

Lyapunov's indirect method:

- Investigate stability from linearization

- To prove the main result:

① we need to learn how to construct Lyapunov Funct. for stable linear sys.

② we need to control the error in linearization.

Review Lyapunov function for linear sys.

- Consider linear system

$$\dot{x} = AX$$

- Solution is given by

$$X(t) = e^{tA} X(0)$$

- $x=0$ is GAS iff $\operatorname{Re}(\lambda_i) < 0$ for all eigenvalues of A . (Thm 4.5)
- Such a A matrix is called "Hurwitz"
- Stability can be studied using Lyapunov function method.

Example :

$$\begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= -x_1 - 2x_2 \end{aligned} \rightsquigarrow \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = A \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\Rightarrow A = \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix}$$

$$\det(\lambda I - A) = 0 \Rightarrow \det \left(\begin{bmatrix} \lambda & 1 \\ -1 & \lambda + 2 \end{bmatrix} \right) = 0$$

$$\Leftrightarrow \lambda(\lambda + 2) + 1 = 0 \Rightarrow \lambda^2 + 2\lambda + 1 = 0$$

$$\Leftrightarrow (\lambda + 1)^2 = 0 \Rightarrow \lambda_1 = \lambda_2 = -1$$

$\Rightarrow A$ is Hurwitz \Rightarrow GAS

- Now we want to show GAS with Lyapunov func.

- Consider quadratic Lyapunov funct

$$V(x) = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \underbrace{\begin{bmatrix} P_{11} & P_{12} \\ P_{12} & P_{22} \end{bmatrix}}_P \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

V is p.d if P is positive definite matrix

Principal minor condition :

$$P_{11} > 0$$

$$P_{11}P_{22} - P_{12}^2 > 0$$

$$\frac{\partial V}{\partial x}(\alpha) = \left[\frac{\partial V}{\partial x_1}(\alpha), \frac{\partial V}{\partial x_2}(\alpha) \right]$$

$$= [2P_{11}x_1 + 2P_{12}x_2, 2P_{22}x_2 + 2P_{12}x_1]$$

$$= [x_1, x_2] \begin{bmatrix} P_{11} & P_{12} \\ P_{12} & P_{22} \end{bmatrix}$$

In general, for $V(x) = x^T P x$, $\frac{\partial V}{\partial x}(\alpha) = 2x^T P$

$$\dot{V}(\alpha) = \frac{\partial V}{\partial x}(\alpha) f(\alpha) = 2x^T P A x$$

$$= x^T \underbrace{(PA + A^T P)}_{-Q} x$$

In general, for any matrix M

$$x^T M x = \frac{1}{2} x^T \underbrace{(M + M^T)}_{\text{Symmetric part of } M} x$$

why?

$$x^T M x = (x^T M x)^T = x^T M^T x$$

$$\Rightarrow 2x^T M x = x^T M^T x + x^T M x = x^T (M + M^T) x$$

$$\Rightarrow x^T M x = \frac{1}{2} x^T (M + M^T) x$$

The Lyapunov function works if

$$\dot{V}(x) = -x^T Q x < 0 \quad \forall x \neq 0$$

or if Q is positive definite.

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

- we choose a positive definite Q and try to find P that gives the Q , or find P that solves

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

Lyapunov equation.

- let $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$,

- the LHS is

$$\begin{aligned} PA + A^T P &= \begin{bmatrix} -2P_{12}, & P_{11} - 2P_{12} - P_{22} \\ P_{11} - 2P_{12} - P_{22}, & 2P_{12} - 4P_{22} \end{bmatrix} \\ &= -Q = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \end{aligned}$$

$$\Rightarrow P_{12} = \frac{1}{2}, \quad P_{22} = \frac{1}{2}, \quad P_{11} = \frac{3}{2}$$

Principal minor: $P_{11} > 0 \checkmark$ $P_{11}P_{22} - P_{12}^2 = \frac{1}{2} > 0 \checkmark$

$$\Rightarrow P = \begin{bmatrix} \frac{3}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \text{ solves } A^T P + P A = -Q$$

$$\text{for } A = \begin{bmatrix} 0 & +1 \\ -1 & -2 \end{bmatrix}$$

How to do this in general? and $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

Thm: (thm 4.6 in book)

- If A is Hurwitz, then for all p.d. matrix Q , there is a p.d. matrix P that solves the Lyapunov eq.

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

The solution is

$$P = \int_0^{\infty} e^{tA^T} Q e^{tA} dt$$

Why is this P a solution?

$$P = \int_0^{\infty} e^{tA^T} Q e^{tA} dt$$

- Because A is Hurwitz, the integral is bounded and P is well-defined

- P is p.d. because $\forall x \neq 0$

$$x^T P x = \int_0^{\infty} x^T e^{tA^T} Q e^{tA} x dt$$

$$= \int_0^{\infty} \underbrace{(e^{tA} x)^T}_{y^T} Q \underbrace{(e^{tA} x)}_{y} dt$$

where $y \neq 0$

> 0 because Q is p.d.

- To show P solves the Lyapunov eq. consider

$$S(t) = e^{tA^T} Q e^{tA}$$

$$\Rightarrow \dot{S}(t) = A^T S(t) + S(t) A$$

$$\Rightarrow \int_0^{\infty} \dot{S}(t) dt = A^T \underbrace{\int_0^{\infty} S(t) dt}_P + \underbrace{\int_0^{\infty} S(t) dt}_P A$$

$$\Rightarrow S(\infty) - S(0) = A^T P + PA$$

but $S(\infty) = \lim_{t \rightarrow \infty} e^{tA^T} Q e^{tA} = 0$ because A is Hurwitz
and $S(0) = Q$

$$\Rightarrow -Q = A^T P + PA \quad \checkmark$$

Thm: (Thm 4.5 and 4.6 in Khalil)

The following statements are equivalent.

① the system $\dot{x} = Ax$ is AS

② matrix A is Hurwitz

③ \forall p.d. matrix Q , \exists p.d. matrix P that

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

Thm: (thm 4.7 in Khalil)

Let $x=0$ be eqib. point of $\dot{x}=f(x)$
where f is C^1 . Let

$$A = \frac{\partial f}{\partial x}(0)$$

- ① If $\operatorname{Re}(\lambda_i) < 0$ for all eigenvalues of A
 $\Rightarrow x=0$ is AS
- ② If $\operatorname{Re}(\lambda_i) > 0$ for some eigenvalue of A
 $\Rightarrow x=0$ is unstable.
- ③ If $\operatorname{Re}(\lambda_i) \leq 0$ and $\operatorname{Re}(\lambda_i) = 0$ for some λ_i
 \Rightarrow no conclusion

Example:

$$\textcircled{1} \quad \dot{x} = -x - x^3 \xrightarrow{\text{linearize}} \dot{x} = -x \Rightarrow \text{AS}$$

$$\textcircled{2} \quad \dot{x} = -x + x^3 \xrightarrow{\text{"}} \dot{x} = -x \Rightarrow \text{AS}$$

$\textcircled{1}$ is actually GAS, but thm is not strong enough to conclude GAS.

Using the Lyapunov funct.

$$\begin{aligned} \textcircled{1} \quad \underbrace{V(x) = x^2}_{\substack{\text{p.d. and} \\ \text{radially unbounded}}} &\Rightarrow \dot{V}(x) = 2x(-x - x^3) \\ &= -2x^2 - 2x^4 < 0 \quad \forall x \neq 0 \\ &\Rightarrow \text{GAS} \end{aligned}$$

$$\begin{aligned} \textcircled{2} \quad V(x) = x^2 &\Rightarrow \dot{V}(x) = 2x(-x + x^3) \\ &= -2x^2 + 2x^4 \\ &= -2x^2(1 - x^2) < 0 \\ &\quad \forall x \in D - \{0\} \end{aligned}$$

where $\underbrace{D = (-1, 1)}_{\text{open set containing 0}} \Rightarrow \text{AS.}$