

Basics of Theory of Dynamic Systems

What is a system: An object, device or Phenomenon which is synthesized as an aggregation or interconnection of Parts.

Automation Engineering

Goal

How does a System evolve?

. Study how a Sequence of events happens in an autonomous way (without human intervention)

↳ Evolve : a Certain number of internal attributes of the System varies over time. (time and space)

Automation Engineering
is Composed of

. Analysis : Predict output of System knowing its inputs
(System's theory)



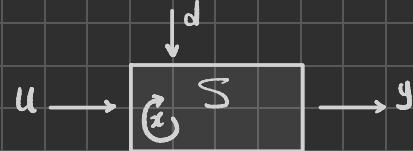
. Synthesis : obtaining a desired outputs from a System by feeding in the right inputs (Automatic Control)



. Identification : Mathematically define a System knowing only inputs and outputs.



Dynamical System An oriented System which evolves over time.



oriented means Producing an "effect" Given a "Cause"
the state 'x' internal variables

u : input independent variables

y : output dependent variables

d : disturbance 'external variables' Not a standard input for the system

Dynamic model

State Space representation :

$$\begin{array}{l} \dot{x} = f(x, u) \\ y = g(x, u) \end{array}$$

$$x = Ax + Bu \quad y = Cx + Du \quad \text{linear time-invariant}$$

. Free evolution (Zero-input response) : $\dot{x} = Ax$ & $y = Cx$

. Forced evolution (Zero-state response) : $\dot{x} = Bu$ & $y = Du$

Problem Setting

$$\dot{x} = Ax + Bu, \quad y = Cx + Du$$

. linear

. Time-invariant

. Continuous time

Dynamic matrices do not depend on time!

$$m \rightarrow F \quad F = m\ddot{v}, \text{ if no force} \rightarrow 0 = m\ddot{v} \quad \therefore \ddot{v} = 0 \Rightarrow v = v_0 \text{ initial velocity}$$

↳ Newton's 1st law

$$\ddot{v} = a = \frac{F}{m} \xrightarrow{\int} v(t) = v_0 + \frac{1}{m} \int_0^t F(\tau) d\tau$$

$$\text{double the Force: } v(t) = \frac{2}{m} \int_0^t F(\tau) d\tau$$

linear Combination of Causes implies linear combination of effects.

Stability of a linear System Study the behavior of the system under Perturbations of the initial state.

Stable → Small Perturbations Cause Small Variations.

Unstable → Small Perturbations Cause big Variations.

Equilibria: a state is an equilibrium if it causes a degenerate trajectory.

$$x_e \in \mathbb{R}^n : x_0 = x_e \Rightarrow x_t = x_e, \forall t > 0$$

$$x_e : Ax_e = 0$$

A system in equilibrium doesn't evolve → but it doesn't mean it's stable!

An equilibrium Point is stable if: $|x_0 - x_e| < \delta \Rightarrow |x(t) - x_e| < \epsilon, \forall t > 0$

If the System
starts close enough
to the equilibrium

↳ the evolution
stays close
enough to it

Asymptotic stability: An equilibrium is asymptotically stable if it is stable and $|x_0 - x_e| < \delta \Rightarrow \lim_{t \rightarrow +\infty} |x(t) - x_e| = 0$

Pendulum is example of Asymptotic stability

local Concept defined by the basin of attraction \mathcal{B}_e

if the system
starts close enough
to the equilibrium

↳ it converges
back to it
eventually

Stability and eigenvalues

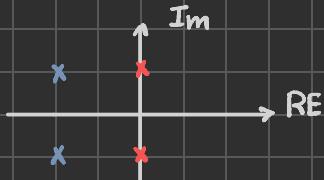
A system is **asymptotically stable** if and only if all eigenvalues have **strictly negative real Part**. $\operatorname{Re}\{\lambda_i\} < 0, \forall \lambda_i : P_A(\lambda_i) = 0$

A System is **(marginally) stable** if and only if all eigenvalues have **non Positive real part**. $\operatorname{Re}\{\lambda_i\} \leq 0, \forall \lambda_i : P_A(\lambda_i) = 0$

A System is **unstable** if and only if there is at least one eigenvalue with **Positive real Part**. $\exists \lambda_i : P_A(\lambda_i) = 0 \wedge \operatorname{Re}\{\lambda\} > 0$



asymptotic stable



marginal stable



Instability

Solution of linear System

$$x(t) = \Phi(t)x_0 + \int_0^t e^{A(t-\tau)} H(t-\tau) u(\tau) d\tau$$

$$y(t) = \Psi(t)x_0 + \int_0^t [Ce^{At} B + D\delta(t-\tau)] W(\tau) u(\tau) d\tau$$

$$x(t) = \Phi(t)x_0 + \int_0^t H(t-\tau) u(\tau) d\tau \quad \Rightarrow \quad \text{Integrals are hard to solve}$$

$$y(t) = \Psi(t)x_0 + \int_0^t W(t-\tau) u(\tau) d\tau \quad \xrightarrow{\text{Laplace Transform}} \quad F(s) = \int_0^\infty f(t) e^{-st} dt$$

Laplace Transform - Properties

linearity: $\mathcal{L}[\alpha f(t) + \beta g(t)] = \alpha \mathcal{L}[f(t)] + \beta \mathcal{L}[g(t)]$

Derivative: $\mathcal{L}\left[\frac{df(t)}{dt}\right] = s \mathcal{L}[f(t)] - f(0)$

Convolution: $\mathcal{L}\left[\int_0^t f(t-\tau) g(\tau) d\tau\right] = F(s) G(s)$

Frequency Domain

$$\dot{x} = Ax + Bu \rightarrow sX(s) - x_0 = A X(s) + B U(s)$$

$$X(s) = (sI - A)^{-1} x_0 + (sI - A)^{-1} B U(s)$$

$$Y(s) = C(sI - A)^{-1} x_0 + [C(sI - A)^{-1} B + D] U(s)$$

$W(s) = C(sI - A)^{-1} B + D$: describes the input-output behavior.
"Transfer function"

Transfer Function

$$W(s) = \frac{N(s)}{D(s)}$$

Poles are eigenvalues.
Not all eigenvalues are Poles.

An eigenvalue appears as a root (Pole) of the transfer function if :

- $\text{rank}(A - \lambda_i I | B) = n$ Controllable
- $\text{rank}\left(\frac{A - \lambda_i I}{c}\right) = n$ and Observable