

Representations of Dynamical Systems

System Representations

- Input/Output representation

Consider relationship between input and output
Black box view

$$y(t) = g(u(t), t) \quad C.T.$$

$$y_k = g(u_k, k) \quad D.T.$$

- Input/State/Output representation

Including the systems internal states

$$\begin{cases} \dot{x}(t) = f(x(t), u(t), t) \\ y(t) = h(x(t), u(t), t) \end{cases} \quad C.T.$$

$$\begin{cases} x_{k+1} = f(x_k, u_k, k) \\ y_k = h(x_k, u_k, k) \end{cases} \quad D.T.$$

Global: describes behavior across full state-space operating range. Used for linear systems

Local: valid around a certain operating point, approximating nonlinear systems with linearization

Properties

- Causality

The output at time t depends only on inputs from time t and earlier.

- Time-invariance

Behavior doesn't depend on time

Shift in input $u(t) \rightarrow u(t + t_0)$ results in a shifted output $y(t) \rightarrow y(t + t_0)$

- Linearity

Has to satisfy two principles

1) Superposition

$$f(u_1 + u_2) = f(u_1) + f(u_2)$$

2) Homogeneity/Scaling

$$f(\alpha u) = \alpha f(u) \text{ for any } \alpha$$

Shift in Basis and Similarity

- Shift in Basis

Consider system

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t), & x(0) = x_0 \\ y(t) = Cx(t) + Du(t) \end{cases}$$

Choose a new basis defined by a non-singular matrix T , we have

$$Z(t) = T^{-1} X(t)$$

The system in the new coordinate become

$$\begin{cases} \dot{Z}(t) = T^{-1} A T Z(t) + T^{-1} B u(t), & Z(0) = T^{-1} X(0) \\ Y(t) = C T Z(t) + D u(t) \end{cases}$$

- Matrix Similarity

A and A' is said similar if there exists a non-singular matrix T s.t.:

$$A' = T^{-1} A T$$

- Similar matrices represent the same linear transformations under different bases

- same eigenvalues, characteristic polynomial, determinant and trace

- Algebraic equivalence

Two systems (A_1, B_1, C_1, D_1) and (A_2, B_2, C_2, D_2) are algebraically equivalent if there exists an invertible matrix T s.t.:

$$A_2 = T A_1 T^{-1}$$

$$B_2 = T B_1$$

$$C_2 = C_1 T^{-1}$$

$$D_2 = D_1$$

This means they represent the same input-output behavior but in different state coordinates

The matrix exponential e^{At}

Defined by power series

$$e^{At} = \sum_{k=0}^{\infty} \frac{(At)^k}{k!} = I + At + \frac{(At)^2}{2!} + \frac{(At)^3}{3!} + \dots$$

- Properties of e^{At}

$$1) e^{A(t_1+t_2)} = e^{At_1} e^{At_2} \quad (\text{linearity})$$

$$2) e^{(A+B)t} = e^{At} e^{Bt} \quad \text{only when } AB = BA$$

$$3) [e^{At}]^{-1} = e^{-At}$$

$$4) \frac{d}{dt}(e^{At}) = Ae^{At} = e^{At} A$$

5) (λ, v) pair of eigenvalue/eigenvector of A , then

$$Av = \lambda v$$

$$e^{At} v = e^{\lambda t} v$$

Lagrange formula

$$x(t) = e^{A(t-t_0)} x_0 + \int_{t_0}^t e^{A(t-\tau)} B u(\tau) d\tau$$

$$y(t) = C e^{A(t-t_0)} x_0 + \int_{t_0}^t C e^{A(t-\tau)} \beta u(\tau) d\tau + D u(t)$$

Jordan form

The Jordan form simplifies the matrix exponential for when A is not diagonalizable. A square can be decomposed into:

$$A = PJP^{-1}$$

where:

P is the matrix of generalized eigen vectors

J is the Jordan block matrix consisting of blocks J_i on the form:

$$J_i = \begin{bmatrix} \lambda_i & 1 & 0 & \cdots & 0 \\ 0 & \lambda_i & 1 & \cdots & 0 \\ 0 & 0 & \ddots & \ddots & \vdots \\ 0 & 0 & \ddots & \ddots & 1 \\ 0 & 0 & 0 & \cdots & \lambda_i \end{bmatrix} \in \mathbb{R}^{n_i \times n_i}$$

We can then express the matrix exponential as:

$$e^{At} = Pe^{Jt}P^{-1}$$

Matrix exponential computation

A diagonalizable ($\lambda_i \in \mathbb{R}$ and distinct)

$$\lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{R}$$

$$v_1, v_2, \dots, v_n \in \mathbb{R}^n$$

$$\begin{aligned}
 & Av_1 = \lambda_1 v_1 \\
 & Av_2 = \lambda_2 v_2 \\
 & \vdots \\
 & Av_n = \lambda_n v_n
 \end{aligned} \Rightarrow A \begin{bmatrix} | & | & | & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & | & | \end{bmatrix} = \begin{bmatrix} | & | & | & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & | & | \end{bmatrix} \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix}$$

$$= \begin{bmatrix} | & | & | & | \\ \lambda_1 v_1 & \lambda_2 v_2 & \cdots & \lambda_n v_n \\ | & | & | & | \end{bmatrix}$$

$$AT = T\Delta$$

$$A = T\Delta T^{-1}, \quad \Delta = T^{-1}AT$$

$$e^{At} = e^{(T\Delta T^{-1})t} = Te^{\Delta t}T^{-1}$$

ARMA Models (Auto-Regressive Moving Average)

Describes an LTI system in terms of its previous outputs (auto-regression) and a moving average of the current and past inputs

AR:

$$Y_k = \alpha_1 Y_{k-1} + \alpha_2 Y_{k-2} + \cdots + \alpha_n Y_{k-n} = \sum_{i=1}^n \alpha_i Y_{k-i}$$

MA:

$$Y_k = \beta_0 U_k + \beta_1 U_{k-1} + \cdots + \beta_m U_{k-m} = \sum_{j=0}^m \beta_j U_{k-j}$$

ARMA:

$$Y_k = \alpha_1 Y_{k-1} + \cdots + \alpha_n Y_{k-n} + \beta_0 U_k + \cdots + \beta_m U_{k-m}$$

$$= \sum_{i=1}^n \alpha_i Y_{k-i} + \sum_{j=0}^m \beta_j U_{k-j}$$

Transfer functions

A frequency-domain representation of an LTI system.

$W_{xu}(s)$: input-to-state transfer matrix

$W_{yu}(s)$: input-to-output transfer matrix

$$W_{xu}(s) = (sI - A)^{-1}B$$

$$W_{yu}(s) = C(sI - A)^{-1}B + D$$

ARMA model of system becomes:

$$x(t) = W_{xu}(s)u(t) = (sI - A)^{-1}Bu(t)$$

$$y(t) = W_{yu}(s)u(t) = [C(sI - A)^{-1}B + D]u(t)$$

More on ARMA models (where s comes from)

$$s = \frac{d}{dt}, \quad s^k = \frac{d^k}{dt^k}, \quad k=-1: \text{integrate}$$

$$(\alpha_1 + \alpha_2 s + \alpha_3 s^2)v(t) = \alpha_1 v(t) + \alpha_2 \dot{v}(t) + \alpha_3 \ddot{v}(t)$$

$$sI x(t) = \begin{bmatrix} \dot{x}_1(t) \\ \vdots \\ \dot{x}_n(t) \end{bmatrix} = \dot{x}(t) = Ax(t) + Bu(t)$$

$$\Rightarrow (sI - A)x(t) = Bu(t) \Rightarrow x(t) = (sI - A)^{-1}Bu(t)$$

Laplace Transform

Tool used for analyzing C.T. LTI systems in the frequency domain. Transforms differential equations into algebraic equations, simplifying the analysis.

For a C.T. signal $f(t)$:

$$F(s) = \mathcal{L}\{f(t)\} = \int_0^{\infty} f(t) e^{-st} dt$$

where $s = \sigma + j\omega$

Transfer functions and Laplace transform

Can derive the transfer function $H(s)$ using the Laplace transform

For the differential equation:

$$\ddot{y}(t) + a_1 \dot{y}(t) + a_2 y(t) = b_0 u(t)$$

Taking the Laplace transform:

$$s^2 Y(s) + a_1 s Y(s) + a_2 Y(s) = b_0 U(s)$$

Solving for $H(s) = \frac{Y(s)}{U(s)}$:

$$H(s) = \frac{b_0}{s^2 + a_1 s + a_2}$$

Impulse response

Characterizes how a system reacts to a single, instantaneous input - a unit impulse.

For C.T., the impulse response $h(t)$ is the output when the input is the Dirac delta function:

$$\begin{cases} \dot{x}(t) = Ax(t) + B\delta(t) \\ y(t) = Cx(t) \end{cases}$$

For D.T., same but input is called Kronecker delta;

$$\begin{cases} x_{k+1} = Ax_k + B\delta_k \\ y_k = Cx_k \end{cases}$$

Impulse response fully describes a system's behavior. If you know the impulse response, you can find the output for any input using convolution.

Convolution

C.T.:

$$y(t) = \int_0^{\infty} h(\tau) u(t-\tau) d\tau$$

D.T.:

$$y_k = \sum_{n=0}^{\infty} h_n u_{k-n}$$

Z-transform

D.T. equivalent of Laplace transform.

For a sequence $\{x_k\}$

$$X(z) = \sum_{k=0}^{\infty} x_k z^{-k}, \quad z \text{ complex variable}$$

$$z x_k = u_{k+1}, \quad z^2 u_k = u_{k+2}, \quad z^\alpha u_k = u_{k+\alpha}$$

$$u_k z^{-1} = u_{k-1}, \quad u_k z^{-\alpha} = u_{k-\alpha}$$

$$(1 + 5z + 6z^2) u_k = u_k + 5u_{k+1} + 6u_{k+2}$$

$$z I x_k = A x_k + B u_k$$

$$\Rightarrow (zI - A) x_k = B u_k$$

$$\Rightarrow x_k = (zI - A)^{-1} B u_k$$

$$\begin{cases} x_k = W_{xu}(z) u_k, & W_{xu}(z) = (zI - A)^{-1} B \\ y_k = W_{yu}(z) u_k, & W_{yu}(z) = C(zI - A)^{-1} B + D \end{cases}$$

Causality condition for $W_{yu}(z)$

$W_{yu}(z)$ causal \Leftrightarrow proper

Degree of numerator not higher than degree of denominator

$$\mathcal{Z}(x_{k+1}) = \mathcal{Z}(A x_k + B u_k)$$

$$\mathcal{Z}(y_k) = \mathcal{Z}(C x_k + D u_k)$$

Realization problems

Problem 1:

Given a transfer function $V_{ya}(\lambda)$, $\lambda \in \{\mathbb{S}, \mathbb{Z}\}$, find a state-space realization (A, B, C, D) that realizes it

$$V_{ya}(\lambda) = C(\lambda I - A)^{-1}B + D$$

Problem 2:

Given a sequence $\{w_0, w_1, \dots, w_k\}_{k=0}^{\infty}$, $w_i \in \mathbb{R} \forall i$ find a state-space realization (A, B, C, D) that realizes it

$$w_k = \begin{cases} D, & k=0 \\ CA^{k-1}B, & k \geq 1 \end{cases}$$

- Canonical realizations

Specific ways to structure the matrices A, B, C, D to emphasize different structural properties of the system.

- RCR1 (Controllable Canonical Form from Transfer function)

Built directly from the coefficients of the transfer function's denominator.

State matrix A is structured with coefficients from the characteristic polynomial

- RCO1 (Observable Canonical form from Transfer function)

Built from the coefficients of the transfer function but emphasizes observability instead.

C matrix captures the output structure clearly

- RCR2 (Controllable Canonical Form from Impulse response)

Built using the Markov parameters (impulse response sequence)

Emphasizes controllability structure, focusing on how inputs can drive states,

- RCO2 (Observable Canonical Form from Impulse response)

Also derived from Markov parameters but emphasizes how states affect outputs.

$$RCR_1 \underset{\substack{* \\ T}}{\approx} RCR_2$$

$$RCO_1 \underset{\substack{* \\ T}}{\approx} RCO_2$$

$$(A, B, C, D) \underset{T}{\approx} (T^{-1}AT, T^{-1}B, CT, D)$$

$$(A, B, C, D) \underset{*}{\approx} (A^T, C^T, B^T, D^T)$$

- Cayley - Hamilton Theorem

A square matrix satisfies its own characteristic polynomial.

If A is an $n \times n$ matrix and its characteristic polynomial

is:

$$p(\lambda) = \det(\lambda I - A) = \lambda^n + a_{n-1}\lambda^{n-1} + \dots + a_1\lambda + a_0$$

then:

$$p(A) = A^n + a_{n-1}A^{n-1} + \dots + a_1A + a_0I = 0$$

Helps in simplifying the computation of matrix exponentials in state-space solutions,

Free Response C.T. LTI Systems

Describes the natural behavior when there is no external input.

$$\dot{x}(t) = Ax(t) \quad / \quad x(t) = e^{At}x_0$$

$$y(t) = Cx(t)$$

- Eigenvalues and modes:

e^{At} can be computed using the eigenvalues and eigenvectors of A . The modes of the system correspond to the eigenmodes, and their behavior is determined by the eigenvalues.

- Eigenvalues determine growth or decay of modes

- Eigenvectors define the shape of the modes.

1) All eigenvalues real and distinct

$\lambda_1, \dots, \lambda_n$ real and distinct

v_1, \dots, v_n linearly independent

A can be written as

$$A = T \Delta T^{-1}$$

where

T : matrix of eigenvectors

Δ : $\text{diag}(\lambda_1, \dots, \lambda_n)$

Solution:

$$x(t) = T e^{\Delta t} T^{-1} x_0$$

Since Δ is diagonal:

$$e^{\Delta t} = \text{diag}(e^{\lambda_1 t}, \dots, e^{\lambda_n t})$$

Expanded form:

$$x(t) = \sum_{i=1}^n c_i v_i e^{\lambda_i t}$$

where

v_i are the eigenvectors

c_i are constants determined by the initial condition

2) Distinct complex-conjugate eigenvalues

Appear in pairs $\lambda_{1,2} = \sigma \pm i\omega$

A is not diagonalizable with real numbers, but can be handled using real-valued decomposition.

General form of solutions

$$x(t) = c_1 e^{\sigma t} \cos(\omega t) v_1 + c_2 e^{\sigma t} \sin(\omega t) v_2$$

where

- (1) determine the growth/decay rate
 (2) determine the frequency of oscillation

v_1, v_2 real-valued vectors derived from the complex eigenvectors

3) A not diagonalizable

There exists fewer eigenvectors than eigenvalues

A Jordan canonical form is used instead

A can be written as

$$A = PJP^{-1}$$

Solution:

Matrix exponential involves both exponentials and polynomials

$$x(t) = e^{\lambda t} (C_1 v + C_2 t v)$$

- Modes

	Diagonalizable	Defective
Real λ	$e^{\lambda t}$	$t^k e^{\lambda t}$ $k = 0, 1, \dots, n_i - 1$ $n_i = \text{am}(\lambda_i)$
Complex conj. $\lambda = \sigma + j\omega$	$e^{\sigma t} \cos(\omega t)$ $e^{\sigma t} \sin(\omega t)$	$t^k e^{\sigma t} \cos(\omega t)$ $t^k e^{\sigma t} \sin(\omega t)$ $k = 0, 1, \dots, n_i - t$

Convergence of modes

1) Convergent mode $m(t)$: if $\lim_{t \rightarrow \infty} m(t) = 0$

2) Bounded mode $m(t)$: $\exists M > 0$ s.t.
 $|m(t)| \leq M, \forall t$

3) Divergent mode $m(t)$: not convergent or bounded.

Free Response D_t, LTI Systems

$$x_{k+1} = Ax_k$$

$$x_k = A^k x_0$$

1) All eigenvalues real and distinct

Can diagonalize A the same way as for C.T.

$$A = T \Delta T^{-1}$$

This time the solution becomes:

$$x_k = T \Delta^k T^{-1} x_0$$

Since Δ diagonal;

$$\Delta^k = \text{diag}(\lambda_1^{(k)}, \dots, \lambda_n^{(k)})$$

2) Distinct Complex-Conjugate eigenvalues

Eigenvalues appear in pairs $\lambda_{1,2} = \sigma \pm j\omega$

Solution:

The complex eigenvalues can be written as,

$$\lambda = \rho e^{j\theta} = \rho \cos j\theta + j \rho \sin j\theta$$

where

$\rho = |\lambda|$ is the magnitude (growth/decay rate)

θ is the angle (frequency of oscillation)

The solution becomes:

$$x_k = \rho^k (c_1 \cos(k\theta) v_1 + c_2 \sin(k\theta) v_2)$$

3) A Not Diagonalizable

Use the Jordan canonical form

Solution

$$x_k = \lambda^k (c_1 v + c_2 k v)$$

- Modes

	Diagonalizable	Defective
Real λ	λ^k	λ^k $i = 0, 1, \dots, n-1$
Complex Conj. $\rho e^{\pm j\omega}$	$\rho^k \cos(\omega k)$ $\rho^k \sin(\omega k)$	$\rho^k \cos(\omega i k)$ $\rho^k \sin(\omega i k)$ $i = 0, 1, \dots, n-1$

$$|\lambda| = \rho$$

$$|\lambda| < 1 \Rightarrow m_k \text{ convergent}$$

$$|\lambda| = \begin{cases} \text{boundedness if diagonalizable} \\ \text{divergence (weak) if defective} \end{cases}$$

$|\lambda| > 1 \Rightarrow \text{divergence (strong)}$

Forced Response

Describes how a system responds to an external input

$$\begin{cases} x(t) = \int_0^t e^{A(t-\tau)} \beta u(\tau) d\tau \\ y(t) = \int_0^t (e^{A(t-\tau)} \beta u(\tau) + Du(t)) d\tau \end{cases}$$

$$\hat{y}(s) = C(SI - A)^{-1} x_0 + [C(SI - A)^{-1} \beta + D] \hat{u}(s)$$

$$y(t) = \mathcal{L}^{-1}(\hat{y}(s))$$

$$u(t) = ue^{\lambda t}, \quad \lambda \in \mathbb{R}/\{0\}, \quad u \in \mathbb{R}$$

$$\sin(\omega t) = \frac{e^{j\omega t} - e^{-j\omega t}}{2j} = \frac{1}{2j} e^{j\omega t} + \frac{-1}{2j} e^{-j\omega t}$$

$$u(t) = ue^{\lambda t} \Rightarrow \hat{u}(s) = u \frac{1}{s-\lambda}$$

Technical Lemma if $\lambda \notin \text{sp}(A)$

$$(SI - A)^{-1} \frac{1}{s-\lambda} = (\lambda I - A)^{-1} \frac{1}{s-\lambda} - (SI - A)^{-1} (\lambda I - A)^{-1}$$

$$y(t) = C e^{\lambda t} (x_0 - \hat{x}_0) + W_{yu}(\lambda) u e^{\lambda t}$$

$$x_f(t) = -e^{At} \hat{x}_0 + W_{xu}(s) u e^{st}$$

$$y_f(t) = -e^{At} \hat{x}_0 + W_{yu}(s) u e^{st}$$

↓
 Transient part ↓
 Permanent part

Step response

$$u(t) = u e^{ot} \quad (\lambda = 0)$$

$$x(t) = e^{At} (x_0 - \hat{x}_0) + W_{xa}(0) u$$

$$y(t) = C e^{At} (x_0 - \hat{x}_0) + W_{ya}(0) u$$

$$\hat{x}_0 = W_{xa}(0) u$$

$$W_{xa}(0) = W_{xa}(s) \Big|_{s=0} \quad \text{input-to-state DC gain}$$

$$W_{ya}(0) = W_{yu}(s) \Big|_{s=0} \quad \text{input-to-output DC gain}$$

For a stable system:

$$u(t) \equiv u, \quad \forall t \geq 0 \Rightarrow \lim_{t \rightarrow \infty} y(t) = W_{ya}(0) u$$

$$\lim_{t \rightarrow \infty} x(t) = W_{xa}(0) u$$

Frequency Response

Describes how a system responds to a sinusoidal input.

Provides information about how different frequencies are amplified or attenuated by the system.

$$\text{input } u(t) = e^{j\omega t}$$

steady-state output can be described using the transfer function

$$V_{yu}(s) = C(SI - A)^{-1}B + D$$

frequency response is obtained by just substituting

$$V_{yu}(j\omega) = C(j\omega I - A)^{-1}B + D$$

Magnitude response $|V_{yu}(j\omega)|$

Phase response $\angle V_{yu}(j\omega)$

$$\Rightarrow V_{yu}(j\omega) = |V_{yu}(j\omega)| e^{j\angle V_{yu}(j\omega)}$$