

Analysis of MIMO System

① ★ Norms

⇒ A norm is a mathematical tool to measure "lengths".

⇒ Let X be a vector space. $\|\cdot\|: X \rightarrow \mathbb{R}$ is a norm if it satisfies the following properties:

① Non-negativity: $\|x\| \geq 0$

② Positive definiteness: $\|x\| = 0 \iff x = 0$

③ Homogeneity: $\|\alpha x\| = |\alpha| \|x\| \quad \forall \alpha \in \mathbb{C}$

④ Triangle inequality: $\|x+y\| \leq \|x\| + \|y\| \quad \forall x, y \in X$

* Vectors norm

Let $x = [x_1, \dots, x_n]^T \in \mathbb{C}^n$

① P-th norm

$$\|x\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}} \quad \forall p \in \mathbb{N}^+ = \{1, 2, \dots\}$$

② Two norm (Euclidean norm)

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

③ Infinity norm

$$\|x\|_\infty = \max_i |x_i|$$

* Matrix norms

⇒ Let $G \in \mathbb{C}^{d \times d}$, with elements g_{ij}

① Frobenius matrix norm

$$\|G\|_F = \sqrt{\sum_{ij} |g_{ij}|^2}$$

② Max element norm

$$\|G\|_{\max} = \max_{i,j} |g_{ij}|$$

* Induced matrix norms

↳ Quantify the maximum gain (amplification) of output vector for any possible input direction.

① Induced p-norm

$$\|G\|_{i,p} = \max_{\omega \neq 0} \frac{\|G\omega\|_p}{\|\omega\|_p}$$

② Induced 2-norm

$$\|G\|_{i,2} = \max_{\omega \neq 0} \frac{\|G\omega\|_2}{\|\omega\|_2} = \sqrt{\max_j |\lambda_j(G^* G)|}$$

③ Induced infinity norm

$$\|G\|_{i,\infty} = \max_i \left(\sum_j |g_{ij}| \right)$$

\Rightarrow All matrix norms satisfy:

$$\|AB\| \leq \|A\| \|B\|$$

* Signal norms

Let $e(t) = [e_1(t) \dots e_m(t)]^T$, $e_i(t) \in \mathbb{C}$, $i=1, \dots, n$

① P-th norm

$$\|e(t)\|_p = \left(\int_{-\infty}^{\infty} \sum_{i=1}^m |e_i(\tau)|^p d\tau \right)^{\frac{1}{p}}$$

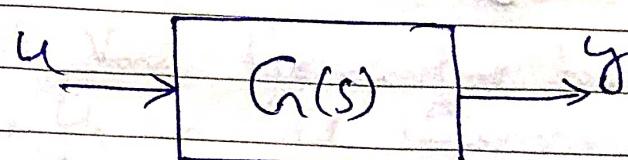
② Two norm ("Energy of the signal")

$$\|e(t)\|_2 = \sqrt{\int_{-\infty}^{\infty} \sum_{i=1}^m |e_i(\tau)|^2 d\tau}$$

③ Infinity norm

$$\|e(t)\|_\infty = \sup_{\tau} (\max_i |e_i(\tau)|)$$

* System norms



① 2-norm

- A measure of the combination of system gains in all directions over all frequencies.
- Does not respect the multiplicative property.

$$\|G(s)\|_2 = \sqrt{\frac{1}{2\pi} \int_{-\infty}^{\infty} \|G(j\omega)\|_F^2 d\omega}$$

② Infinity norm

- A measure of the peak of the maximum singular value (i.e. the biggest amplification the system may bring at any frequency for any input direction).
- Respects the multiplicative property.

$$\|G(s)\|_\infty = \sup_{\omega} \|G(j\omega)\|_{1,2}$$

② MIMO Analysis (directionality)

* Singular value Decomposition (SVD)

→ Let $G \in \mathbb{C}^{lxm}$

→ It can always be written as

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

Where:

$$U = [U_1, \dots, U_p] \in \mathbb{C}^{lxl} \quad \begin{matrix} \text{Unitary} \\ \text{Orthogonal} \end{matrix}$$

$$V = [V_1, \dots, V_m] \in \mathbb{C}^{m \times m} \quad \begin{matrix} \text{Unitary} \\ \text{Orthogonal} \end{matrix}$$

$$\Sigma = \begin{bmatrix} \sigma_1 & & 0 \\ & \ddots & \\ 0 & & 0 \end{bmatrix} \quad \text{with } \Sigma = \begin{bmatrix} G_1 & & 0 \\ & G_2 & \\ 0 & & G_p \end{bmatrix}$$

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_p \quad \{P \text{ is rank of } G\}$$

$$\sigma_i = \sqrt{\lambda_i(G^* G)}$$

⇒ V_1, V_2, \dots, V_m forms the basis of input direction and U_1, U_2, \dots, U_p forms the basis of output direction.

⇒ Singular values

↳ Are directional gains along each input/output directions.

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

$$\Rightarrow Gv_i = U \Sigma$$

$$\Rightarrow Gv_i = \sigma_i u_i$$

$$\sigma_i = \frac{\|Gv_i\|_2}{\|u_i\|_2}$$

* SVD observations

⇒ For any input, the input-output's two-norm ratio is bounded by the singular values.

$$\underline{\sigma}(A) \leq \frac{\|Au\|_2}{\|u\|_2} \leq \overline{\sigma}(A)$$

Condition number

↳ Coarse measure of how "directional" a MIMO plant is.

$$\kappa(A) = \frac{\overline{\sigma}(A)}{\underline{\sigma}(A)} \geq 1$$

⇒ The higher the condition number is, the more difficult it is (typically) to control a MIMO plant.

* SVD Computation recipe

1. Find eigenvalues of G^*G
2. Calculate singular values as the roots of the eigenvalues found in ①.
 - ↳ Name the singular values in decreasing order such that $\sigma_1 \geq \sigma_2 \geq \dots$
3. Build the singular value matrix, recalling it has same dimension of G .
 - ↳ Place the singular values on the main diagonal in decreasing order.
4. Compute the right singular vectors V_i as eigen vectors of G^*G
 - ↳ Make matrix unitary (vector orthogonal) by using Gram-Schmidt if necessary.
5. Compute the left singular vectors as

$$U_i = \frac{1}{\sigma_i} G V_i$$

* Poles and zeros direction

(a) Zero

$$0 = G(z) = [U_1 \ U_2] \begin{bmatrix} G_1 > 0 & 0 \\ 0 & G_2 = 0 \end{bmatrix} [V_1 \ V_2]^*$$

Output direction Input direction

(b) Poles

$$G(p+\epsilon) \underset{\epsilon \rightarrow 0}{\approx} [U_1 \ U_2] \begin{bmatrix} G_1 \gg 0 & 0 \\ 0 & G_2 \ll G_1 \end{bmatrix} [V_1 \ V_2]^*$$

Output direction Input direction

