

# EECS 16B CSM

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CSM

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Stability

Feedback

Least Squares

# Stability

# Continuous

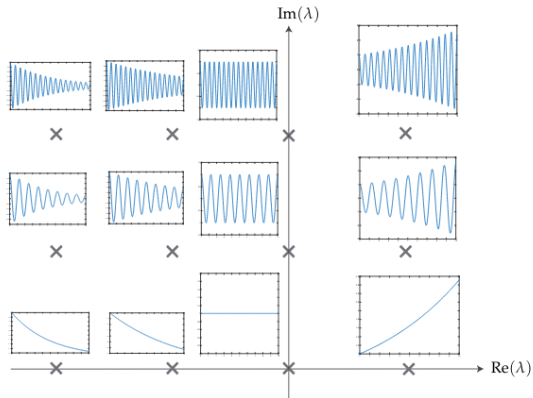
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# Discrete

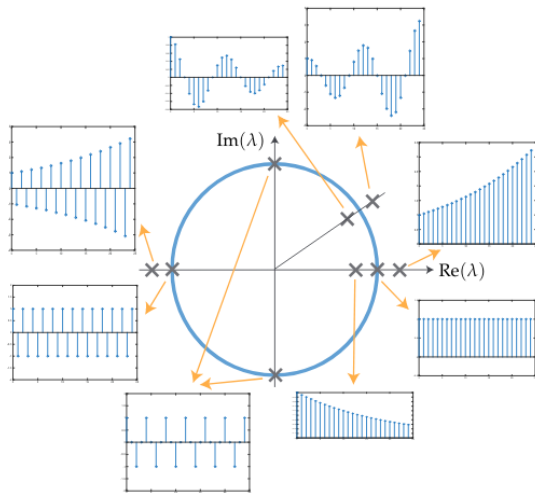
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# Feedback

# Open-Loop

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$$\mathbf{x}[t + 1] = \mathbf{A}\mathbf{x}[t] + \mathbf{B}\mathbf{u}[t] \quad (1)$$

- define a certain range of use
- simpler
- no restraints apart from stability

# Closed-Loop

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$$\mathbf{u}[t] = \begin{bmatrix} k_1 & k_2 & \cdots & k_n \end{bmatrix} \begin{bmatrix} x_1[t] \\ x_2[t] \\ \vdots \\ x_n[t] \end{bmatrix} \quad (2)$$

$$\mathbf{x}[t+1] = \mathbf{A}\mathbf{x}[t] + \mathbf{B}\mathbf{K}\mathbf{x}[t] = (\mathbf{A} + \mathbf{B}\mathbf{K})\mathbf{x}[t] \quad (3)$$

- adaptable to a wide range of use
- more complex
- self-correcting
- requires more constraints

# Controller Canonical Form

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$$\mathbf{x}[t+1] = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ a_1 & a_2 & a_3 & \cdots & a_n \end{bmatrix} \mathbf{x}[t] + \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \mathbf{u}[t] \quad (4)$$

$$\det(\mathbf{A} + \mathbf{BK} - \lambda \mathbf{I}) = \lambda^n - \sum_{i=1}^n (a_i + k_i) \lambda^i \quad (5)$$



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# Least Squares

# Quick Review

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$$\mathbf{A}\mathbf{x} = \mathbf{b} \implies \hat{\mathbf{x}} \approx ([\mathbf{A}]^{\top}\mathbf{A})^{-1} [\mathbf{A}]^{\top}\mathbf{b} \quad (6)$$

- we want to minimize the error vector  $\mathbf{e} = \mathbf{b} - \mathbf{A}\hat{\mathbf{x}}$