

Lecture 4

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Diagonalization, Model Analysis, Intro to Feedback

★ Similarity Transform

⇒ Let T be an invertible matrix, and consider a coordinate transformation $X = T\tilde{X}$.

↳ This is called similarity transform.

⇒ The standard state-space model can be written as:

$$\Rightarrow \begin{cases} \dot{X} = AX + Bu \\ Y = CX + Du \end{cases} \Rightarrow \begin{cases} T\dot{\tilde{X}} = AT\tilde{X} + Bu \\ Y = CT\tilde{X} + Du \end{cases}$$

$$\Rightarrow \begin{cases} \dot{\tilde{X}} = (T^{-1}AT)\tilde{X} + (T^{-1}B)u \\ Y = (CT)\tilde{X} + Du \end{cases}$$

$$\Rightarrow \begin{cases} \dot{\tilde{X}} = \tilde{A}\tilde{X} + \tilde{B}u \\ Y = \tilde{C}\tilde{X} + \tilde{D}u \end{cases} \text{ where } \left\{ \begin{array}{l} \tilde{A} = T^{-1}AT \\ \tilde{B} = T^{-1}B \\ \tilde{C} = CT \\ \tilde{D} = D \end{array} \right.$$

⇒ The choice of a state-space model for a given system is not unique.

★ Controllability and Observability

⇒ A LTI system of the form $\dot{x} = Ax + Bu$ is said to be **controllable** if for any given initial state $x(0) = x_0$ there exists a control signal that takes the state to the origin $x(t) = 0$ for some finite time t .

⇒ A LTI system of the form $\dot{x} = Ax + Bu$, $y = Cx + Du$ is said to be **observable** if any given initial condition $x(0) = x_0$ can be reconstructed based on the knowledge of the output and input signal only over a finite time interval $[0, t]$.

⇒ An LTI system is **stabilizable** if all unstable modes are controllable.

⇒ An LTI system is **detectable** if all unstable modes are observable.

★ Diagonalization

⇒ ~~Let's~~ Lets assume we have n independent eigenvectors; then we can assemble the eigenvectors into an invertible matrix V whose columns are the eigenvectors v_i .

$$V = [v_1 \ v_2 \ \dots \ v_n] \quad \Lambda = \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \ddots \\ & & & \lambda_n \end{bmatrix}$$

So $AV = VA$

* Modal Coordinates

⇒ Eigenvalue and eigen vectors of A define the modes of the system.

⇒ The transformed coordinates $X(t) = V\tilde{X}$ are called modal coordinates.

⇒ The eigen vector V_i defines the shape of the i th mode.

⇒ The eigenvalue λ_i defines how the amplitude of the mode evolves over time.

⇒ Let $\dot{\tilde{X}}(t) = A\tilde{X}(t)$ be the system.

⇒ Let $V = [V_1, V_2, \dots, V_n]$ be the eigen vector matrix.

⇒ Let $\Lambda = \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \ddots \\ & & & \lambda_n \end{bmatrix}$ be eigen value matrix.

⇒ Let $X(t) = V\tilde{X}(t)$

⇒ $\dot{\tilde{X}} = (V^{-1}AV)\tilde{X} \quad \left\{ \Lambda = V^{-1}AV \right\}$

⇒ $\dot{\tilde{X}}(t) = \Lambda\tilde{X}(t)$

⇒ $\tilde{X}(t) = e^{\Lambda t} \tilde{X}(0)$

⇒ $V^{-1}X(t) = \cancel{e^{\Lambda t} V^{-1}X(0)} e^{\Lambda t} \tilde{X}(0)$

$$X(t) = V e^{At} \tilde{X}(0)$$

$$\Rightarrow [V_1 \ V_2 \ \dots \ V_m] \begin{bmatrix} e^{\lambda_1 t} & & \\ & e^{\lambda_2 t} & \\ & & \ddots \\ & & & e^{\lambda_m t} \end{bmatrix} \begin{bmatrix} \tilde{X}_1(0) \\ \tilde{X}_2(0) \\ \vdots \\ \tilde{X}_m(0) \end{bmatrix}$$

$$\Rightarrow \sum_{i=1}^m e^{\lambda_i t} \tilde{X}_i(0) V_i = X(t)$$

\Rightarrow Any trajectory can be expressed as a linear combination of modes.

★ Pole placement - First order systems

\Rightarrow Consider a first order control system with dynamics $\dot{x} = ax + bu$, and assume that we are not happy about its behaviour.

\Rightarrow Feedback control: Choose $u = -Kx$, then the dynamics would be

$$\dot{x} = (a - bK)x$$

\Rightarrow As long as $b \neq 0$ (i.e. the system is controllable)

by choosing $K = (a - a^*)/b$, we can place the "closed-loop" eigenvalue at a desired value a^* , on anywhere we want on the real axis!

⇒ This is the simplest example of a general technique called "pole placement".

★ Effect on feedback for closed-loop dynamics

⇒ If we have an open-loop LTI system

$$\dot{X}(t) = AX(t) + Bu(t)$$

$$Y(t) = CX(t)$$

⇒ By choosing a linear feedback $u = -K$ ~~Y~~ $Y = -KCX$, we can transform it into another, closed-loop LTI system:

$$\boxed{\begin{aligned}\dot{X}(t) &= (A - BKC)X(t) \\ Y(t) &= CX(t)\end{aligned}}$$

⇒ An general "negative feedback" has stabilizing effects and the bigger K , the faster the closed-loop system is, and the smaller the error are.

⇒ However this is not generally the case.