

Control Systems And Their Components (EE391)

Lec. 6: SS Dynamic Solution and Realizations

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Lecture Outline

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- ❑ Diagonalization of the system dynamics matrix **A**
- ❑ Going from SS model to TF
- ❑ Relationship between poles of TF and eigenvalues of **A**
- ❑ Equivalent SS equations
- ❑ SS realizations

State Space Equations (Reminder)

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For an n dimensional system with p inputs and m outputs

$$\begin{array}{c} n \times 1 \\ \dot{\mathbf{x}}(t) = \mathbf{A} \mathbf{x}(t) + \mathbf{B} \mathbf{u}(t) \\ \begin{array}{cc} n \times n & n \times p \end{array} \end{array}$$

$$\begin{array}{c} m \times 1 \\ \mathbf{y}(t) = \mathbf{C} \mathbf{x}(t) + \mathbf{D} \mathbf{u}(t) \\ \begin{array}{cc} m \times n & m \times p \end{array} \end{array}$$

- $t \in \mathbb{R}$ denotes time
- $\mathbf{x} \in \mathbb{R}^n$ denotes the state vector
- $\mathbf{u} \in \mathbb{R}^p$ denotes the input vector
- $\mathbf{y} \in \mathbb{R}^m$ denotes the output vector

- $\mathbf{A} \in \mathbb{R}^{n \times n}$ denotes the system dynamic matrix
- $\mathbf{B} \in \mathbb{R}^{n \times p}$ denotes the input matrix
- $\mathbf{C} \in \mathbb{R}^{m \times n}$ denotes the output or sensor matrix
- $\mathbf{D} \in \mathbb{R}^{m \times p}$ denotes the feedthrough matrix

- For LTI systems, the matrices $\mathbf{A}, \mathbf{B}, \mathbf{C}$ and \mathbf{D} are all constant, i.e. not $f(t)$
- For time variant systems $\rightarrow \mathbf{A}(t), \mathbf{B}(t), \mathbf{C}(t), \mathbf{D}(t)$

Solution of SS Equations (Time response)

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Interpretation of solution using diagonalization

Back to Homog. Sol.

$$\mathbf{x}(t) = e^{\mathbf{A}t} \mathbf{x}(0)$$

$$\begin{aligned} \mathbf{x}(t) &= \mathbf{Q} \begin{bmatrix} e^{\lambda_1 t} & & \\ & \ddots & \\ & & e^{\lambda_n t} \end{bmatrix} \mathbf{Q}^{-1} \mathbf{x}(0) \\ &= \begin{bmatrix} \vdots & & \vdots \\ \mathbf{v}_1 & \cdots & \mathbf{v}_n \\ \vdots & & \vdots \end{bmatrix} \begin{bmatrix} e^{\lambda_1 t} & & \\ & \ddots & \\ & & e^{\lambda_n t} \end{bmatrix} \begin{bmatrix} \cdots & \mathbf{w}_1^T & \cdots \\ & \vdots & \\ \cdots & \mathbf{w}_n^T & \cdots \end{bmatrix} \mathbf{x}(0) \\ &= \begin{bmatrix} \vdots & & \vdots \\ \mathbf{v}_1 & \cdots & \mathbf{v}_n \\ \vdots & & \vdots \end{bmatrix} \begin{bmatrix} e^{\lambda_1 t} & & \\ & \ddots & \\ & & e^{\lambda_n t} \end{bmatrix} \begin{bmatrix} \mathbf{w}_1^T \mathbf{x}(0) \\ \vdots \\ \mathbf{w}_n^T \mathbf{x}(0) \end{bmatrix} \end{aligned}$$

Solution of SS Equations (Time response)

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Interpretation of solution using diagonalization

Back to Homog. Sol.

$$\mathbf{x}(t) = e^{\mathbf{A}t} \mathbf{x}(0)$$

$$\mathbf{x}(t) = \begin{bmatrix} \vdots & & \vdots \\ \mathbf{v}_1 & \cdots & \mathbf{v}_n \\ \vdots & & \vdots \end{bmatrix} \begin{bmatrix} e^{\lambda_1 t} & & \\ & \ddots & \\ & & e^{\lambda_n t} \end{bmatrix} \begin{bmatrix} \mathbf{w}_1^T \mathbf{x}(0) \\ \vdots \\ \mathbf{w}_n^T \mathbf{x}(0) \end{bmatrix}$$

$$= \begin{bmatrix} \vdots & & \vdots \\ \mathbf{v}_1 & \cdots & \mathbf{v}_n \\ \vdots & & \vdots \end{bmatrix} \begin{bmatrix} e^{\lambda_1 t} \mathbf{w}_1^T \mathbf{x}(0) \\ \vdots \\ e^{\lambda_n t} \mathbf{w}_n^T \mathbf{x}(0) \end{bmatrix}$$

$$= \sum_{i=1}^n \mathbf{v}_i e^{\lambda_i t} \left(\mathbf{w}_i^T \mathbf{x}(0) \right)$$

Solution of SS Equations (Time response)

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Interpretation of solution using diagonalization

Back to Homog. Sol. $\mathbf{x}(t) = e^{\mathbf{A}t} \mathbf{x}(0)$

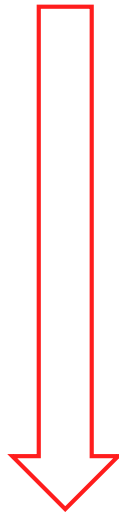
$$\mathbf{x}(t) = \sum_{i=1}^n \mathbf{v}_i e^{\lambda_i t} \left(\mathbf{w}_i^T \mathbf{x}(0) \right)$$

- Solution is a linear combination of all individual modes ($e^{\lambda_i t}$)
- Eigenvalues λ_i determine the time behavior of each mode
- Eigenvectors \mathbf{v}_i determine how much each mode impacts each of the state variables
- Rows of \mathbf{Q}^{-1} , denoted by \mathbf{w}_i^T , determine how much each initial state variable contribute to each mode
- Benefit of diagonalization or eigen decomposition is to decouple the modes and write the full time solution as a linear combination of them
- You can also expect that **eigenvalues are related to poles**

From SS model to Transfer Function

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SS model



Transfer Function

$$\begin{array}{l} \overset{n \times 1}{\dot{\mathbf{x}}(t)} = \underset{n \times n}{\mathbf{A}} \overset{n \times 1}{\mathbf{x}(t)} + \underset{n \times p}{\mathbf{B}} \overset{p \times 1}{\mathbf{u}(t)} \\ \overset{m \times 1}{\mathbf{y}(t)} = \underset{m \times n}{\mathbf{C}} \overset{n \times 1}{\mathbf{x}(t)} + \underset{m \times p}{\mathbf{D}} \overset{p \times 1}{\mathbf{u}(t)} \end{array}$$

$$\frac{Y(s)}{U(s)}$$

From SS model to Transfer Function

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$$\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t)$$

Take LT $s\mathbf{X}(s) - \mathbf{x}(0) = \mathbf{A}\mathbf{X}(s) + \mathbf{B}\mathbf{U}(s)$

$$(s\mathbf{I} - \mathbf{A})\mathbf{X}(s) = \mathbf{x}(0) + \mathbf{B}\mathbf{U}(s)$$

$$\mathbf{X}(s) = (s\mathbf{I} - \mathbf{A})^{-1} \mathbf{x}(0) + (s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B}\mathbf{U}(s)$$

since $\mathbf{y}(t) = \mathbf{C}\mathbf{x}(t) + \mathbf{D}\mathbf{u}(t)$

$$\therefore \mathbf{Y}(s) = \mathbf{C}\mathbf{X}(s) + \mathbf{D}\mathbf{U}(s)$$

$$\therefore \mathbf{Y}(s) = \mathbf{C}\mathbf{X}(s) + \mathbf{D}\mathbf{U}(s)$$

$$\mathbf{Y}(s) = \underbrace{\mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{x}(0)}_{\text{Initial state response}} + \underbrace{\left[\mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D} \right]}_{\text{Transfer Function Matrix } m \times p} \mathbf{U}(s)$$

Initial state response

Transfer Function Matrix $m \times p$

From SS model to Transfer Function

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$$\mathbf{Y}(s) = \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{x}(0) + \left[\mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D} \right] \mathbf{U}(s)$$

To obtain TF, set $\mathbf{x}(0) = 0$

$$\mathbf{Y}(s) = \left[\mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D} \right] \mathbf{U}(s)$$

For a SISO system, $\mathbf{Y}(s)$ and $\mathbf{U}(s)$ are scalars

$$\frac{Y(s)}{U(s)} = \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D}$$

MATLAB

tf2ss

ss2tf

From SS model to Transfer Function

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Example

Find the transfer function of the following state space model

$$\dot{\mathbf{x}} = \begin{bmatrix} 4 & -5 \\ 2 & -3 \end{bmatrix} \mathbf{x} + \begin{bmatrix} -2 \\ 1 \end{bmatrix} u$$
$$y = [-0.5 \quad 1] \mathbf{x}$$

Solution

$$\frac{Y(s)}{U(s)} = \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D}$$

From a previous example with the same matrix \mathbf{A}

$$(s\mathbf{I} - \mathbf{A})^{-1} = \begin{bmatrix} \frac{s+3}{(s-2)(s+1)} & \frac{-5}{(s-2)(s+1)} \\ \frac{2}{(s-2)(s+1)} & \frac{s-4}{(s-2)(s+1)} \end{bmatrix}$$

From SS model to Transfer Function

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Example

$$\begin{aligned}\frac{Y(s)}{U(s)} &= \begin{bmatrix} -0.5 & 1 \end{bmatrix} \begin{bmatrix} \frac{s+3}{(s-2)(s+1)} & \frac{-5}{(s-2)(s+1)} \\ \frac{2}{(s-2)(s+1)} & \frac{s-4}{(s-2)(s+1)} \end{bmatrix} \begin{bmatrix} -2 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} -0.5 & 1 \end{bmatrix} \begin{bmatrix} \frac{-2s-1}{(s-2)(s+1)} \\ \frac{s-8}{(s-2)(s+1)} \end{bmatrix} \\ &= \frac{2s-7.5}{(s-2)(s+1)}\end{aligned}$$

We notice that the poles at $s = 2, -1$ are exactly the eigenvalues of \mathbf{A} we found before

From SS model to Transfer Function

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HW problem

Find the transfer function matrix of the following SS model having 3 inputs and 2 outputs

$$\dot{\mathbf{x}} = \begin{bmatrix} 4 & -5 \\ 2 & -3 \end{bmatrix} \mathbf{x} + \begin{bmatrix} -2 & 3 & 1 \\ 1 & -5 & 0 \end{bmatrix} \mathbf{u}$$
$$\mathbf{y} = \begin{bmatrix} -0.5 & 1 \\ -1 & 2 \end{bmatrix} \mathbf{x}$$

Hint

You should still find $[\mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{B} + \mathbf{D}]$ which is a 2x3 matrix that relates the input and output vectors as follows

$$\mathbf{Y}(s) = \left[\mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{B} + \mathbf{D} \right] \mathbf{U}(s)$$

$$\begin{bmatrix} Y_1(s) \\ Y_2(s) \end{bmatrix} = \left[\mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{B} + \mathbf{D} \right] \begin{bmatrix} U_1(s) \\ U_2(s) \\ U_3(s) \end{bmatrix}$$

From SS model to Transfer Function

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Back to SISO case and general TF

$$\frac{Y(s)}{U(s)} = \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D}$$

$$\frac{Y(s)}{U(s)} = \mathbf{C} \frac{\text{adj}(s\mathbf{I} - \mathbf{A})}{|s\mathbf{I} - \mathbf{A}|} \mathbf{B} + \mathbf{D}$$

$$= \frac{\mathbf{C} \text{adj}(s\mathbf{I} - \mathbf{A}) \mathbf{B} + \mathbf{D} |s\mathbf{I} - \mathbf{A}|}{|s\mathbf{I} - \mathbf{A}|}$$

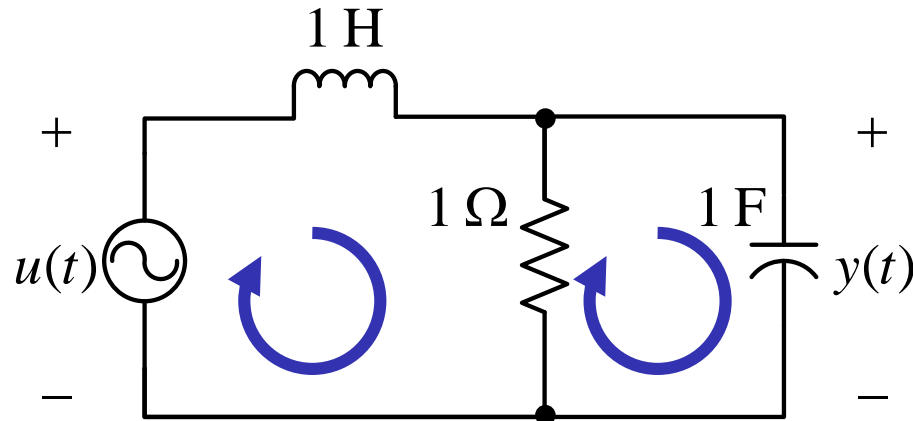
- Clearly the poles of the TF are the values of s that makes $|s\mathbf{I} - \mathbf{A}| = 0$ which are the also the eigenvalues of \mathbf{A}
- We can easily predict that what was said on poles can be exactly said on eigenvalues (e.g. the condition of BIBO stability is $\text{Re}\{\text{eigenvalues}\} < 0$)

Equivalent SS Equations

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Example

Find two SS representations for the this circuit. Use the underneath two assignments of state variables



State variables (1st)

- x_1 : inductor current i_L
- x_2 : capacitor voltage v_c

State variables (2nd)

- \tilde{x}_1 : current of left loop
- \tilde{x}_2 : current of right loop

Find the relation (transformation \mathbf{P}) between the two state vectors in the 1st and 2nd realizations

$$\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x}$$

Equivalent SS Equations

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State variables (1st)

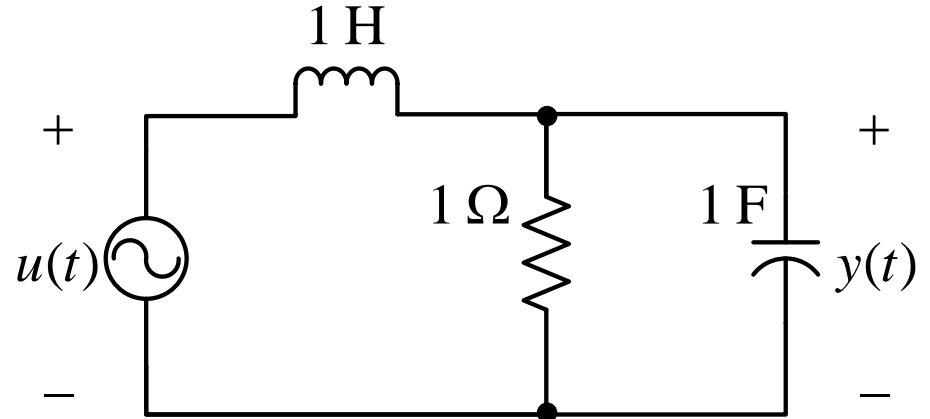
- x_1 : inductor current i_L
- x_2 : capacitor voltage v_C

$$u = \dot{x}_1 + x_2 \Rightarrow \dot{x}_1 = -x_2 + u$$

$$x_1 = x_2 + \dot{x}_2 \Rightarrow \dot{x}_2 = x_1 - x_2$$

$$\therefore \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u$$

$$y = \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + 0u$$

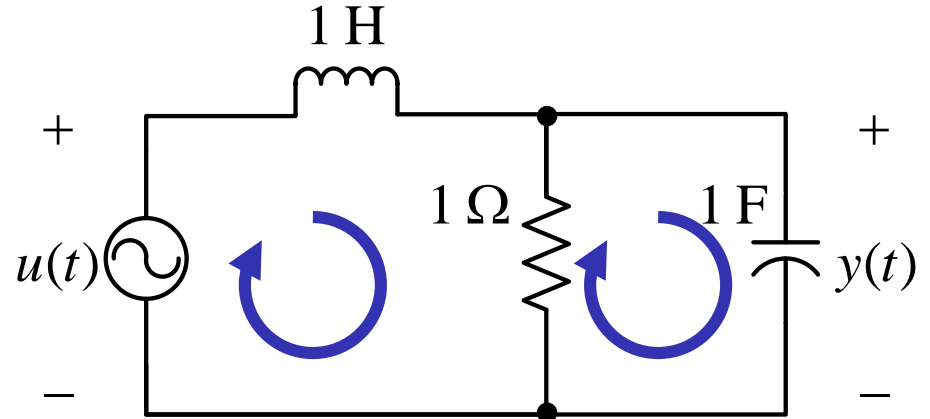


Equivalent SS Equations

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State variables (2nd)

- \tilde{x}_1 : current of left loop
- \tilde{x}_2 : current of right loop



$$u = \dot{\tilde{x}}_1 + (\tilde{x}_1 - \tilde{x}_2) \Rightarrow \dot{\tilde{x}}_1 = -\tilde{x}_1 + \tilde{x}_2 + u$$

$$\tilde{x}_2 = 1 \cdot \frac{d}{dt} \{1 \cdot (\tilde{x}_1 - \tilde{x}_2)\} \Rightarrow \dot{\tilde{x}}_1 - \dot{\tilde{x}}_2 = \tilde{x}_2 \Rightarrow \dot{\tilde{x}}_2 = -\tilde{x}_1 + u$$

$$\therefore \begin{bmatrix} \dot{\tilde{x}}_1 \\ \dot{\tilde{x}}_2 \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u$$

$$y = [1 \quad -1] \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} + 0u$$

The two State vectors are equivalent and can be related by the transformation \mathbf{P}

$$\begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x}$$

Equivalent SS Equations

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- Consider an n -dimensional state space equations:

$$\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t)$$

$$\mathbf{y}(t) = \mathbf{C}\mathbf{x}(t) + \mathbf{D}\mathbf{u}(t)$$

- Let \mathbf{P} be an $n \times n$ real nonsingular matrix, and let $\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x}$. Then, the SS equations

$$\dot{\tilde{\mathbf{x}}}(t) = \tilde{\mathbf{A}}\tilde{\mathbf{x}}(t) + \tilde{\mathbf{B}}\mathbf{u}(t)$$

$$\mathbf{y}(t) = \tilde{\mathbf{C}}\tilde{\mathbf{x}}(t) + \tilde{\mathbf{D}}\mathbf{u}(t)$$

where

$$\tilde{\mathbf{A}} = \mathbf{P}\mathbf{A}\mathbf{P}^{-1}, \quad \tilde{\mathbf{B}} = \mathbf{P}\mathbf{B}, \quad \tilde{\mathbf{C}} = \mathbf{C}\mathbf{P}^{-1}, \quad \tilde{\mathbf{D}} = \mathbf{D}$$

is said to be algebraically equivalent with the original state space equations

- $\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x}$ is called an equivalence (or similarity) transformation

Equivalent SS Equations

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start from $\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t)$

$\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x} \Rightarrow$ substitute with $\mathbf{x} = \mathbf{P}^{-1}\tilde{\mathbf{x}}$

$$\mathbf{P}^{-1}\dot{\tilde{\mathbf{x}}}(t) = \mathbf{A}\mathbf{P}^{-1}\tilde{\mathbf{x}}(t) + \mathbf{B}\mathbf{u}(t)$$

$$\dot{\tilde{\mathbf{x}}}(t) = \underbrace{\mathbf{P}\mathbf{A}\mathbf{P}^{-1}}_{\tilde{\mathbf{A}}}\tilde{\mathbf{x}}(t) + \underbrace{\mathbf{P}\mathbf{B}}_{\tilde{\mathbf{B}}}\mathbf{u}(t)$$

start from $\mathbf{y}(t) = \mathbf{C}\mathbf{x}(t) + \mathbf{D}\mathbf{u}(t)$

$\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x} \Rightarrow$ substitute with $\mathbf{x} = \mathbf{P}^{-1}\tilde{\mathbf{x}}$

$$\mathbf{y}(t) = \underbrace{\mathbf{C}\mathbf{P}^{-1}}_{\tilde{\mathbf{C}}}\tilde{\mathbf{x}}(t) + \underbrace{\mathbf{D}}_{\tilde{\mathbf{D}}}\mathbf{u}(t)$$

Equivalent SS Equations

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Problem

Prove that the similarity transformation $\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x}$ does not change the eigenvalues of \mathbf{A} , i.e. prove that eigenvalues of \mathbf{A} are the same as $\tilde{\mathbf{A}}$

Problem

Prove that the similarity transformation $\tilde{\mathbf{x}} = \mathbf{P}\mathbf{x}$ does not change the Transfer function matrix

$$\begin{aligned}\tilde{\mathbf{T}} &= \tilde{\mathbf{C}}(s\mathbf{I} - \tilde{\mathbf{A}})^{-1} \tilde{\mathbf{B}} + \tilde{\mathbf{D}} \\ &= \mathbf{C}\mathbf{P}^{-1} (s\mathbf{I} - \mathbf{P}\mathbf{A}\mathbf{P}^{-1})^{-1} \mathbf{P}\mathbf{B} + \mathbf{D} \\ &= \mathbf{C}\mathbf{P}^{-1} (s\mathbf{P}\mathbf{P}^{-1} - \mathbf{P}\mathbf{A}\mathbf{P}^{-1})^{-1} \mathbf{P}\mathbf{B} + \mathbf{D} \\ &= \mathbf{C}\mathbf{P}^{-1} \left[\mathbf{P}(s\mathbf{I} - \mathbf{A})\mathbf{P}^{-1} \right]^{-1} \mathbf{P}\mathbf{B} + \mathbf{D} \\ &= \mathbf{C}\mathbf{P}^{-1} \mathbf{P}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{P}^{-1} \mathbf{P}\mathbf{B} + \mathbf{D} \\ &= \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D} = \mathbf{T}\end{aligned}$$

SS Realizations

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- The problem concerning how to describe a system in state space equations, provided that the transfer function of a system, $G(s)$, is available, is called **Realization Problem**.

$$G(s) \longrightarrow A, B, C, D$$

- There are infinite realizations for the same TF
- We are only interested in minimal realizations (least number of state variables n)
- There are standard realizations that we will study
 - Controller Canonical Form (CCF)
 - Observer Canonical Form (OCF)
 - Modal (Diagonal) Canonical Form (DCF)

SS Realizations

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Controller Canonical Form (CCF)

Assume order of numerator is less than that of denominator by 1

$$\begin{aligned} G(s) &= \frac{Y(s)}{U(s)} = \frac{b_{n-1}s^{n-1} + b_{n-2}s^{n-2} + \dots + b_1s + b_0}{s^n + a_{n-1}s^{n-1} + a_{n-2}s^{n-2} + \dots + a_1s + a_0} \\ &= (b_{n-1}s^{n-1} + b_{n-2}s^{n-2} + \dots + b_1s + b_0) \times \frac{1}{s^n + a_{n-1}s^{n-1} + a_{n-2}s^{n-2} + \dots + a_1s + a_0} \\ &= \frac{Y(s)}{Z(s)} \quad \times \quad \frac{Z(s)}{U(s)} \end{aligned}$$

For the first part

$$\frac{Z(s)}{U(s)} = \frac{1}{s^n + a_{n-1}s^{n-1} + a_{n-2}s^{n-2} + \dots + a_1s + a_0}$$

$$z^{(n)}(t) + a_{n-1}z^{(n-1)}(t) + \dots + a_1\dot{z}(t) + a_0z(t) = u(t)$$

SS Realizations

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Controller Canonical Form (CCF)

For the first part

$$\frac{Z(s)}{U(s)} = \frac{1}{s^n + a_{n-1}s^{n-1} + a_{n-2}s^{n-2} + \dots + a_1s + a_0}$$

$$z^{(n)}(t) + a_{n-1}z^{(n-1)}(t) + \dots + a_1\dot{z}(t) + a_0z(t) = u(t)$$

$$\text{let } x_1 = z, \quad x_2 = \dot{z}, \quad x_3 = \ddot{z}, \quad \dots, \quad x_n = z^{(n-1)}$$

$$\therefore \dot{x}_1 = x_2, \quad \dot{x}_2 = x_3, \quad \dot{x}_3 = x_4, \dots,$$

$$\begin{aligned} \dot{x}_n = z^{(n)} &= -a_{n-1}z^{(n-1)} - \dots - a_1\dot{z} - a_0z + u \\ &= -a_{n-1}x_n - \dots - a_1x_2 - a_0x_1 + u \end{aligned}$$

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \\ \vdots \\ \dot{x}_n(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ -a_0 & -a_1 & \dots & -a_{n-3} & -a_{n-2} & -a_{n-1} \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \\ \vdots \\ x_n(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} u(t)$$

SS Realizations

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Controller Canonical Form (CCF)

For the second part

$$\frac{Y(s)}{Z(s)} = (b_{n-1}s^{n-1} + b_{n-2}s^{n-2} + \dots + b_1s + b_0)$$

$$\begin{aligned} y(t) &= b_{n-1}z^{(n-1)}(t) + b_{n-2}z^{(n-2)}(t) + \dots + b_1\dot{z}(t) + b_0z(t) \\ &= b_{n-1}x_n + b_{n-2}x_{n-1} + \dots + b_1x_2 + b_0x_1 \end{aligned}$$

$$y(t) = [b_0 \quad b_1 \quad b_2 \quad \dots \quad b_{n-2} \quad b_{n-1}] \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix} + 0 \cdot u(t)$$

SS Realizations

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Controller Canonical Form (CCF)

$$G(s) = \frac{Y(s)}{U(s)} = \frac{b_{n-1}s^{n-1} + b_{n-2}s^{n-2} + \dots + b_1s + b_0}{s^n + a_{n-1}s^{n-1} + a_{n-2}s^{n-2} + \dots + a_1s + a_0}$$

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \\ \vdots \\ \dot{x}_n(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ -a_0 & -a_1 & \dots & -a_{n-3} & -a_{n-2} & -a_{n-1} \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \\ \vdots \\ x_n(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} u(t)$$

$$y(t) = [b_0 \quad b_1 \quad b_2 \quad \dots \quad b_{n-2} \quad b_{n-1}] \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix} + 0 \cdot u(t)$$

SS Realizations

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Diagonal Canonical Form (DCF)

$$G(s) = \frac{Y(s)}{U(s)} = \left\{ \sum_{i=1}^n \frac{r_i}{s - \lambda_i} + r_0 \right\} U(s)$$

Use Partial
Fraction Expansion
for TF

- In case **all poles are distinct**, we define:

$$X_1(s) = \frac{1}{s - \lambda_1} U(s) \longrightarrow \dot{x}_1(t) = \lambda_1 x_1(t) + u(t)$$

$$X_2(s) = \frac{1}{s - \lambda_2} U(s) \longrightarrow \dot{x}_2(t) = \lambda_2 x_2(t) + u(t)$$

$$\vdots$$
$$X_n(s) = \frac{1}{s - \lambda_n} U(s) \longrightarrow \dot{x}_n(t) = \lambda_n x_n(t) + u(t)$$

$$Y(s) = r_1 X_1(s) + r_2 X_2(s) + \cdots + r_n X_n(s) + r_0 U(s) \longrightarrow y(t) = r_1 x_1(t) + r_2 x_2(t) + \cdots + r_n x_n(t) + r_0 u(t)$$

SS Realizations

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Diagonal Canonical Form (DCF)

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \vdots \\ \dot{x}_n(t) \end{bmatrix} = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} r_1 & r_2 & \cdots & r_n \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix} + r_0 u(t)$$

- Matrix \mathbf{A} for DCF is diagonal, i.e. all state variables are decoupled
- DCF can be obtained by finding the eigenvalues λ 's and eigenvector matrix \mathbf{Q}
- Having distinct eigenvalues is exactly the same as having distinct poles (non-repeated poles) in which case \mathbf{A} is diagonalizable

SS Realizations

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Observer Canonical Form (OCF)

$$G(s) = \frac{Y(s)}{U(s)} = \frac{b_{n-1}s^{n-1} + b_{n-2}s^{n-2} + \dots + b_1s + b_0}{s^n + a_{n-1}s^{n-1} + \dots + a_1s + a_0}$$

$$s^n Y(s) + a_{n-1}s^{n-1}Y(s) + \dots + a_1sY(s) + a_0Y(s) = \\ b_{n-1}s^{n-1}U(s) + b_{n-2}s^{n-2}U(s) + \dots + b_1sU(s) + b_0U(s)$$

$$Y(s) + a_{n-1} \frac{Y(s)}{s} + \dots + a_1 \frac{Y(s)}{s^{n-1}} + a_0 \frac{Y(s)}{s^n} = \\ b_{n-1} \frac{U(s)}{s} + b_{n-2} \frac{U(s)}{s^2} + \dots + b_1 \frac{U(s)}{s^{n-1}} + b_0 \frac{U(s)}{s^n}$$

$$Y(s) = \frac{1}{s} \left\{ (b_{n-1}U(s) - a_{n-1}Y(s)) + \frac{1}{s} \left\{ (b_{n-2}U(s) - a_{n-2}Y(s)) + \frac{1}{s} (\dots) + \right. \right. \\ \left. \left. \frac{1}{s} \{ b_0U(s) - a_0Y(s) \} \right\} \dots \right\}$$

$X_1(s)$

SS Realizations

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Observer Canonical Form (OCF)

$$X_1(s) = \frac{1}{s} \{b_0 U(s) - a_0 Y(s)\} \longrightarrow \dot{x}_1(t) = b_0 u(t) - a_0 y(t)$$

$$X_2(s) = \frac{1}{s} \{(b_1 U(s) - a_1 Y(s)) + X_1(s)\} \longrightarrow \dot{x}_2(t) = b_1 u(t) - a_1 y(t) + x_1(t)$$

\vdots

\vdots

$$X_n(s) = \frac{1}{s} \{(b_{n-1} U(s) - a_{n-1} Y(s)) + X_{n-1}(s)\} \longrightarrow \dot{x}_n(t) = b_{n-1} u(t) - a_{n-1} y(t) + x_{n-1}(t)$$

$$Y(s) = X_n(s) \longrightarrow y(t) = x_n(t)$$

SS Realizations

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Observer Canonical Form (OCF)

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \vdots \\ \dot{x}_n(t) \end{bmatrix} = \begin{bmatrix} 0 & 0 & \cdots & -a_0 \\ 1 & 0 & \cdots & -a_1 \\ \vdots & \ddots & & \vdots \\ 0 & 0 & 1 & -a_{n-1} \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix} + \begin{bmatrix} b_0 \\ b_1 \\ \vdots \\ b_{n-1} \end{bmatrix} u(t)$$
$$y(t) = [0 \quad \cdots \quad 0 \quad 1] \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix} + 0u(t)$$

SS Realizations

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Example

Find the SS realization of the following TF in CCF, DCF, OCF

$$G(s) = \frac{4s^3 + 25s^2 + 45s + 34}{2s^3 + 12s^2 + 20s + 16}$$

And prove that all 3 realizations are equivalent

Solution

$$G(s) = \frac{4s^3 + 25s^2 + 45s + 34}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{s^2 + 5s + 2}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{\frac{1}{2}s^2 + 2\frac{1}{2}s + 1}{s^3 + 6s^2 + 10s + 8}$$

SS Realizations

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Solution

$$G(s) = \frac{4s^3 + 25s^2 + 45s + 34}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{s^2 + 5s + 2}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{\frac{1}{2}s^2 + 2\frac{1}{2}s + 1}{s^3 + 6s^2 + 10s + 8}$$

CCF

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -8 & -10 & -6 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} 1 & 2.5 & 0.5 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + 2 \cdot u(t)$$

SS Realizations

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Solution

$$G(s) = \frac{4s^3 + 25s^2 + 45s + 34}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{s^2 + 5s + 2}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{\frac{1}{2}s^2 + 2\frac{1}{2}s + 1}{s^3 + 6s^2 + 10s + 8}$$

OCF

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \end{bmatrix} = \begin{bmatrix} 0 & 0 & -8 \\ 1 & 0 & -10 \\ 0 & 1 & -6 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + \begin{bmatrix} 1 \\ 2.5 \\ 0.5 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + 2 \cdot u(t)$$

SS Realizations

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Solution

$$G(s) = \frac{4s^3 + 25s^2 + 45s + 34}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{s^2 + 5s + 2}{2s^3 + 12s^2 + 20s + 16} = 2 + \frac{\frac{1}{2}s^2 + 2\frac{1}{2}s + 1}{s^3 + 6s^2 + 10s + 8}$$

DCF

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \end{bmatrix} = \begin{bmatrix} -1+j & 0 & 0 \\ 0 & -1-j & 0 \\ 0 & 0 & -4 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} 0.3 + j0.15 & 0.3 - j0.15 & -0.1 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + 2 \cdot u(t)$$

Use MATLAB to find partial fraction expansion

```
[r,p,k] = residue([4 25 45 34],[2 12 20 16])
```