$

Thus,

$$T^{-1} = \begin{bmatrix} \bar{A} & \bar{B} \\ \bar{C} & \bar{D} \end{bmatrix} := \begin{bmatrix} A_1 + BD^{-1}C & BD^{-1} \\ D^{-1}C & D^{-1} \end{bmatrix}$$

where

$$\begin{aligned} \bar{D} = D^{-1} &= \begin{bmatrix} I & -D_K \\ -D_{22} & I \end{bmatrix}^{-1} \\ &= \begin{bmatrix} I + (I - D_{22}D_K)^{-1}D_{22} & D_K(I - D_{22}D_K)^{-1} \\ (I - D_{22}D_K)^{-1}D_{22} & (I - D_{22}D_K)^{-1} \end{bmatrix} \\ &= \begin{bmatrix} I & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} (I - D_{22}D_K)^{-1}D_{22} & D_K(I - D_{22}D_K)^{-1} \\ (I - D_{22}D_K)^{-1}D_{22} & (I - D_{22}D_K)^{-1} \end{bmatrix} \\ &= \begin{bmatrix} I & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} D_K \\ I \end{bmatrix} (I - D_{22}D_K)^{-1} \begin{bmatrix} D_{22} & I \end{bmatrix} \end{aligned}$$

Thus,

$$\bar{A} = A_1 + BD^{-1}C = \begin{bmatrix} I & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} B_2D_K \\ B_K \end{bmatrix} (I - D_{22}D_K)^{-1} \begin{bmatrix} C_2 \\ D_{22}C_K \end{bmatrix}$$

Prove that the following are equivalent:

1.  $(\bar{A}, \bar{B}, \bar{C}, \bar{D})$  is stabilizable and detectable.
2.  $(A, B_2, C_2, D_{22})$  and  $(A_K, B_K, C_K, D_K)$  are stabilizable and detectable.

### 3 Problem 3

For the feedback configuration in Figure 2, prove that:

1. If  $K$  is stable then the closed loop interconnection is stable if and only if  $G_{22}(I - KG_{22})^{-1}$  is stable.
2. If  $G_{22}$  is stable then the closed loop interconnection is stable if and only if  $K(I - G_{22}K)^{-1}$  is stable.

## 4 Problem 4

Consider the standard negative feedback loop with the nominal plant dynamics  $P(s) = \frac{2}{s+1}$  and controller  $K(s) = 20$ . Assume the “true” dynamics lie within the following multiplicative uncertainty set:

$$\mathcal{M} := \left\{ \hat{P} = P(1 + W_u \Delta) : \|\Delta\|_\infty < 1 \text{ and } \Delta \text{ stable} \right\}$$

Assume the uncertainty weight is  $W_u(s) = \frac{2s+1}{s+10}$ .

1. Provide an interpretation for the uncertainty describe by the weight  $W_u$ .
2. Is the nominal feedback system stable? What are the gain and phase margins of the nominal loop  $L = PK$ ?
3. The robust stability condition for this type of multiplicative uncertainty is stated as:  $K$  stabilizes all  $\hat{P} \in \mathcal{M}$  if and only if  $\|W_u T\|_\infty \leq 1$ . Does  $K$  robustly stabilize all models in  $\mathcal{M}$  based on this condition?
4. We can construct the uncertainty set  $\mathcal{M}$  in MATLAB using the following commands:

```
>> Delta = ultidyn(0Delta0; [1 1]);
>> Phat = P * (1 + Wu * Delta);
>> Lhat = Phat * K;
>> That = feedback(Lhat; 1);
```

The ultidyn command constructs an uncertain, LTI transfer function object that satisfies  $\|\Delta\|_\infty < 1$ . The function inputs are the name and input/output size of the object. Construct the uncertain model Phat using the commands above. Generate a Bode magnitude plot with 10 samples drawn from the uncertainty set and draw the nominal response P on the same plot. Note: The command bodemag(Phat) will automatically generate 10 samples and draw their plots.

5. Finally, we can perform the robustness test using the following command:

```
>> [stabmarg; destabunc; report] = robstab(That)
```

Refer to the function documentation for robstab for a short description of the input/output arguments. Does the result obtained with robstab agree with your conclusions in part (c)?

## **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>