

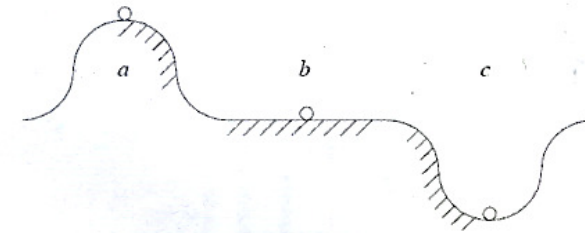
EN5101 Digital Control Systems

Lyapunov Stability

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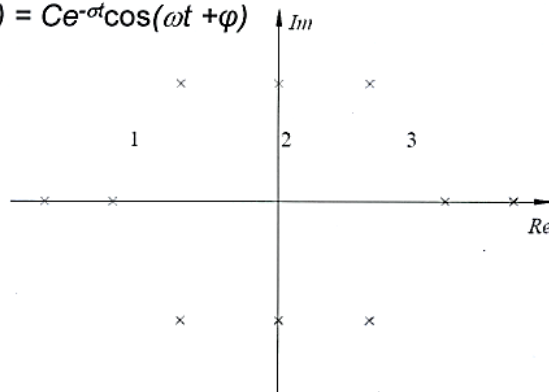
Stability Definition

- A system is stable if the output is bounded for all bounded inputs. Stability is property of a system, independent of input signal.
- Equilibrium states can be unstable equilibrium (point a), neutral equilibrium (region b), or stable equilibrium (point c); this is demonstrated in the diagram



Observer, Plant, and Control Law

- Simple test for system stability: The real part of all poles must be negative. Poles are eigenvalues of system dynamics matrix **A**
- Characteristic equation $|sI - \mathbf{A}| = 0$
- Poles, eigen values: $s_i = \sigma_i \pm j\omega_i$
- Transient solution $y(t) = Ce^{-\sigma t} \cos(\omega t + \varphi)$



Classical Methods

1. If all $\sigma_i < 0$ stable
 2. If any $\sigma_i = 0$ marginally stable (assuming the remaining σ_i are negative)
 3. If any $\sigma_i > 0$ unstable
- **Routh-Hurwitz criterion** - determine stability based on transfer function coefficients without actually calculating poles.
 - **Root-locus method** - graphical method to vary feedback gain k to determine ranges for stability and control transient response
 - **Nyquist Method** – OLTF frequency response (graphical)

Stability from Energy Point of View

- Applicable to linear/nonlinear as well as static/dynamic systems
 - If $dE/dt < 0$ then stable
 - If $dE/dt < 0$ then marginally stable
 - If $dE/dt > 0$ then unstable

Lyapunov Background

- Any time varying nonlinear system can be represented as

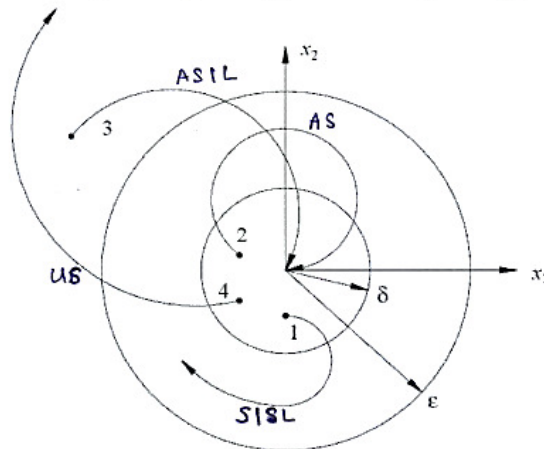
$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t)$$

$$y = \mathbf{C}\mathbf{x}$$

neglect forcing input u

- A state is an equilibrium state \mathbf{x}_e if $\mathbf{f}(\mathbf{x}_e, t) = 0$ for all t .
- For linear time invariant systems, $d\mathbf{x}/dt = \mathbf{f}(\mathbf{X}, t) = \mathbf{A}\mathbf{x}$ and there is one unique equilibrium state \mathbf{x}_e if \mathbf{A} is *nonsingular*. There can be infinitely many equilibrium states \mathbf{x}_e if \mathbf{A} is *singular*
- We can always shift an equilibrium state \mathbf{x}_e to zero by coordinate shifts: $\mathbf{f}(0, t) = 0$ for all t .

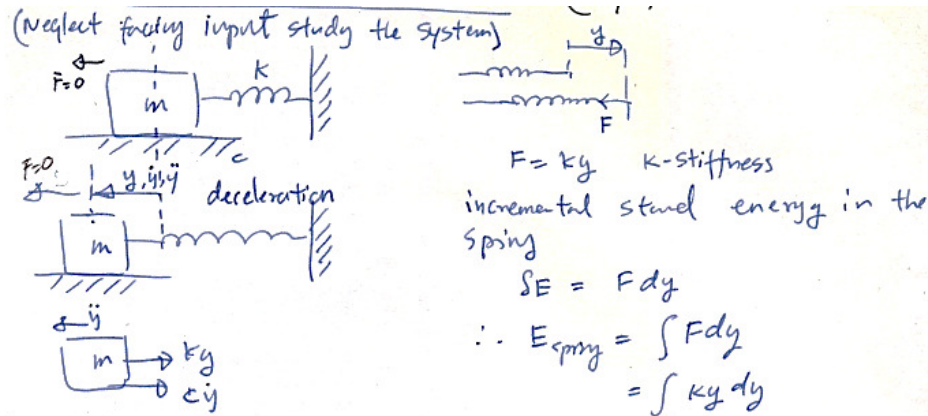
- Consider a hyperspherical region of radius k about an equilibrium state \mathbf{x}_e (using Euclidean norm)
- $\|\mathbf{x} - \mathbf{x}_e\| = k$ where $\|\mathbf{x} - \mathbf{x}_e\| = \sqrt{(x - x_{1e})^2 + (x - x_{2e})^2 + \dots + (x - x_{ne})^2}$
- Define two such spherical regions $\|\mathbf{x} - \mathbf{x}_e\| = \delta$ and $\|\mathbf{x} - \mathbf{x}_e\| = \varepsilon$ with $\delta < \varepsilon$



1. An equilibrium state \mathbf{x}_e is said to be **stable in the sense of Lyapunov** (stability I.S.L.) if trajectories starting within δ do not leave the ε region as t increases indefinitely
2. An equilibrium state \mathbf{x}_e is said to be **asymptotically stable** if trajectories starting within δ converge to \mathbf{x}_e without leaving the ε region as t increases indefinitely. [This case is preferable to stability I.S.L.]
3. An equilibrium state \mathbf{x}_e is said to be **asymptotically stable in the large** if trajectories starting from anywhere in the hyperspace converge to \mathbf{x}_e . There must be only one equilibrium state in the whole state space
4. An equilibrium state \mathbf{x}_e is said to be **unstable** if trajectories starting within δ leaves the ε region as t increases

Lyapunov 1st Method (Indirect, Intuitive)

- For any system $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t)$
- No need of solving state space model
- Non-unique, intuitive method



Class Exercise: Show that the System is stable using energy

From (1) $\Rightarrow m\ddot{y} + ky = -c\dot{y}$ put in (4)

$$\frac{dE}{dt} = -c\dot{y}\dot{y}$$

$$= -c(\dot{y})^2$$

< 0 + $c > 0$ (positive damping)

System is stable for $c > 0$

Assignment: Check stability of the non linear plant

$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t)$ where

$$\dot{x}_1 = -x_1 + x_2 + x_1(x_1^2 + x_2^2) \text{ — (1)}$$

$$\dot{x}_2 = -x_1 - x_2 + x_2(x_1^2 + x_2^2) \text{ — (2)}$$

Force balance: $m\ddot{y} = F - ky - c\dot{y}$ — (1)
 $m\ddot{y} = -(ky + c\dot{y})$ — (1)

$E_{\text{spring}} = \frac{1}{2}ky^2$ — (2)

kinetic Energy of the mass
 $E_k = \frac{1}{2}m\dot{y}^2$ — (3)

Total Energy of the System
 $E_{\text{total}} = E_{\text{spring}} + E_{\text{kinetic}}$ — (2)
 $E = \frac{1}{2}ky^2 + \frac{1}{2}m\dot{y}^2$
 $\frac{dE}{dt} = ky\dot{y} + m\dot{y}\ddot{y}$
 $= (ky + m\ddot{y})\dot{y}$ — (4)

Scalar Functions of State Vector

- Scalar function of state $v(\mathbf{x})$ is +ve definite if $V(\mathbf{x}) > 0 \forall \mathbf{x} \in \Omega$ and $V(\mathbf{x})=0$ for $\mathbf{x}=0$
- Scalar function of state $v(\mathbf{x})$ is -ve definite if $V(\mathbf{x}) < 0 \forall \mathbf{x} \in \Omega$ and $V(\mathbf{x})=0$ for $\mathbf{x}=0$

$V_1(\mathbf{x}) = x_1^2 + x_2^2$ is +ve definite $\begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

$V_2(\mathbf{x}) = (x_1 + x_2)^2$ is +ve semi-definite

$V_3(\mathbf{x}) = x_1^2 + x_1x_2$ is indefinite

$V_4(\mathbf{x}) = -x_1^2 - (x_1 + x_2)^2$ is -ve definite

Lyapunov 2nd Method (Direct)

- If a +ve definite function $V(\mathbf{x})$ can be found such that $dV(\mathbf{x})/dt$ is -ve definite, this equilibrium state is asymptotically stable
- Quadratic form of $V(\mathbf{x}) = \mathbf{x}^T \mathbf{P} \mathbf{x}$; \mathbf{P} is real symmetric and +ve definite

- Sylvester's Criterion – \mathbf{P} is +ve definite if all principal minors are positive. Principal minors are submatrix determinants starting with scalar p_{11} and proceeding (with p_{11} included as the first term in each) until the determinant of the entire \mathbf{P}
- \mathbf{P} is +ve semi-definite if all principal minors are non-negative (at least one zero)

Eg:

$$\mathbf{P} = \begin{array}{|c|cc|} \hline & \text{pm2} & \\ \hline \text{pm1} & 1 & 0 \\ \hline & 1.5 & 0 \\ \hline & 1 & 2 \\ \hline & & \text{pm3} \\ \hline \end{array}$$

In MatLab
 $\mathbf{P} = [1 \ 0 \ 2; 1.5 \ 0.5 \ 2; 1 \ 2 \ 3]$
 $\text{submat1} = \mathbf{P}(1,1)$
 $\text{submat2} = \mathbf{P}(1:2,1:2)$
 $\text{submat3} = \mathbf{P}(1:3,1:3)$
 $\text{pm1} = \det(\text{submat1})$
 $\text{pm2} = \det(\text{submat2})$
 $\text{pm3} = \det(\text{submat3})$

Answer
 $\text{pm1} = 1.0$
 $\text{pm2} = 0.5$
 $\text{pm3} = 2.5$

$$V(\mathbf{x}) = \mathbf{x}^T \mathbf{P} \mathbf{x}$$

$$\dot{V}(\mathbf{x}) = \dot{\mathbf{x}}^T \mathbf{P} \mathbf{x} + \mathbf{x}^T \dot{\mathbf{P}} \mathbf{x} + \mathbf{x}^T \mathbf{P} \dot{\mathbf{x}} = \dot{\mathbf{x}}^T \mathbf{P} \mathbf{x} + \mathbf{x}^T \mathbf{P} \dot{\mathbf{x}}$$

for constant \mathbf{A} and constant \mathbf{P}

$$\dot{V}(\mathbf{x}) = (\mathbf{A}\mathbf{x})^T \mathbf{P} \mathbf{x} + \mathbf{x}^T \mathbf{P} (\mathbf{A}\mathbf{x}); \text{ as } \dot{\mathbf{x}} = \mathbf{A}\mathbf{x}$$

$$\dot{V}(\mathbf{x}) = \mathbf{x}^T \mathbf{A}^T \mathbf{P} \mathbf{x} + \mathbf{x}^T \mathbf{P} \mathbf{A} \mathbf{x} = \mathbf{x}^T (\mathbf{A}^T \mathbf{P} + \mathbf{P} \mathbf{A}) \mathbf{x}$$

If $\dot{V}(\mathbf{x})$ is to be -ve definite $(\mathbf{A}^T \mathbf{P} + \mathbf{P} \mathbf{A})$ has to be -ve definite

Lets introduce any known +ve definite \mathbf{Q} such that

$$\mathbf{A}^T \mathbf{P} + \mathbf{P} \mathbf{A} = -\mathbf{Q}$$

Determine symmetric \mathbf{P} and check for +ve definiteness
 (easy way)

Example (2nd method)

Determine the stability condition

of $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}u$ where $\mathbf{A} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$

The stability condition is

$$\mathbf{A}^T \mathbf{P} + \mathbf{P} \mathbf{A} = -\mathbf{Q} \quad \text{lets set } \mathbf{Q} = \mathbf{I}_2 \text{ +ve def}^{\text{te}}$$

solve for symmetric $\mathbf{P} = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$ and

verify its +ve definiteness

$$\begin{bmatrix} 0 & -2 \\ 1 & -3 \end{bmatrix} \begin{bmatrix} a & b \\ b & c \end{bmatrix} + \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\begin{bmatrix} -2b & -2c \\ a-3b & b-3c \end{bmatrix} + \begin{bmatrix} -2b & a-3b \\ -2c & b-3c \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\begin{bmatrix} -4b & a-3b-2c \\ a-3b-2c & 2b-6c \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \quad \begin{array}{l} -4b = -1 \quad b = 0.25 // \\ 2b - 6c = -1 \\ 2(0.25) - 6c = -1 \\ \frac{2(0.25) + 1}{6} = c = 0.25 // \end{array}$$

due to symmetry of P we have only $\frac{(n+1)n}{2}$ equations

$$\begin{bmatrix} 0 & -4 & 0 \\ 1 & -3 & -2 \\ 0 & 2 & -6 \end{bmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \\ -1 \end{pmatrix} \quad \text{these linear equations are decoupled.}$$

$$\therefore P = \begin{bmatrix} a & b \\ b & c \end{bmatrix} = \begin{bmatrix} 1.25 & 0.25 \\ 0.25 & 0.25 \end{bmatrix}$$

use Sylvester's theorem.

$$p_{m1} = 1.25 > 0$$

$$p_{m2} = \begin{vmatrix} 1.25 & 0.25 \\ 0.25 & 0.25 \end{vmatrix} = 0.25 > 0 \quad \therefore P \text{ is true def}^{+e}$$

\therefore System is asymptotically stable.

This answer can be verified by solving $|sI - A| = 0$, which shows that the system poles are located at -1 and -2 (stable).