

# Nonlinear Systems

## Lecture 12

02/28/13

### Last time

LaSalle's invariance principle

- applies to time-invariant systems

- conditions for asymptotic stability even when  $\dot{V}(x) \not< 0$

$$\mathcal{L}_c = \{x ; V(x) \leq c\} ; \text{ bounded}$$

$$\dot{V}(x) < 0 \text{ in } \mathcal{L}_c$$

$$S = \{x \in \mathcal{L}_c ; \dot{V}(x) = 0\}$$

$M$ : largest invariant set in  $S$        $\Rightarrow x(t) \xrightarrow{t \rightarrow \infty} M$   
 $\forall x(0) \in \mathcal{L}_c$

Consequence  $\Rightarrow$  If  $M = \{0\} \Rightarrow$  asymptotic stability of the origin

Ex 1  $\dot{x}_1 = x_2$

$$\dot{x}_2 = -ag(x_1) - bx_2$$

$a > 0, b \geq 0 ; g(0) = 0 \quad x_1 g(x_1) > 0, \text{ for all } x \in (-b, c)$

$$\dot{V}(x) = -ax_1^2 - bx_2^2 = -[x_1 \ x_2] \begin{bmatrix} 0 & 0 \\ 0 & b \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq 0$$

$$V(x) = a \int_0^{x_1} g(\xi) d\xi + \frac{1}{2} x_2^2$$

$$\Rightarrow S = \{(x_1, x_2), x_2=0\}$$

$$x_2=0 \Rightarrow \dot{x}_2=0 \Rightarrow 0 = -ag(x_1) - b \cdot 0^2$$

$g(x_1)=0 \Rightarrow$  from properties of  $g$   
 $x_1=0$

$\Rightarrow M=\{0\} \rightarrow$  local asymptotic stability

In the linear case:

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = ax_1 - bx_2$$

LaSalle applies (use argument from Ex1) with:

$$V(x) = ax_1^2 + \frac{1}{2}x_2^2$$

In general such Lyapunov functions can be written as:

$$V = x^T P x ; \text{ In this example } P = \begin{bmatrix} a & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$$

Note!  $\dot{V}$  can be written as

$$\dot{V}(x) = -x^T Q x \quad \text{with} \quad Q = \begin{bmatrix} 0 & 0 \\ 0 & b \end{bmatrix}$$

Lyapunov indirect method for Linear time-invariant systems:

$$\dot{x} = Ax$$

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

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

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

$$\dot{V}(x) = -x^T Q x$$

Given  $P = P^T > 0$  show that  $\lim_{\|x\| \rightarrow \infty} x^T P x = +\infty$

$$\lambda_{\min}(P) \|x\|^2 \leq x^T P x \leq \lambda_{\max}(P) \|x\|^2$$

from positive definiteness of  $P = P^T$  (all eigenvalues positive)  
 $x^T P x > 0, \forall x \neq 0$

For linear systems

A: Hurwitz ( $\operatorname{Re}(\lambda_i(A)) < 0, \forall i=1, \dots, n$ )



For any  $Q = Q^T > 0$ , there is  $P = P^T > 0$  st.

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



Note! this  $P$  is unique (for any given  $Q$ )

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



observability gramian

(remember w/A transpose comes  
↓ first)

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

$$V(x) = x^T Q x$$

$$= -x^T C^T C x = -y^T y$$

LaSalle's invariance principle for linear systems:

start with  $V(x) = x^T P x$

$$\dot{V}(x) = -x^T Q x \leq 0 \text{ on } \mathcal{S}_c \quad (\text{not } <)$$

Assume that  $\text{rank}(Q) = r < n$ . Write  $Q = C^T C$   $C \in \mathbb{R}^{r \times n}$   
of rank  $r$

$$\dot{V}(x) = -(Cx)^T (Cx) = 0$$

$$\Rightarrow Cx = 0$$

$$\text{Introduce } y(t) = C \cdot x(t)$$

$$\text{Want, } y = Cx = 0 \Leftrightarrow x = 0$$

$$\begin{cases} \dot{x} = Ax \\ y = Cx \end{cases} \quad \begin{array}{l} y = 0 \Leftrightarrow x = 0 \\ x(t) = e^{At} \cdot x_0 \end{array}$$

Thus asymptotic stability with  $Q = Q^T \geq 0$



$$Q = C^T C$$

Pair  $(A, C)$ : observable.

$$\text{Ex} \quad A = \begin{pmatrix} 0 & 1 \\ -a & -b \end{pmatrix} \quad Q = \begin{pmatrix} 0 & 0 \\ 0 & b \end{pmatrix}$$

$$Q = \underbrace{\begin{bmatrix} 0 \\ \sqrt{b} \end{bmatrix}}_{C^T} \underbrace{\begin{bmatrix} 0 & \sqrt{b} \end{bmatrix}}_C$$

Observability matrix :

$$\begin{bmatrix} C \\ CA \end{bmatrix} = \begin{bmatrix} 0 & \sqrt{b} \\ -a\sqrt{b} & -b\sqrt{b} \end{bmatrix}$$

$$\det \begin{pmatrix} C \\ CA \end{pmatrix} = ab \neq 0$$

$b = 0 \rightarrow$  conservative system & this analysis would not be needed.

$\Rightarrow$  thus  $a \neq 0$  since  $a > 0 \Rightarrow (a > 0 \Leftrightarrow \text{stability})$

Controllability gramian

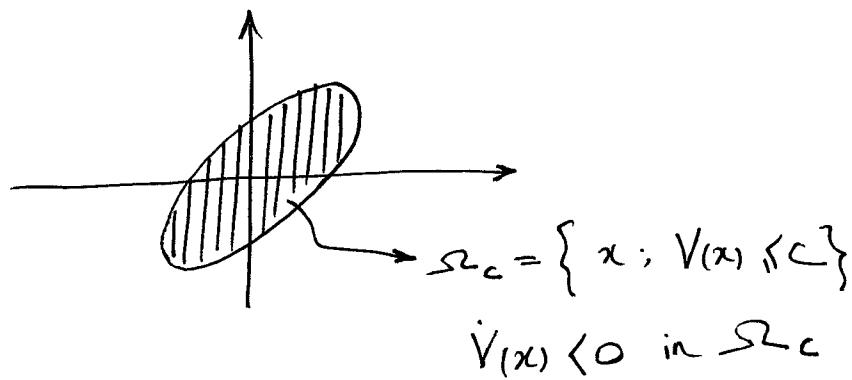
$$AW_c + W_c A^T = -[B] \left[ \begin{array}{c} B^T \\ \downarrow \end{array} \right]$$

for Observability grammian

$$A^T W_o + W_o A = - \begin{bmatrix} C^T \\ \vdots \end{bmatrix} \begin{bmatrix} C \\ \vdots \end{bmatrix}$$

Region of Attraction :

Set of points s.t.  $\left\{ x_0, \underbrace{\phi(t, x_0)}_{x(t) \text{ s.t. } x(0)=x_0} \rightarrow 0 \right\}$



Q: How to estimate the region of attraction?

A possible approach : use Lyapunov based analysis

(Caveat : Maybe conservative!)

For example: use  $V(x) = x^T P x$  where  $P$  is a Lyapunov function for corresponding linearization

Use of sum-of-squares (SOS) techniques can reduce conservatism. (Talk to Gary & Pete)

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Lyapunov direct method (Linearization)

LDM

$$\dot{x} = f(x) = Ax + g(x)$$

$$\frac{\|g(x)\|}{\|x\|} \xrightarrow{\|x\| \rightarrow 0} 0$$

Choose  $V(x) = x^T P x \quad P = P^T > 0$

$$\text{s.t. } A^T P + P A = -Q, \quad Q = Q^T > 0$$

$$\dot{V}(x) = -x^T Q x + g^T(x) P x + x^T P g(x)$$

$$= -x^T Q x + 2x^T P g(x) \leq \underbrace{-x^T Q x}_{\text{negative}} + \underbrace{2\|x\|\|P\|\|g(x)\|}_{\text{positive}}$$

$$\lambda_{\min}(Q) \|x\|^2 \leq x^T Q x \leq \lambda_{\max}(Q) \|x\|^2$$

$$-x^T Q x \leq -\lambda_{\min}(Q) \|x\|^2$$

$$\leq -\lambda_{\min}(Q) \|x\|^2 + 2\|x\|\|P\|\|g(x)\|$$

$$= -\|x\|^2 \left( \lambda_{\min}(Q) - 2\|P\| \frac{\|g(x)\|}{\|x\|} \right)$$

From  $\frac{\|g(x)\|}{\|x\|} \xrightarrow{\|x\| \rightarrow 0} 0$ , for small enough  $\|x\|$

the quantity inside the parentheses on the right will be positive.