

Input-output norms of LTI systems

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Amplification of disturbances

- **Harmonic forcing**

$$d(t) = \hat{d}(\omega) e^{j\omega t} \xrightarrow{\text{steady-state response}} y(t) = \hat{y}(\omega) e^{j\omega t}$$

- ★ **Frequency response**

$$\hat{y}(\omega) = \underbrace{C (j\omega I - A)^{-1} B}_{H(\omega)} \hat{d}(\omega)$$

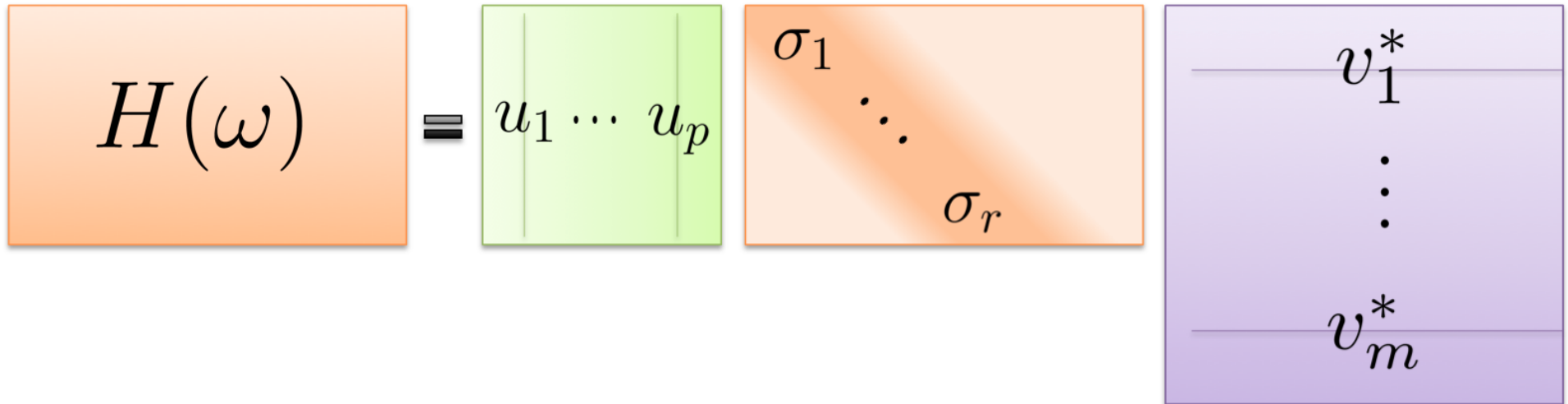
example: 3 inputs, 2 outputs

$$\begin{bmatrix} \hat{y}_1(\omega) \\ \hat{y}_2(\omega) \end{bmatrix} = \begin{bmatrix} H_{11}(\omega) & H_{12}(\omega) & H_{13}(\omega) \\ H_{21}(\omega) & H_{22}(\omega) & H_{23}(\omega) \end{bmatrix} \begin{bmatrix} \hat{d}_1(\omega) \\ \hat{d}_2(\omega) \\ \hat{d}_3(\omega) \end{bmatrix}$$

$H_{ij}(\omega)$ – response from j th input to i th output

Input-output gains

- Determined by **singular values** of $H(\omega)$



left and **right** singular vectors:

$$H(\omega)H^*(\omega) u_i(\omega) = \sigma_i^2(\omega) u_i(\omega)$$

$$H^*(\omega)H(\omega) v_i(\omega) = \sigma_i^2(\omega) v_i(\omega)$$

$\{u_i\}$ orthonormal basis of output space

$\{v_i\}$ orthonormal basis of input space

- **Action of $H(\omega)$ on $\hat{d}(\omega)$**

$$\hat{y}(\omega) = H(\omega) \hat{d}(\omega) = \sum_{i=1}^r \sigma_i(\omega) \mathbf{u}_i(\omega) \langle \mathbf{v}_i(\omega), \hat{d}(\omega) \rangle$$

- **Right singular vectors**

★ **identify input directions with simple responses**

$$\sigma_1(\omega) \geq \sigma_2(\omega) \geq \dots > 0$$

$$\hat{y}(\omega) = \sum_{i=1}^r \sigma_i(\omega) \mathbf{u}_i(\omega) \langle \mathbf{v}_i(\omega), \hat{d}(\omega) \rangle$$

$$\downarrow \hat{d}(\omega) = \mathbf{v}_k(\omega)$$

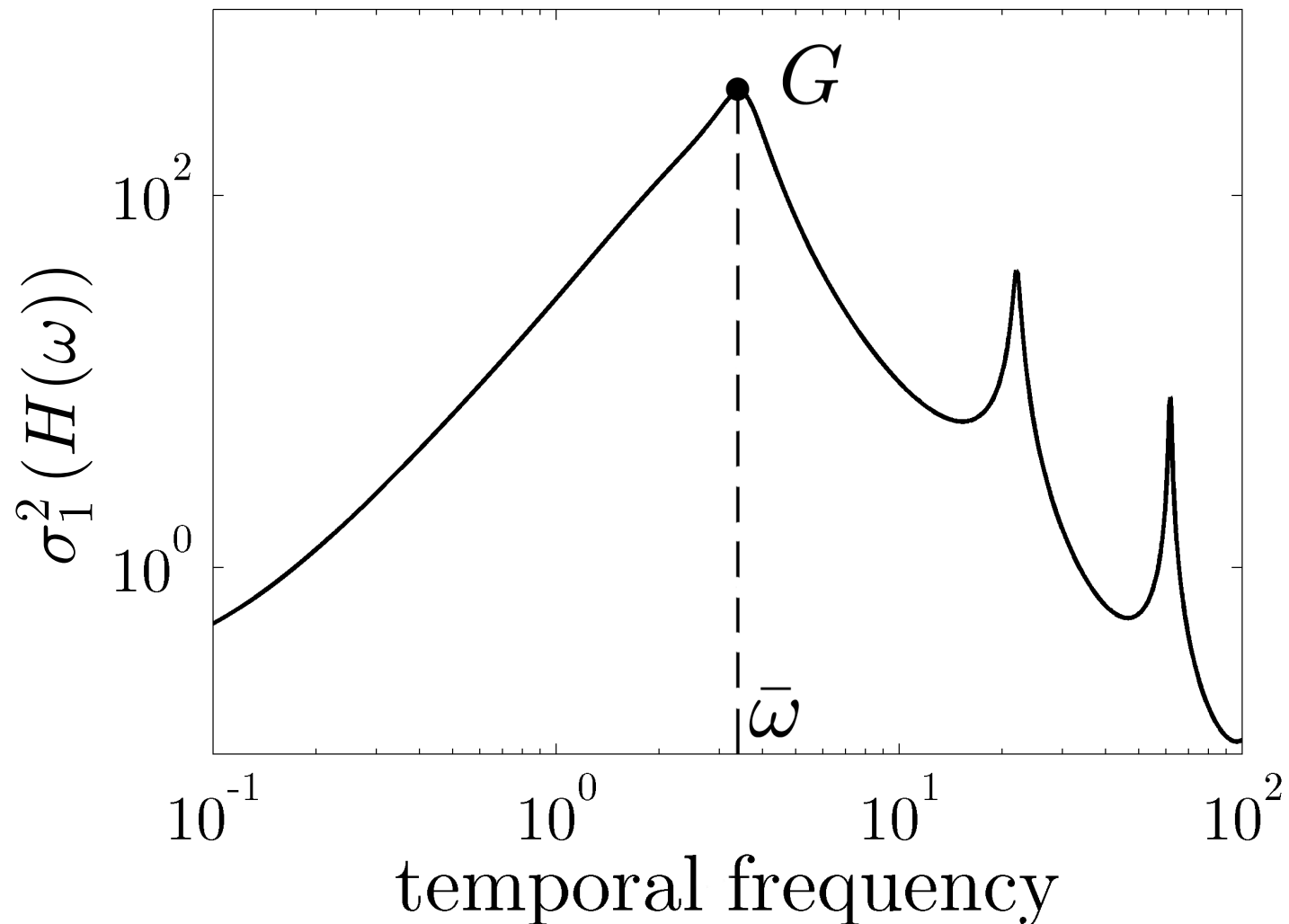
$$\hat{y}(\omega) = \sigma_k(\omega) \mathbf{u}_k(\omega)$$

$\sigma_1(\omega)$: the largest amplification at any frequency

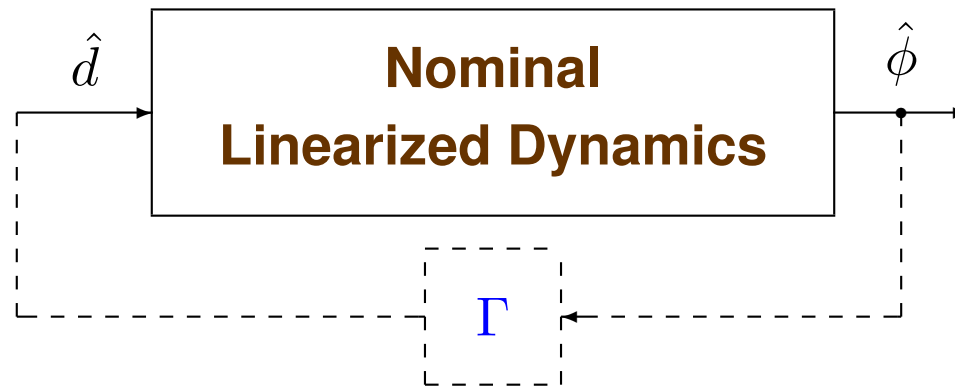
Worst case amplification

- H_∞ norm: an induced L_2 gain (of a system)

$$G = \|H\|_\infty^2 = \max \frac{\text{output energy}}{\text{input energy}} = \max_{\omega} \sigma_1^2(H(\omega))$$



Robustness interpretation: small-gain theorem



modeling uncertainty

(can be nonlinear or time-varying)

- Closely related to **pseudospectra** of linear operators

$$\dot{x}(t) = (A + B \Gamma C) x(t)$$

LARGE
worst case amplification



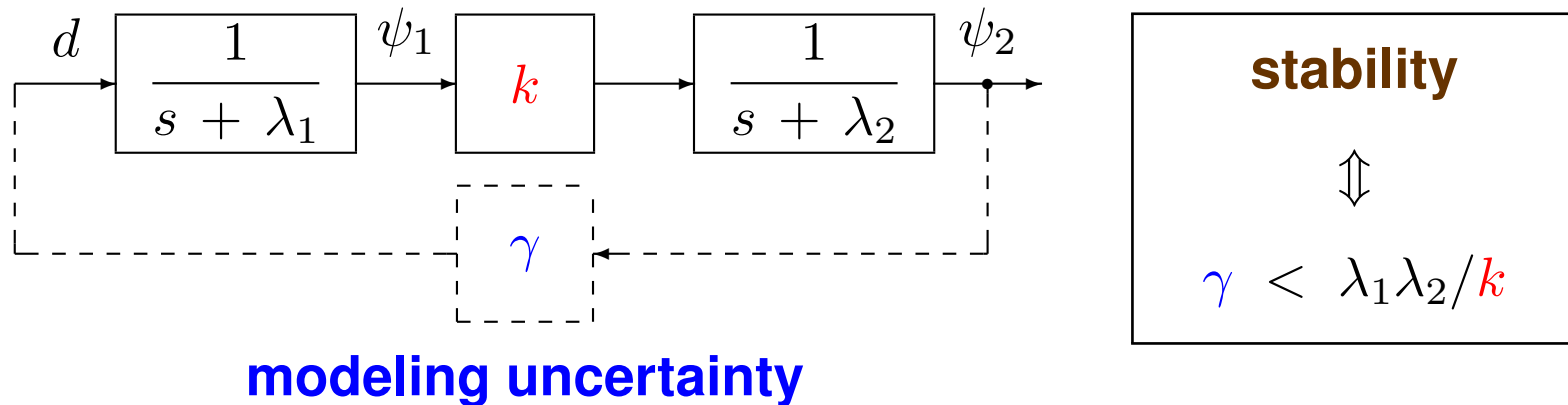
small
stability margins

Back to a toy example

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -\lambda_1 & \mathbf{0} \\ k & -\lambda_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} d$$

$$G = \max_{\omega} |H(j\omega)|^2 = \frac{k^2}{(\lambda_1 \lambda_2)^2}$$

ROBUSTNESS



$$\det \left(\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} -\lambda_1 & \gamma \\ k & -\lambda_2 \end{bmatrix} \right) = s^2 + (\lambda_1 + \lambda_2) s + \underbrace{(\lambda_1 \lambda_2 - \gamma k)}_{> 0}$$

Response to stochastic forcing

- **White-in-time forcing**

$$\mathcal{E} (d(t_1) d^*(t_2)) = I \delta(t_1 - t_2)$$

- ★ **Frobenius norm**

power spectral density:

$$\|H(\omega)\|_F^2 = \text{trace} (H(\omega) H^*(\omega)) = \sum_{i=1}^r \sigma_i^2 (\omega)$$

- ★ **H_2 norm**

variance amplification:

$$\|H\|_2^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} \|H(\omega)\|_F^2 d\omega = \int_0^{\infty} \|H(t)\|_F^2 dt$$

Computation of H_2 and H_∞ norms

$$\begin{aligned}\dot{x}(t) &= A x(t) + B d(t) \\ y(t) &= C x(t)\end{aligned}$$

- H_2 norm

- ★ Lyapunov equation

$$\mathcal{E}(d(t_1) d^*(t_2)) = W \delta(t_1 - t_2) \Rightarrow \begin{cases} \|H\|_2^2 = \text{trace}(C P C^*) \\ A P + P A^* = -B W B^* \end{cases}$$

- H_∞ norm

- ★ E-value decomposition of Hamiltonian in conjunction with bisection

$$\|H\|_\infty \geq \gamma \Leftrightarrow \begin{bmatrix} A & \frac{1}{\gamma} B B^* \\ -\frac{1}{\gamma} C^* C & -A^* \end{bmatrix} \text{ has at least one imaginary e-value}$$