

## Lecture 2

- Lyapunov theory *cont'd.*
- Storage function and dissipation
- Absolute stability
- The Kalman-Yakubovich-Popov lemma
- Circle Criterion
- Popov Criterion

# Krasovskii's method

Consider

$$\dot{x} = f(x), \quad f(0) = 0, \quad f(x) \neq 0, \quad \forall x \neq 0$$

and

$$A = \frac{\partial f}{\partial x}$$

$$\text{If } \boxed{A + A^T < 0} \quad \forall x \neq 0$$

then use  $V = f(x)^T f(x) > 0, \forall x \neq 0,$

$$\begin{aligned}\dot{V} &= f^T \dot{f} + \dot{f}^T f \\ &= \left\{ \dot{f} = \frac{\partial f}{\partial x} \dot{x} = Af \right\} \\ &= f^T \left\{ A + A^T \right\} f < 0, \quad \forall x \neq 0\end{aligned}$$

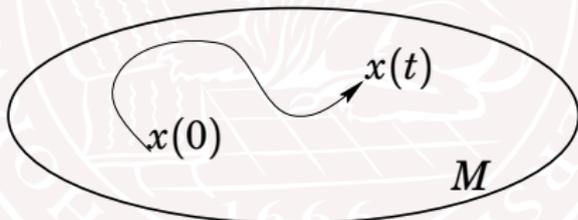
See more general case in [Khalil, Exercise 4.10]

# Invariant Sets

**Definition** A set  $M$  is called **invariant** if for the system

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

$x(0) \in M$  implies that  $x(t) \in M$  for all  $t \geq 0$ .



# Invariant Set Theorem

**Theorem** Let  $\Omega \in \mathbf{R}^n$  be a bounded and closed set that is invariant with respect to

$$\dot{x} = f(x).$$

Let  $V : \mathbf{R}^n \rightarrow \mathbf{R}$  be a radially unbounded  $C^1$  function such that  $\dot{V}(x) \leq 0$  for  $x \in \Omega$ . Let  $E$  be the set of points in  $\Omega$  where  $\dot{V}(x) = 0$ . If  $M$  is the largest invariant set in  $E$ , then every solution with  $x(0) \in \Omega$  approaches  $M$  as  $t \rightarrow \infty$  (see proof in textbook)



Common use is to try to show that the origin is the largest invariant set of  $E$ , ( $M = \{0\}$ ).

## Example – saturated control

*Exercise - 5 min (revisited)*

Find a bounded control signal  $u = \text{sat}(v)$ , which **globally** stabilizes the system

$$\begin{aligned} \dot{x}_1 &= x_1 x_2 \\ \dot{x}_2 &= u \\ u &= \text{sat}(v(x_1, x_2)) \end{aligned} \tag{1}$$

Hint: Use the Lyapunov function candidate

$$V_2 = \ln(1 + x_1^2) + \alpha x_2^2$$

for some appropriate value of  $\alpha$ .

$$V_2 = \ln(1 + x_1^2) + \alpha x_2^2/2$$

$$\dot{V}_2 = 2 \frac{x_1 \dot{x}_1}{1 + x_1^2} + 2\alpha x_2 \dot{x}_2$$

$$= 2x_2 \left( \underbrace{\frac{x_1^2}{1 + x_1^2}}_{0 \leq \dots < 1} + \alpha \text{sat}(v) \right)$$

Can use some part to cancel  $\frac{x_1^2}{1+x_1^2}$  and some to add bounded negative damping in  $x_2$  (like  $\text{sign}(x_2)$  or  $\text{sat}(x_2)$  or ...)

With this type of control law, we end up with

$$\dot{V} = -q(x_2) \leq 0$$

for some  $q(\cdot)$  which only depends on the state  $x_2$ .

$E = \{x | q(x) = 0\}$ , i. e.,  $E$  is the line  $x_2 = 0$ .

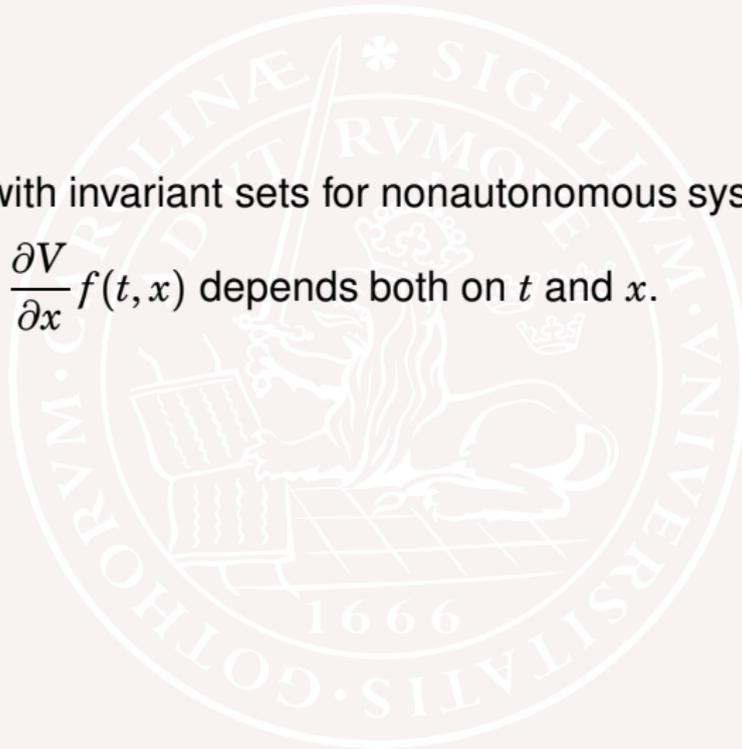
Can solutions stay on that line?

$\dot{x}_2 = 0$  only for also  $x_1 = 0$  (insert control law and check) so the solution curves will not stay on the line  $x_2 = 0$  except for the origin. Thus, the origin is the largest invariant set and asymptotic stability follows from the invariant set theorem.

# Invariant sets - nonautonomous systems

Problems with invariant sets for nonautonomous systems.

$$\dot{V} = \frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x) \text{ depends both on } t \text{ and } x.$$



# Barbalat's Lemma - nonautonomous systems

Let  $\phi : \mathbb{R} \rightarrow \mathbb{R}$  be a uniformly continuous function on  $[0, \infty)$ .  
Suppose that

$$\lim_{t \rightarrow \infty} \int_0^t \phi(\tau) d\tau$$

exists and is finite. Then

$$\phi(t) \rightarrow 0 \text{ as } t \rightarrow \infty$$

Common tool in adaptive control.

- $V(t, x)$  is lower bounded
- $\dot{V}(t, x) \leq 0$
- $\dot{V}(t, x)$  uniformly cont. in time

then  $\dot{V}(t, x) \rightarrow 0$  as  $t \rightarrow \infty$

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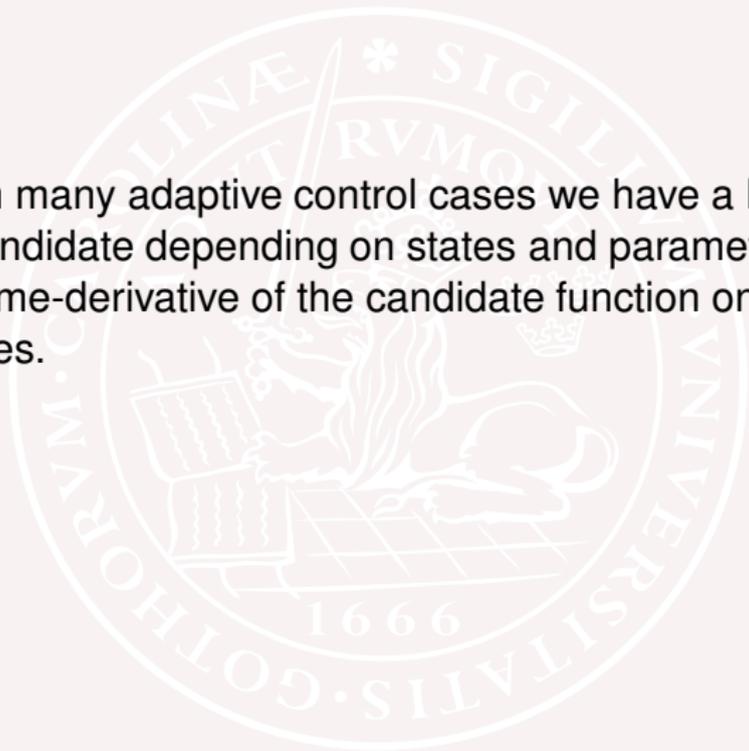
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Remark: In many adaptive control cases we have a Lyapunov function candidate depending on states and parameter errors, while the time-derivative of the candidate function only depends on the states.



# Nonautonomous systems —cont'd

[Khalil, Theorem 4.8 & 4.9]

Assume there exists  $V(t, x)$  such that

$$\underbrace{W_1(x)}_{\text{positive definite}} \leq V(t, x) \leq \underbrace{W_2(x)}_{\text{decrecent}}$$
$$\dot{V}(t, x) = \frac{\partial V}{\partial t} + \frac{\partial V}{\partial x} f(t, x) \leq -W_3(x)$$

$W_3$  is a continuous positive **semi**-definite function.

Solutions to  $\dot{x} = f(t, x)$  starting in  $x(t_0) \in \{x \in B_r | \dots\}$  are bounded and satisfy

$$W_3(x(t)) \rightarrow 0 \quad t \rightarrow \infty$$

See example in Khalil.

# An instability result - Chetaev's Theorem

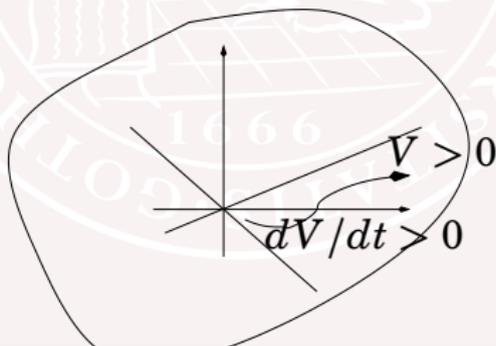
Idea: show that a solution arbitrarily close to the origin have to leave.

Let  $f(0) = 0$  and let  $V : D \rightarrow \mathbf{R}$  be a continuously differentiable function on a neighborhood  $D$  of  $x = 0$ , such that  $V(0) = 0$ .

Suppose that the set

$$U = \{x \in D : \|x\| < r, V(x) > 0\}$$

is nonempty for every  $r > 0$ . If  $\dot{V} > 0$  in  $U$ , then  $x = 0$  is unstable.



Exercise - 5 min [Slotine]

Consider the system

$$\dot{x}_1 = x_1^2 + x_2^3$$

$$\dot{x}_2 = -x_2 + x_1^3$$

Use Chetaev's theorem to show that the origin is an unstable equilibrium point.

You may consider

$$V = x_1 - x_2^2/2$$

for a certain region.

# Dissipativity

Consider a nonlinear system

$$\begin{cases} \dot{x}(t) = f(x(t), u(t), t), & t \geq 0 \\ y(t) = h(x(t), u(t), t) \end{cases}$$

and a locally integrable function

$$r(t) = r(u(t), y(t), t).$$

The system is said to be *dissipative* with respect to the *supply rate*  $r$  if there exists a *storage function*  $S(t, x)$  such that for all  $t_0, t_1$  and inputs  $u$  on  $[t_0, t_1]$

$$S(t_0, x(t_0)) + \int_{t_0}^{t_1} r(t) dt \geq S(t_1, x(t_1)) \geq 0$$

## Example—Capacitor

A capacitor

$$i = C \frac{du}{dt}$$

is dissipative with respect to the supply rate  $r(t) = i(t)u(t)$ .

A storage function is

$$S(u) = \frac{Cu^2}{2}$$

In fact

$$\frac{Cu(t_0)^2}{2} + \int_{t_0}^{t_1} i(t)u(t)dt = \frac{Cu(t_1)^2}{2}$$

# Example—Inductance

An inductance

$$u = L \frac{di}{dt}$$

is dissipative with respect to the supply rate  $r(t) = i(t)u(t)$ .

A storage function is

$$S(i) = \frac{Li^2}{2}$$

In fact

$$\frac{Li(t_0)^2}{2} + \int_{t_0}^{t_1} i(t)u(t)dt = \frac{Li(t_1)^2}{2}$$

# Memoryless Nonlinearity

The memoryless nonlinearity  $w = \phi(v, t)$  with sector condition

$$\alpha \leq \phi(v, t)/v \leq \beta, \quad \forall t \geq 0, v \neq 0$$

is dissipative with respect to the quadratic supply rate

$$r(t) = -[w(t) - \alpha v(t)][w(t) - \beta v(t)]$$

with storage function

$$S(t, x) \equiv 0$$

# Linear System Dissipativity

The linear system

$$\dot{x}(t) = Ax(t) + Bu(t), \quad t \geq 0$$

is dissipative with respect to the supply rate

$$- \begin{bmatrix} x \\ u \end{bmatrix}^T M \begin{bmatrix} x \\ u \end{bmatrix}$$

and storage function  $x^T P x$  if and only if

$$M + \begin{bmatrix} A^T P + PA & PB \\ B^T P & 0 \end{bmatrix} \geq 0$$

# Storage function as Lyapunov function

For a system without input, suppose that

$$r(y) \leq -k|x|^c$$

for some  $k > 0$ . Then the dissipation inequality implies

$$S(t_0, x(t_0)) - \int_{t_0}^{t_1} k|x(t)|^c dt \geq S(t_1, x(t_1))$$

which is an integrated form of the Lyapunov inequality

$$\frac{d}{dt}S(t, x(t)) \leq -k|x|^c$$

# Interconnection of dissipative systems

If the two systems

$$\dot{x}_1 = f_1(x_1, u_1) \quad \dot{x}_2 = f_2(x_2, u_2)$$

are dissipative with supply rates  $r_1(u_1, x_1)$  and  $r_2(u_2, x_2)$  and storage functions  $S(x_1)$ ,  $S(x_2)$ , then their interconnection

$$\begin{cases} \dot{x}_1 = f_1(x_1, h_2(x_2)) \\ \dot{x}_2 = f_2(x_2, h_1(x_1)) \end{cases}$$

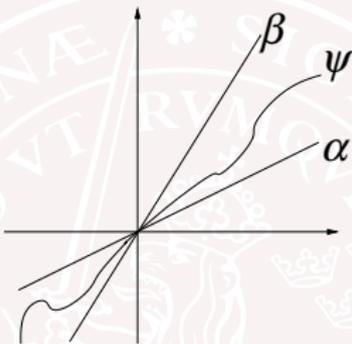
is dissipative with respect to every supply rate of the form

$$\tau_1 r_1(h_2(x_2), x_1) + \tau_2 r_2(h_1(x_1), x_2) \quad \tau_1, \tau_2 \geq 0$$

The corresponding supply rate is

$$\tau_1 S_1(x_1) + \tau_2 S_2(x_2)$$

# Global Sector Condition

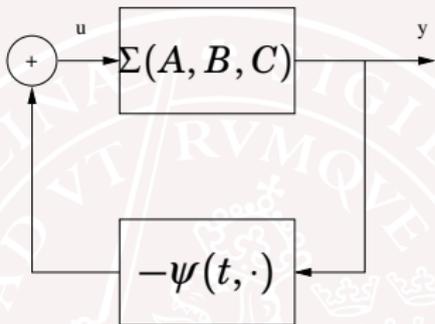


Let  $\psi(t, y) \in \mathbf{R}$  be piecewise continuous in  $t \in [0, \infty)$  and locally Lipschitz in  $y \in \mathbf{R}$ .

Assume that  $\psi$  satisfies the *global sector condition*

$$\alpha \leq \psi(t, y)/y \leq \beta, \quad \forall t \geq 0, y \neq 0 \quad (2)$$

# Absolute Stability

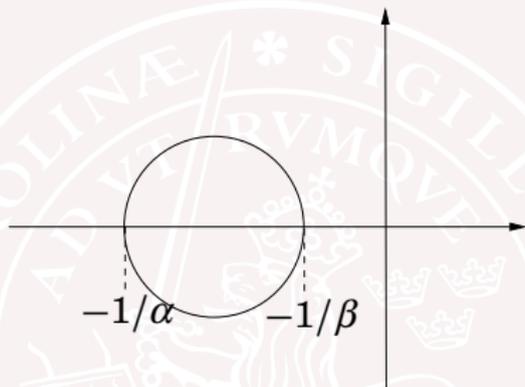


The system

$$\begin{cases} \dot{x} = Ax + Bu, & t \geq 0 \\ y = Cx \\ u = -\psi(t, y) \end{cases} \quad (3)$$

with sector condition (2) is called *absolutely stable* if the origin is globally uniformly asymptotically stable for any nonlinearity  $\psi$  satisfying (2).

# The Circle Criterion

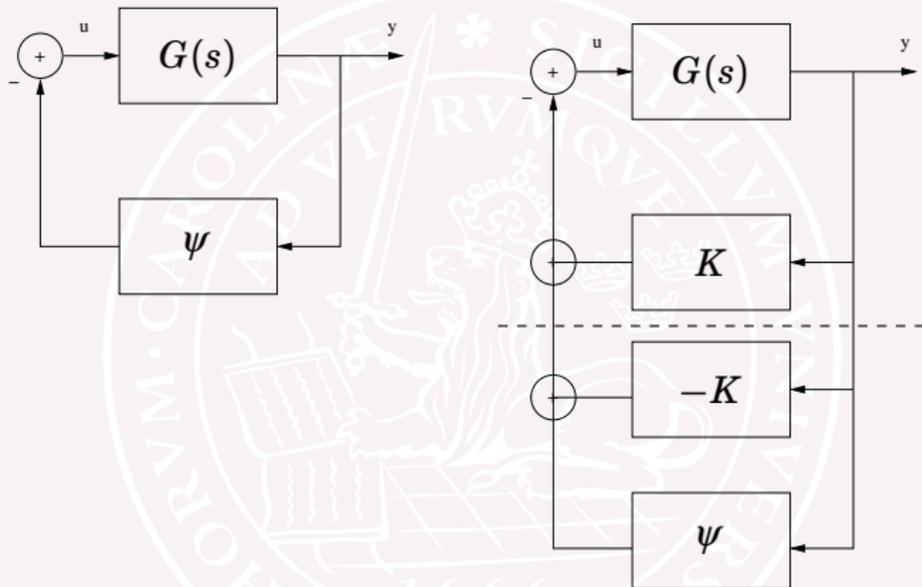


The system (3) with sector condition (2) is absolutely stable if the origin is asymptotically stable for  $\psi(t, y) = \alpha y$  and the Nyquist plot

$$C(j\omega I - A)^{-1}B + D, \quad \omega \in \mathbf{R}$$

does not intersect the closed disc with diameter  $[-1/\alpha, -1/\beta]$ .

# Loop Transformation



Common choices:  $K = \alpha$  or  $K = \frac{\alpha + \beta}{2}$

## Special Case: Positivity

Let  $M(j\omega) = C(j\omega I - A)^{-1}B + D$ , where  $A$  is Hurwitz. The system

$$\begin{cases} \dot{x} = Ax + Bu, & t \geq 0 \\ y = Cx + Du \\ u = -\psi(t, y) \end{cases}$$

with sector condition

$$\psi(t, y)/y \geq 0 \quad \forall t \geq 0, y \neq 0$$

is absolutely stable if

$$M(j\omega) + M(j\omega)^* > 0, \quad \forall \omega \in [0, \infty)$$

Note: For SISO systems this means that the Nyquist curve lies strictly in the right half plane.

# Proof

Set

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

$$\begin{aligned} \Rightarrow \dot{V} &= 2x^T P \dot{x} \\ &= 2x^T P [A \quad B] \begin{bmatrix} x \\ -\psi \end{bmatrix} \leq 2x^T P [A \quad B] \begin{bmatrix} x \\ -\psi \end{bmatrix} + 2\psi y \\ &= 2 \begin{bmatrix} x^T & -\psi \end{bmatrix} \begin{bmatrix} P & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} A & B \\ -C & -D \end{bmatrix} \begin{bmatrix} x \\ -\psi \end{bmatrix} \end{aligned}$$

By the Kalman-Yakubovich-Popov Lemma, the inequality  $M(j\omega) + M(j\omega)^* > 0$  guarantees that  $P$  can be chosen to make the upper bound for  $\dot{V}$  strictly negative for all  $(x, \psi) \neq (0, 0)$ .

Stability by Lyapunov's theorem.

# The Kalman-Yakubovich-Popov Lemma

- Exists in numerous versions
- Idea: **Frequency dependence** is replaced by **matrix equations/inequalities** or vice versa<sup>1</sup>

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# The Kalman-Yakubovich-Popov Lemma

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# The K-Y-P Lemma, version I

Let  $M(j\omega) = C(j\omega I - A)^{-1}B + D$ , where  $A$  is Hurwitz. Then the following statements are equivalent.

- (i)  $M(j\omega) + M(j\omega)^* > 0$  for all  $\omega \in [0, \infty)$
- (ii)  $\exists P = P^T > 0$  such that

$$\begin{bmatrix} P & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} A & B \\ C & D \end{bmatrix} + \begin{bmatrix} A & B \\ C & D \end{bmatrix}^T \begin{bmatrix} P & 0 \\ 0 & I \end{bmatrix} < 0$$

Compare Khalil (5.10-12):

$M$  is strictly positive real if and only if  $\exists P, W, L, \epsilon :$

$$\begin{bmatrix} PA + A^T P & PB - C^T \\ B^T P - C & D + D^T \end{bmatrix} = - \begin{bmatrix} \epsilon P + L^T L & L^T W \\ W^T L & W^T W \end{bmatrix}$$

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Mini-version a la [ Slotine & Li ]:

$$\begin{aligned} \dot{x} &= Ax + bu, & A \text{ Hurwitz, (i. e., } \operatorname{Re}\{\lambda_i(A)\} < 0\}) \\ y &= cx \end{aligned}$$

The following statements are equivalent

- $\operatorname{Re}\{c(j\omega I - A)^{-1}b\} > 0, \forall \omega \in [0, \infty)$
- There exist  $P = P^T > 0$  and  $Q = Q^T > 0$  such that

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

$$Pb = c^T$$

# The K-Y-P Lemma, version II

For

$$\begin{bmatrix} \Phi(s) \\ \tilde{\Phi}(s) \end{bmatrix} = \begin{bmatrix} C \\ \tilde{C} \end{bmatrix} (s\tilde{A} - A)^{-1}(B - s\tilde{B}) + \begin{bmatrix} D \\ \tilde{D} \end{bmatrix},$$

with  $s\tilde{A} - A$  nonsingular for some  $s \in \mathbf{C}$ , the following two statements are equivalent.

## The K-Y-P Lemma, version II - *cont.*

- (i)  $\Phi(j\omega)^* \tilde{\Phi}(j\omega) + \tilde{\Phi}(j\omega)^* \Phi(j\omega) \leq 0$  for all  $\omega \in \mathbf{R}$  with  $\det(j\omega \tilde{A} - A) \neq 0$ .
- (ii) There exists a nonzero pair  $(p, P) \in \mathbf{R} \times \mathbf{R}^{n \times n}$  such that  $p \geq 0$ ,  $P = P^*$  and

$$\begin{aligned} & \begin{bmatrix} A & B \\ C & D \end{bmatrix}^* \begin{bmatrix} P & 0 \\ 0 & pI \end{bmatrix} \begin{bmatrix} \tilde{A} & \tilde{B} \\ \tilde{C} & \tilde{D} \end{bmatrix} \\ & + \begin{bmatrix} \tilde{A} & \tilde{B} \\ \tilde{C} & \tilde{D} \end{bmatrix}^* \begin{bmatrix} P & 0 \\ 0 & pI \end{bmatrix} \begin{bmatrix} A & B \\ C & D \end{bmatrix} \leq 0 \end{aligned}$$

The corresponding equivalence for strict inequalities holds with  $p = 1$ .

# Some Notation Helps

Introduce

$$M = \begin{bmatrix} A & B \end{bmatrix}, \quad \widetilde{M} = \begin{bmatrix} I & 0 \end{bmatrix},$$

$$N = \begin{bmatrix} C & D \end{bmatrix}, \quad \widetilde{N} = \begin{bmatrix} 0 & I \end{bmatrix}.$$

Then

$$y = [C(j\omega I - A)^{-1}B + D]u$$

if and only if

$$\begin{bmatrix} y \\ u \end{bmatrix} = \begin{bmatrix} N \\ \widetilde{N} \end{bmatrix} w$$

for some  $w \in \mathbf{C}^{n+m}$  satisfying  $Mw = j\omega \widetilde{M}w$ .

# Lemma 1

Given  $y, z \in \mathbf{C}^n$ , there exists an  $\omega \in [0, \infty)$  such that  $y = j\omega z$ , if and only if  $yz^* + zy^* = 0$ .

**Proof** Necessity is obvious. For sufficiency, assume that  $yz^* + zy^* = 0$ . Then

$$|v^*(y+z)|^2 - |v^*(y-z)|^2 = 2v^*(yz^* + zy^*)v = 0.$$

Hence  $y = \lambda z$  for some  $\lambda \in \mathbf{C} \cup \{\infty\}$ . The equality  $yz^* + zy^* = 0$  gives that  $\lambda$  is purely imaginary.

# Proof of the K-Y-P Lemma

See handout (Rantzer)



(i) and (ii) can be connected by the following sequence of equivalent statements.

(a)  $w^*(\tilde{N}^*N + N^*\tilde{N})w < 0$  for  $w \neq 0$  satisfying

$$Mw = j\omega\tilde{M}w \text{ with } \omega \in \mathbf{R}.$$

(b)  $\Theta \cap \mathcal{P} = \emptyset$ , where

$$\Theta = \left\{ \left( w^*(\tilde{N}^*N + N^*\tilde{N})w, \tilde{M}ww^*M^* + Mww^*\tilde{M}^* \right) : w^*w = 1 \right\}$$

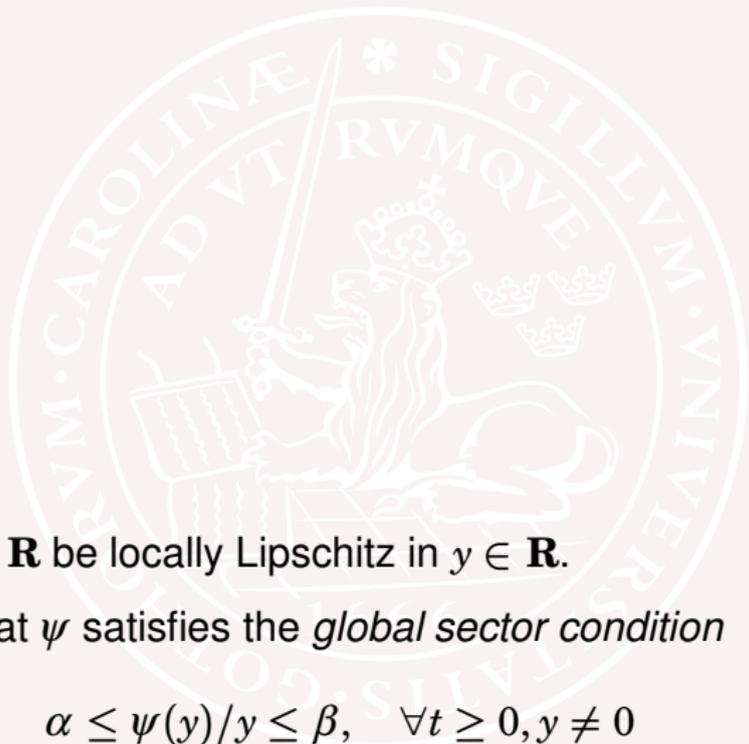
$$\mathcal{P} = \{(r, 0) : r > 0\}$$

(c)  $(\text{conv } \Theta) \cap \mathcal{P} = \emptyset$ .

(d) There exists a hyperplane in  $\mathbf{R} \times \mathbf{R}^{n \times n}$  separating  $\Theta$  from  $\mathcal{P}$ , i.e.  $\exists P$  such that  $\forall w \neq 0$

$$0 > w^* \left( \tilde{N}^*N + N^*\tilde{N} + M^*PM + \tilde{M}^*PM \right) w$$

# Time-invariant Nonlinearity

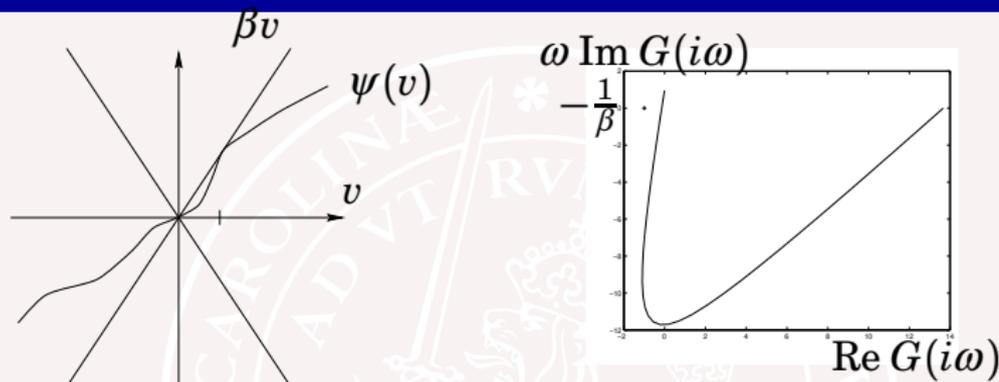


Let  $\psi(y) \in \mathbf{R}$  be locally Lipschitz in  $y \in \mathbf{R}$ .

Assume that  $\psi$  satisfies the *global sector condition*

$$\alpha \leq \psi(y)/y \leq \beta, \quad \forall t \geq 0, y \neq 0$$

# The Popov Criterion



Suppose that  $\psi : \mathbf{R} \rightarrow \mathbf{R}$  is Lipschitz and  $0 \leq \psi(v)/v \leq \beta$ . Let  $G(i\omega) = C(i\omega I - A)^{-1}B$  with  $A$  Hurwitz and  $(A, B, C)$  minimal. If there exists  $\eta \in \mathbf{R}$  such that

$$\text{Re} [(1 + i\omega\eta)G(i\omega)] > -\frac{1}{\beta} \quad \omega \in \mathbf{R} \quad (4)$$

then the differential equation  $\dot{x}(t) = Ax(t) - B\psi(Cx(t))$  is exponentially stable.

# Popov proof I

Set

$$V(x) = x^T P x + 2\eta\beta \int_0^{Cx} \psi(\sigma) d\sigma$$

where  $P$  is an  $n \times n$  positive definite matrix. Then

$$\begin{aligned}\dot{V} &= 2(x^T P + \eta k \psi C) \dot{x} \\ &= 2(x^T P + \eta\beta\psi C) \begin{bmatrix} A & B \end{bmatrix} \begin{bmatrix} x \\ -\psi \end{bmatrix} \\ &\leq 2(x^T P + \eta\beta\psi C) \begin{bmatrix} A & B \end{bmatrix} \begin{bmatrix} x \\ -\psi \end{bmatrix} - 2\psi(\psi - \beta y) \\ &= 2 \begin{bmatrix} x^T & -\psi \end{bmatrix} \begin{bmatrix} PA & PB \\ -\beta C - \eta\beta CA & -1 - \eta\beta CB \end{bmatrix} \begin{bmatrix} x \\ -\psi \end{bmatrix}\end{aligned}$$

By the K-Y-P Lemma there is a  $P$  that makes the upper bound for  $\dot{V}$  strictly negative for all  $(x, \psi) \neq (0, 0)$ .

# Popov proof II

For  $\eta \geq 0$ ,  $V > 0$  is obvious for  $x \neq 0$ .

Stability for linear  $\psi$  gives  $V \rightarrow 0$  and  $\dot{V} < 0$ , so  $V$  must be positive also for  $\eta < 0$ .

Stability for nonlinear  $\psi$  from Lyapunov's theorem.

# The Kalman-Yakubovich-Popov lemma – III

Given  $A \in \mathbf{R}^{n \times n}$ ,  $B \in \mathbf{R}^{n \times m}$ ,  $M = M^T \in \mathbf{R}^{(n+m) \times (n+m)}$ , with  $i\omega I - A$  nonsingular for  $\omega \in \mathbf{R}$  and  $(A, B)$  controllable, the following two statements are equivalent.

(i)

$$\begin{bmatrix} (i\omega I - A)^{-1}B \\ I \end{bmatrix}^* M \begin{bmatrix} (i\omega I - A)^{-1}B \\ I \end{bmatrix} \leq 0 \quad \forall \omega \in \mathbf{R}$$

(ii) There exists a matrix  $P \in \mathbf{R}^{n \times n}$  such that  $P = P^*$  and

$$M + \begin{bmatrix} A^T P + PA & PB \\ B^T P & 0 \end{bmatrix} \leq 0$$

# Proof techniques

(ii)  $\Rightarrow$  (i) simple

Multiply from right and left by  $\begin{bmatrix} (i\omega I - A)^{-1}B \\ I \end{bmatrix}$

(i)  $\Rightarrow$  (ii) difficult

- Spectral factorization (Anderson)
- Linear quadratic optimization (Yakubovich)
- Find  $(1, P)$  as separating hyperplane between the sets

$$\left\{ \left( \begin{bmatrix} x \\ u \end{bmatrix}^T M \begin{bmatrix} x \\ u \end{bmatrix}, x(Ax + Bu)^* + (Ax + Bu)x^* \right) : (x, u) \in \mathbf{C}^{n+m} \right\}$$
$$\{(r, 0) : r \geq 0\}$$

