

MATLAB #4 Reachability, Controllability

1) Input \rightarrow discrete-time
 \rightarrow continuous-time
 \rightarrow Optimality

2) Controllability canonical form

3) Kalman decomposition

If a system is completely reachable:

$$\mathcal{R}^+ \subseteq \mathcal{R}^- \quad \begin{matrix} \text{completely} \\ \text{reachable} \end{matrix} \Rightarrow \begin{matrix} \text{completely} \\ \text{controllable} \end{matrix}$$

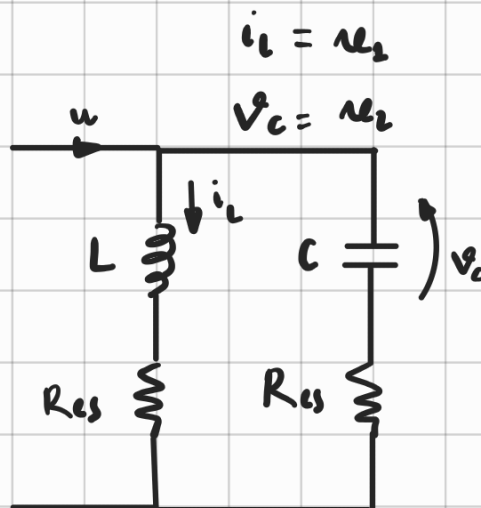
The inverse holds only for reversible system:

- 1.) discrete \rightarrow sampled data
- 2.) continuous \rightarrow reversible

Example: (RLC)

$$A = \begin{bmatrix} -2R_{es}/L & 1/L \\ -1/C & 0 \end{bmatrix}$$

$$B = \begin{bmatrix} R_{es}/L \\ 1/C \end{bmatrix}$$



$$R = \begin{bmatrix} R_{es}/L & -2R_{es}^2/L^2 + 1/LC \\ \underbrace{-1/C}_B & \underbrace{-R_{es}/LC}_{AB} \end{bmatrix} \quad \text{rank}(R) = 1$$

If $\text{rank}(R) = 2$, $\mathcal{R}^+ \subset \mathbb{R}^2$.

For $\sigma(A) = \lambda$ with $a = 2$ and $g = 1$, the Jordan block would be:

$$J = \begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix} \rightarrow e^{Jt} = \begin{bmatrix} e^{\lambda t} u_1(0) + t e^{\lambda t} u_2(0) & u_2(t) \\ e^{\lambda t} u_2(0) & u_2(t) \end{bmatrix}$$

Discrete-time control input

$$u_t = A^t u_0 + R_t \begin{bmatrix} u_{t-1} \\ \vdots \\ u_0 \end{bmatrix}$$

$$u_t = \underbrace{-\lambda^t u_0}_{\text{a point } \in \mathbb{R}^n} \in \underbrace{\text{image}(R_t)}_{\text{a subspace}}$$

linear variety

$$1) \underline{u}_{tf} - A^{tf} \underline{u}_0 = R_{tf} \underline{u}_{tf-1} \quad (\text{particular solution})$$

$$2) 0 = R_{tf} \cdot \underline{v} \quad (\text{homogenous solution})$$



Solution:

$$1) (R_{tf} R_{tf}^T) \underline{\eta} = \underline{u}_{tf} - A^{tf} \underline{u}_0$$

$$2) \underline{u}^* = R_{tf}^T \underline{\eta}$$

$\forall \underline{\eta}_1, \underline{\eta}_2$ solution of 1)
 $\underline{\eta}_1 - \underline{\eta}_2 \in \text{kernel}(R_{tf} R_{tf}^T)$

$$\boxed{R_{tf}^T \underline{\eta}_1 = R_{tf}^T \underline{\eta}_2}$$

Let $\underline{u}_0 \rightarrow \underline{u}_f$ as "minimum energy input"

$$\|\underline{u}\| = \langle \underline{u}, \underline{u} \rangle^{1/2} = \left(\sum_{k=0}^{t_f-1} \underline{u}_k^T \underline{u}_k \right)^{1/2}$$

Proof (Minimum energy) $y = A \underline{u}$

$$\underline{u}^* = R_{tf}^T \underline{\eta} \in \text{image}(R_{tf}^T) \\ \equiv \text{kernel}(R_{tf})^\perp$$

Every other solution $\underline{u}^* + v$, $v \in \text{kernel}(R_{tt})$

$$\|\underline{u}^* + v\|^2 = \|\underline{u}^*\|^2 + \|v\|^2$$

Chickens, Wolves, Bonns

$$\text{Let } A = \begin{bmatrix} 0.8 & -0.2 & -0.1 \\ 0.1 & 0.5 & 0.1 \\ 0.1 & -0.1 & 0.5 \end{bmatrix} \quad B = \begin{bmatrix} 2 & 0 & -0.1 & 0 \\ 0 & 0.3 & -0.4 & -0.5 \\ 0.5 & 0.2 & -0.5 & -0.4 \end{bmatrix}$$

For a discrete-time system: $R_t = n$,

$$R_1 = [B] = n$$

$$R_3 = [\dots AB A^{t-1}] = n$$

Continuous-time system: $R_t = n$
finite

Kalman Decomposition

$$A_c = \begin{bmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \\ -\alpha_0 & \dots & \alpha_{n-1} \end{bmatrix} \quad B_c = \begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}$$

$$\sigma(A_c) = \lambda^n + \alpha_{n-1} \lambda^{n-1} + \dots + \alpha_0$$

In cont.-time system, it is completely reachable \Leftrightarrow completely controllable \Rightarrow stabilizable.

$\exists K : (A + BK)$ is Hurwitz, $\forall \lambda \in \sigma(A + BK)$,
real part of $\{\lambda\} < 0$

$K = \text{place}(A, B, [\quad])$
 \hookrightarrow obs. eigenval.

$$u = Kx$$

$$y = 1x$$

$$A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$B = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

$$u = Kx$$

$$R = \begin{bmatrix} 1 & 1 & 3 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$$

$$\det(R) = -2$$

$$\text{rank}(R) = 3$$

$$B \quad AB \quad A^2B$$

$$1) \sigma_A(\lambda) = \sigma_{A_c}(\lambda)$$

$$\det(\lambda I - A) = \underset{a_0}{1} - \underset{a_1}{\lambda} - \underset{a_2}{\lambda^2} + \lambda^3$$

$$A_c = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & 1 & 1 \end{bmatrix}$$

$$B_c = \begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}$$

$$2) T_c = R_c R^{-1}$$

$$\sigma(A + BK) = \{-2, -2, -5\}$$

$$\det(\lambda I - A - BK) = \{K_i\}$$

► Create a char. polynom.

$$P(A+BK)(\lambda) = (\lambda+2)^2(\lambda+5)$$

$$= 20 + 24\lambda + 9\lambda^2 + \lambda^3$$

$$K_c = [K_{c0} \quad K_{c1} \quad K_{c2}]$$

$$-20 = -1 + K_{c0} \quad -9 = 1 + K_{c1}$$

$$A_c + B_c K_c = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1+K_{c0} & 1+K_{c1} & 1+K_{c2} \end{bmatrix} \quad B_c = \begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}$$

$-d_0 \quad -d_1 \quad -d_2$

$$K_{c0} = -19$$

$$K_{c1} = -25$$

$$K_{c2} = -10$$

$$K_c = [-19 \quad -25 \quad -10]$$

$$\begin{array}{l} \lambda_1 = -2 \\ \lambda_2 = -2 \\ \lambda_3 = -5 \end{array}$$

eigvals

in $(A+BK)$

$$K = K_c T_c$$