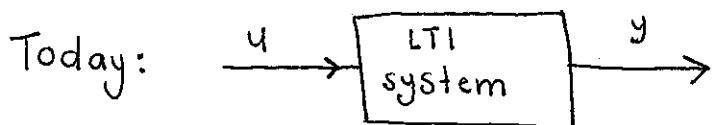


## Make-up Lecture 11/8



Consider finite energy input systems

$$\Rightarrow \|U\|_2^2 = \int_0^\infty \|U(t)\|_2^2 dt = \int_0^\infty U^T(t) \cdot U(t) dt = \int_0^\infty \sum_{i=1}^m |u_i(t)|^2 dt < \infty$$

output:

$$y(t) = \int_0^t \underbrace{H(t-\tau)}_{\text{impulse response}} \cdot U(\tau) d\tau$$

↓ fourier transform

$$Y(j\omega) = H(j\omega) \cdot U(j\omega)$$

↪ study energy of output:  $\|y\|_2^2 = \int_0^\infty y^T(t) \cdot y(t) dt$

and remember → transforms preserve norms

$$\|y\|_2^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} Y^*(j\omega) Y(j\omega) d\omega$$

↪ Parsevals equality

$$\text{ex/ } Y = \begin{bmatrix} 1 + 3j \\ 2 - 4j \end{bmatrix} \Rightarrow Y^* = (\bar{Y})^T = \begin{bmatrix} 1 - 3j & 2 + 4j \end{bmatrix}$$

$$\Rightarrow \|y\|_2^2 = \int_{-\infty}^{+\infty} U^*(j\omega) H^*(j\omega) H(j\omega) U(j\omega) d\omega$$

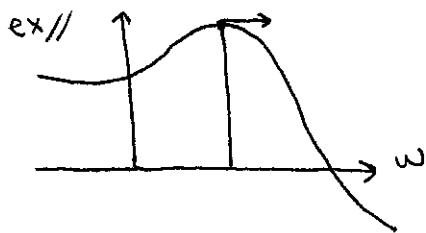
In SISO case,  $U(j\omega), Y(j\omega), H(j\omega) \in \mathbb{C}$  at each  $\omega$

$$\Rightarrow \|y\|_2^2 = \frac{1}{2\pi} \int_{-\infty}^{+\infty} |H(j\omega)|^2 \cdot |U(j\omega)|^2 \leq \sup_\omega |H(j\omega)|^2 \cdot \frac{1}{2\pi} \int_{-\infty}^{+\infty} |U(j\omega)|^2 d\omega$$

$$\leq \sup_\omega |H(j\omega)|^2 \cdot \|U\|_2^2$$

$$\text{If } \|U\|_2^2 \neq 0 \Rightarrow \frac{\|y\|_2^2}{\|U\|_2^2} = \frac{\text{output energy}}{\text{input energy}}$$

$$\leq \underbrace{\sup_\omega |H(j\omega)|^2}_{\text{peak rating on Bode mag. plot}}$$



$$H(s) = \frac{k}{s+1} \Rightarrow \sup_{\omega} |H(j\omega)|^2 = k^2$$

Want to determine "worst case" ratio of  $\|y\|_2^2/\|u\|_2^2$

$$\sup_{\omega} \|y\|_2^2/\|u\|_2^2 ; 0 < \|u\|_2^2 \leq 1$$

[street lingo: "maximize output/input (energy) under constraint that 0 ≠ input energy ≤ 1"]

(turns out)  $\Rightarrow \sup_{\omega} |H(j\omega)|^2$

BIG Q: What do we do in MIMO case?

(what is equivalent of  $\sup_{\omega} |H(j\omega)|^2$  in MIMO system?)

$$\Rightarrow \boxed{\sup_{\omega} \sigma_{\max}^2 (H(j\omega))}$$

↳ largest singular value of  $H(j\omega)$

BIG ASIDE: Singular Value Decomposition (of matrix M)

but only  $M \in \mathbb{C}^{n \times n}$  square  $\left\{ \begin{array}{l} \rightarrow \text{for square matrix we can do e-value decomposition} \\ M \cdot \mathbf{v}_i = \lambda_i \mathbf{v}_i \end{array} \right.$

$\rightarrow$  If M is Hermitian ( $M = M^*$ )  $\Rightarrow V^{-1} = V^*$  (unitary diag.)  
 $\Rightarrow M = V \cdot \Lambda \cdot V^{-1} \equiv V \cdot \Lambda \cdot V^*$

Q: how do we figure out "personality" of rectangular matrix?  
(e-value decomp reveals personality of square matrix)  
↳ Fact of Life (oh + Linear Algebra)

ANY matrix  $M \in \mathbb{C}^{m \times n}$  can be represented by:

$$m \left\{ \underbrace{\begin{bmatrix} M \\ \vdots \\ m \end{bmatrix}}_n \right\} \equiv \left\{ \underbrace{\begin{bmatrix} U \\ \vdots \\ m \end{bmatrix}}_n \left[ \underbrace{\Sigma}_{n} \right] \left[ \begin{bmatrix} V^* \\ \vdots \\ n \end{bmatrix} \right] \right\}_n$$

$\Sigma$ -matrix of singular values

$$\text{where } UU^* = U^*U = I_m \\ VV^* = V^*V = I_n$$