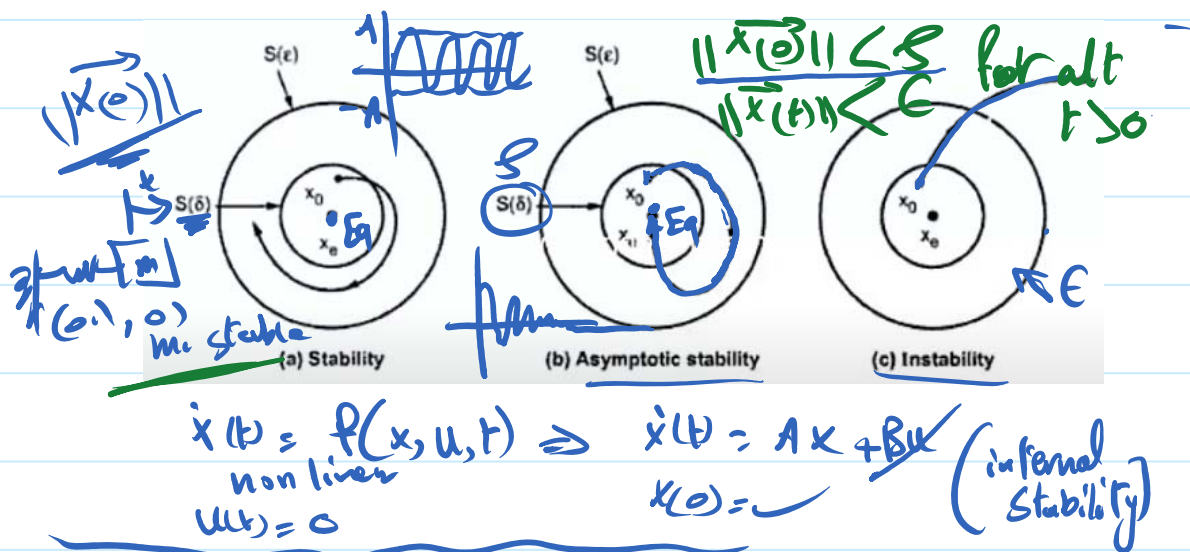


Lyapunov Stability theorem for linear and non-linear systems

- * Assume $\vec{x}(0)$ represent the vector of IC for given Sys. with a stable equilibrium point if any arbitrary number ϵ that exist some the number δ such that whenever:
 - * $\|\vec{x}(0)\| < \delta$ then $\|\vec{x}(t)\| < \epsilon$ for all $t > 0$ "stability Condition"
- if $\lim_{t \rightarrow \infty} \|\vec{x}(t)\| = 0$, then the Sys. is said Asy. Stable.



- * According to Lyapunov theory, one can check the stability of the Sys by finding some scalar Fun of the vector \vec{x} denoted by $V(x)$, which has continuous first partial derivative and satisfy

the following conditions:

① $V(\vec{x}) > 0$ for all values of $\vec{x}(t) \neq \vec{0}$

② $\dot{V}(\vec{x}) = \frac{\partial V}{\partial \vec{x}} \dot{\vec{x}} \leq 0$ for all values of $\vec{x}(t) \neq \vec{0}$

③ $V(\vec{0}) = 0$ $\frac{\partial V}{\partial \vec{x}} \frac{d\vec{x}}{dt} \leq 0$

if you can find such fun satisfied the previous condition in the case the sys. is Asy. stable but if the fun. $V(\vec{x})$ doesn't satisfy the previous condition in this case you can't know if the sys is Asy. stable, unstable, or unstable.

* Study the Stability of Non/linear Sys by LV.

Ex: Study the stability of the following sys.

$\dot{x}_1 = x_2$

$\dot{x}_2 = -x_1 - e^{-t} x_2$

Non linear
* Variant Sys

$\dot{x} = f(x, u, t)$, let $V(x) = x_1^2 + x_2^2$

① $V(x) > 0$ for all values of $\vec{x}(t) \neq \vec{0}$

$V(x) > 0$ (OK)

$\frac{\partial V}{\partial x_1} = 2x_1$

② $\dot{V}(x) = \frac{\partial V}{\partial \vec{x}} \dot{\vec{x}} \leq 0 =$

$\frac{\partial V}{\partial x_2} = 2x_2$

$\dot{V}(x) = 2x_1 \dot{x}_1 + 2x_2 \dot{x}_2$ Sub Eq (*) in (*)

$\dot{V}(x) = 2x_1 x_2 + 2x_2 (-x_1 - e^{-t} x_2) \leq 0??$

$\dot{V}(x) = -2x_2^2 e^{-t} < 0$ satisfied



$$\begin{aligned} x_1 &= 0 \\ x_2 &= 0 \end{aligned}$$

$$\textcircled{3} V(0) = 0 \Rightarrow V(x) = x_1^2 + x_2^2$$

$$V(0) = (0)^2 + (0)^2 \text{ Satisfied}$$

$V(x)$ is Lyapunov Fun, Thus the Sys is Asy. Stable

Ex: Check the stability $\dot{x}_1 = x_2$

$$\dot{x}_2 = -(2 + \cos(x_1)) + \sin(x_1)$$

Let $V(x) = a x_1^2 + b x_2^2$ where $a, b > 0$

$$\textcircled{3} V(0) = 0 \text{ for } x_1 = 0, x_2 = 0 \text{ Satisfied}$$

$$\textcircled{1} V(x) > 0 \text{ for all } \vec{x}(t) \neq \vec{0}$$

Satisfied

$$\textcircled{2} \frac{\partial V}{\partial x} \dot{x} = 2a x_1 \dot{x}_1 + 2b x_2 \dot{x}_2$$

$$\frac{\partial V}{\partial x} \dot{x} \leq 0$$

$$= 2a x_1 x_2 - 2b x_2^2 (2 + \cos x_1) + 2b x_2 \sin x_1$$

$V(x)$ is not Lyapunov Fun. So we can determine anything

Let try $V(x) = a + \cos x_1 + b x_2^2$

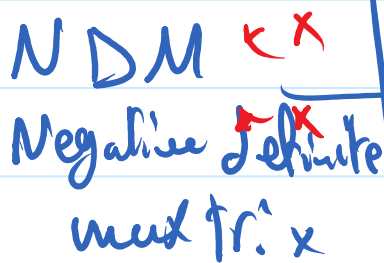
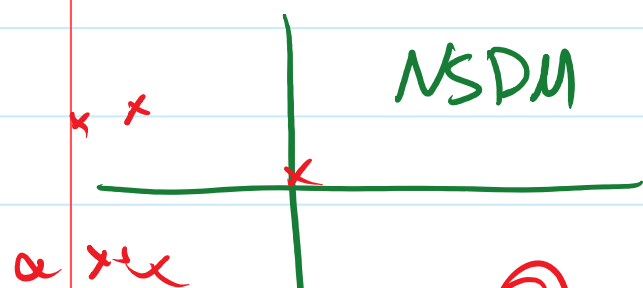
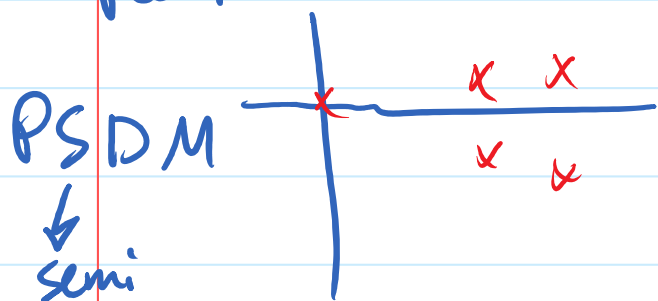
$$\begin{cases} a > 1 \\ b > 0 \end{cases}$$

Lyapunov Stability For LTI

- A Sys is stable in the sense of Lyapunov if it is able to find Lyapunov fun.

* For all LTI Sys $\dot{x}(t) = Ax(t)$, $\vec{x}(0) = \vec{x}_0$
the Lyapunov fun $V(x) = x^T P x > 0$

P symmetric definite matrix $P = P^T > 0$
Positive



αx^2 $\rightarrow Px^2$ (similar)

$V(x) = x^T P x$, Lyapunov fun or not??
(1) $V(0) = 0$ satisfied

(2) $V(x) > 0$ for all $\vec{x}(t) \neq \vec{0}$

$$\dot{V}(x) = \dot{\vec{x}}^T P \vec{x} + \vec{x}^T P \dot{\vec{x}} \quad \text{--- (1)} \quad \vec{\dot{x}} = A\vec{x}$$

$$\begin{aligned} \dot{V}(x) &= (A\vec{x})^T P \vec{x} + \vec{x}^T P (A\vec{x}) \\ &= \vec{x}^T A^T P \vec{x} + \vec{x}^T P A \vec{x} = \vec{x}^T (A^T P + P A) \vec{x} \end{aligned}$$

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

if $\dot{V} =$

$$A^T P + P A = -Q$$

$Q = \begin{bmatrix} \Sigma x \\ \Sigma x \end{bmatrix}$ PDM

< 0 ??

$Q = Q^T > 0$
is any symmetric
positive definite matrix

let suppose and Q matrix

which is PDM and if you find P matrix
is PDM in this case the second
condition is satisfied

Ex: $\dot{x} = \begin{bmatrix} 0 & 1 \\ 2 & 3 \end{bmatrix} x$ check the stability

$|\lambda I - A| =$ \leftarrow Lyapunov.

$V(x) = x^T P x$ $V(0) = 0$ $V(x) > 0$

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

let $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ $(\lambda I - Q) = \lambda_{1,2} = 1$
PDM

$P = \begin{bmatrix} P_{11} & P_{12} \\ P_{12} & P_{22} \end{bmatrix}$

$\begin{bmatrix} 0 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} P_{11} & P_{12} \\ P_{12} & P_{22} \end{bmatrix} + \begin{bmatrix} P_{11} & P_{12} \\ P_{12} & P_{22} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 2 & 3 \end{bmatrix}$

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

if P PDM then the system is stable

