

# Lecture 4 — Lyapunov Stability

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## Material

- ▶ Glad & Ljung Ch. 12.2
- ▶ Khalil Ch. 4.1-4.3
- ▶ Lecture notes

# Today's Goal

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*To be able to*

- ▶ *prove local and global stability of an equilibrium point using Lyapunov's method*
- ▶ *show stability of a set (e.g., an equilibrium, or a limit cycle) using La Salle's invariant set theorem.*

# Alexandr Mihailovich Lyapunov (1857–1918)

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Master thesis “On the stability of ellipsoidal forms of equilibrium of rotating fluids,” St. Petersburg University, 1884.

Doctoral thesis “The general problem of the stability of motion,” 1892.

# Main idea

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Lyapunov formalized the idea:

*If the total energy is dissipated, then the system must be stable.*

**Main benefit:** By looking at **how** an energy-like function **V** (a so called *Lyapunov function*) **changes over time**, we might **conclude** that a system is stable or asymptotically stable **without solving** the nonlinear differential equation.

**Main question:** **How to find** a Lyapunov function?

# Examples

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Start with a Lyapunov *candidate*  $V$  to measure e.g.,

- ▶ "size"<sup>1</sup> of state and/or output error,
- ▶ "size" of deviation from true parameters,
- ▶ energy difference from desired equilibrium,
- ▶ weighted combination of above
- ▶ ...

Example of common choice in adaptive control

$$V = \frac{1}{2} \left( e^2 + \gamma_a \tilde{a}^2 + \gamma_b \tilde{b}^2 \right)$$

(here weighted sum of output error and parameter errors)

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<sup>1</sup>Often a magnitude measure or (squared) norm like  $|e|_2^2, \dots$

**Analysis:** Check if  $V$  is decreasing with time

- ▶ Continuous time:  $\frac{dV}{dt} < 0$
- ▶ Discrete time:  $V(k+1) - V(k) < 0$

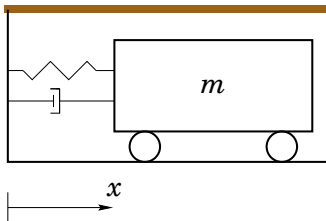
**Synthesis:** Choose, e.g., control law and/or parameter update law to satisfy  $\dot{V} \leq 0$

$$\begin{aligned}\frac{dV}{dt} &= e\dot{e} + \gamma_a \tilde{a}\dot{\tilde{a}} + \gamma_b \tilde{b}\dot{\tilde{b}} = \\ &= \tilde{x}(-a\tilde{x} - \tilde{a}\hat{x} + \tilde{b}u) + \gamma_a \tilde{a}\dot{\tilde{a}} + \gamma_b \tilde{b}\dot{\tilde{b}} = \dots\end{aligned}$$

If  $a$  is constant and  $\tilde{a} = a - \hat{a}$  then  $\dot{\tilde{a}} = -\dot{\hat{a}}$ .

Choose update law  $\frac{d\hat{a}}{dt}$  in a "good way" to influence  $\frac{dV}{dt}$ .  
(more on this later...)

## A Motivating Example



$$m\ddot{x} = - \underbrace{b\dot{x}|\dot{x}|}_{\text{damping}} - \underbrace{k_0x - k_1x^3}_{\text{spring}}$$

$$b, k_0, k_1 > 0$$

Total energy = kinetic + pot. energy:  $V = \frac{mv^2}{2} + \int_0^x F_{\text{spring}} ds \Rightarrow$

$$V(x, \dot{x}) = m\dot{x}^2/2 + k_0x^2/2 + k_1x^4/4 > 0, \quad V(0, 0) = 0$$

$$\begin{aligned} \frac{d}{dt}V(x, \dot{x}) &= m\ddot{x}\dot{x} + k_0x\dot{x} + k_1x^3\dot{x} = \{\text{plug in system dynamics}^2\} \\ &= -b|\dot{x}|^3 < 0, \text{ for } \dot{x} \neq 0 \end{aligned}$$

What does this mean?

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<sup>2</sup>Also referred to evaluate “along system trajectories”.

# Stability Definitions

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An equilibrium point  $x^*$  of  $\dot{x} = f(x)$  (i.e.,  $f(x^*) = 0$ ) is

- ▶ **locally stable**, if for every  $R > 0$  there exists  $r > 0$ , such that

$$\|x(0) - x^*\| < r \quad \Rightarrow \quad \|x(t) - x^*\| < R, \quad t \geq 0$$

- ▶ **locally asymptotically stable**, if locally stable and

$$\|x(0) - x^*\| < r \quad \Rightarrow \quad \lim_{t \rightarrow \infty} x(t) = x^*$$

- ▶ **globally asymptotically stable**, if asymptotically stable for all  $x(0) \in \mathbf{R}^n$ .



# Lyapunov Theorem for Local Stability

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**Theorem** Let  $\dot{x} = f(x)$ ,  $f(x^*) = 0$  where  $x^*$  is in the interior of  $\Omega \subset \mathbf{R}^n$ . Assume that  $V : \Omega \rightarrow \mathbf{R}$  is a  $C^1$  function. If

(1)  $V(x^*) = 0$

(2)  $V(x) > 0$ , for all  $x \in \Omega$ ,  $x \neq x^*$

(3)  $\dot{V}(x) \leq 0$  along all trajectories of the system in  $\Omega$

$\implies x^*$  is locally stable.

Furthermore, if also

(4)  $\dot{V}(x) < 0$  for all  $x \in \Omega$ ,  $x \neq x^*$

$\implies x^*$  is locally asymptotically stable.

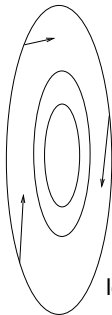
# Lyapunov Functions ( $\approx$ Energy Functions)

A function  $V$  that fulfills (1)–(3) is called a *Lyapunov function*.

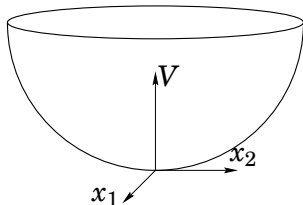
Condition (3) means that  $V$  is non-increasing along all trajectories in  $\Omega$ :

$$\dot{V}(x) = \frac{\partial V}{\partial x} \dot{x} = \sum_i \frac{\partial V}{\partial x_i} f_i(x) \leq 0$$

where  $\frac{\partial V}{\partial x} = \left[ \frac{\partial V}{\partial x_1}, \frac{\partial V}{\partial x_2}, \dots, \frac{\partial V}{\partial x_n} \right]$



level sets where  $V = \text{const.}$



# Conservation and Dissipation

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**Conservation of energy:**  $\dot{V}(x) = \frac{\partial V}{\partial x} f(x) = 0$ , i.e., the vector field  $f(x)$  is everywhere orthogonal to the normal  $\frac{\partial V}{\partial x}$  to the level surface  $V(x) = c$ .

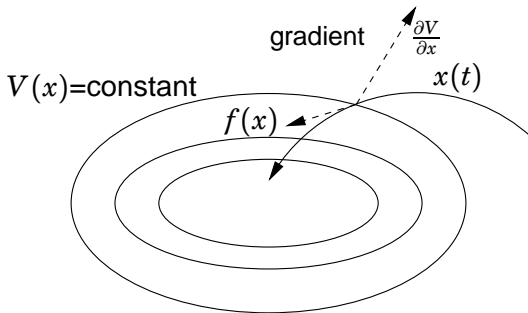
*Example:* Total energy of a lossless mechanical system or total fluid in a closed system.

**Dissipation of energy:**  $\dot{V}(x) = \frac{\partial V}{\partial x} f(x) \leq 0$ , i.e., the vector field  $f(x)$  and the normal  $\frac{\partial V}{\partial x}$  to the level surface  $\{z : V(z) = c\}$  make an obtuse angle (Sw. “trubbig vinkel”).

*Example:* Total energy of a mechanical system with damping or total fluid in a system that leaks.

# Geometric interpretation

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Vector field points into sublevel sets

Trajectories can only go to lower values of  $V(x)$

## Boundedness:

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For any trajectory  $x(t)$

$$V(x(t)) = V(x(0)) + \int_0^t \dot{V}(x(\tau)) d\tau \leq V(x(0))$$

which means that the whole trajectory lies in the set

$$\{z \mid V(z) \leq V(x(0))\}$$

For stability it is thus important that the sublevel sets

$$\{z \mid V(z) \leq c\} \text{ bounded } \forall c \geq 0 \iff V(x) \rightarrow \infty \text{ as } \|x\| \rightarrow \infty.$$

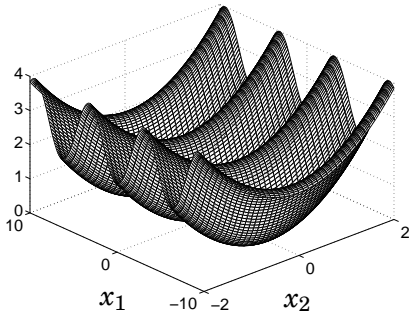
## 2 min exercise—Pendulum

Show that the origin is locally stable for a mathematical pendulum.

$$\dot{x}_1 = x_2, \quad \dot{x}_2 = -\frac{g}{\ell} \sin x_1$$

Use as a Lyapunov function candidate

$$V(x) = (1 - \cos x_1)g\ell + \ell^2 x_2^2/2$$



## Example—Pendulum

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(1)  $V(0) = 0$

(2)  $V(x) > 0$  for  $-2\pi < x_1 < 2\pi$  and  $(x_1, x_2) \neq 0$

(3)

$$\dot{V}(x) = \dot{x}_1 \sin x_1 g \ell + \ell^2 x_2 \dot{x}_2 = 0, \quad \text{for all } x$$

Hence,  $x = 0$  is locally stable.

Note that  $x = 0$  is not asymptotically stable, because  $\dot{V}(x) = 0$  and not  $< 0$  for all  $x \neq 0$ .

# Positive Definite Matrices

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**Definition:** Symmetric matrix  $M = M^T$  is

- ▶ **positive definite** ( $M > 0$ ) if  $x^T M x > 0, \forall x \neq 0$
- ▶ **positive semidefinite** ( $M \geq 0$ ) if  $x^T M x \geq 0, \forall x$

**Lemma:**

- ▶  $M = M^T > 0 \iff \lambda_i(M) > 0, \forall i$
- ▶  $M = M^T \geq 0 \iff \lambda_i(M) \geq 0, \forall i$

$$M = M^T > 0 \quad V(x) := x^T M x$$

$$\Downarrow$$

$$V(0) = 0, \quad V(x) > 0, \forall x \neq 0$$

$V(x)$  candidate Lyapunov function



## More matrix results

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- ▶ for symmetric matrix  $M = M^T$

$$\lambda_{\min}(M)\|x\|^2 \leq x^T M x \leq \lambda_{\max}(M)\|x\|^2, \quad \forall x$$

Proof idea: factorize  $M = U \Lambda U^T$ , unitary  $U$  (i.e.,  $\|Ux\| = \|x\| \ \forall x$ ),  $\Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$

- ▶ for any matrix  $M$

$$\|Mx\| \leq \sqrt{\lambda_{\max}(M^T M)}\|x\|, \quad \forall x$$

## Example- Lyapunov function for linear system

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$$\dot{x} = Ax = \begin{bmatrix} -1 & 4 \\ 0 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad (1)$$

Eigenvalues of  $A : \{-1, -3\} \Rightarrow$  (global) asymptotic stability.

Find a quadratic Lyapunov function for the system (1):

$$V(x) = x^T P x = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} p_{11} & p_{12} \\ p_{12} & p_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \quad P = P^T > 0$$

Take any  $Q = Q^T > 0$ , say  $Q = I_{2 \times 2}$ . Solve  $A^T P + P A = -Q$ .

## Example cont'd

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

$$\begin{bmatrix} -1 & 0 \\ 4 & -3 \end{bmatrix} \begin{bmatrix} p_{11} & p_{12} \\ p_{12} & p_{22} \end{bmatrix} + \begin{bmatrix} p_{11} & p_{12} \\ p_{12} & p_{22} \end{bmatrix} \begin{bmatrix} -1 & 4 \\ 0 & -3 \end{bmatrix} = \begin{bmatrix} -2p_{11} & -4p_{12} + 4p_{11} \\ -4p_{12} + 4p_{11} & 8p_{12} - 6p_{22} \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \quad (2)$$

Solving for  $p_{11}$ ,  $p_{12}$  and  $p_{22}$  gives

$$2p_{11} = -1$$

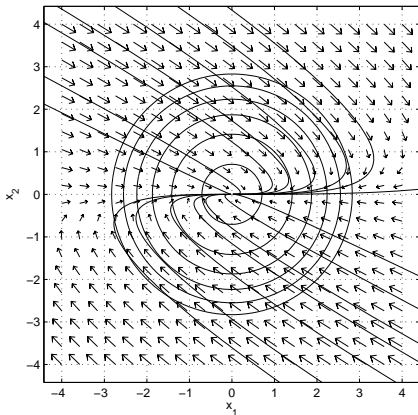
$$-4p_{12} + 4p_{11} = 0$$

$$8p_{12} - 6p_{22} = -1$$

$$\Rightarrow \begin{bmatrix} p_{11} & p_{12} \\ p_{12} & p_{22} \end{bmatrix} = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 5/6 \end{bmatrix} > 0$$

$$\begin{aligned}x_1' &= -x_1 + 4x_2 \\ x_2' &= -3x_2\end{aligned}$$

$$x_1^2 + x_2^2 - 8 = 0$$



Print

Quit

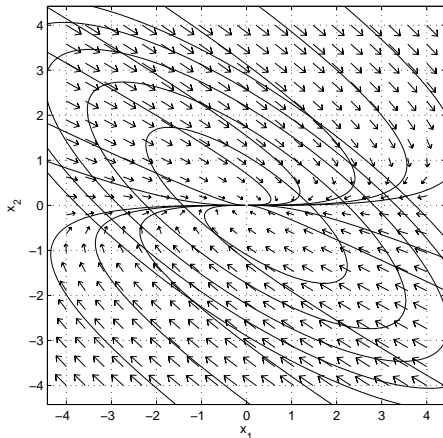
The backward orbit from (0.45, -5.9) left the computation window.  
Ready.  
Preparing to print the PPLANES Display Window. Please be patient.  
Printing the PPLANES Display Window.  
Ready.

Phase plot showing that

$$V = \frac{1}{2}(x_1^2 + x_2^2) = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \text{ does NOT work.}$$

$$\begin{aligned}x_1' &= -x_1 + 4x_2 \\x_2' &= -3x_2\end{aligned}$$

$$(1/2 x_1 + 1/2 x_2) x_1 + (1/2 x_1 + 5/6 x_2) x_2 - 7 = 0$$



Print

Quit

The backward orbit from (3.6,-0.88) left the competition window.  
Ready.  
The forward orbit from (3.1, 3.6) --> a possible eq. pt. near (0.029, 0.3e-08).  
The backward orbit from (3.1, 3.6) left the competition window.  
Ready.

Phase plot with level curves  $x^T P x = c$  for  $P$  found in example.

# Lyapunov Stability for Linear Systems

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**Linear system:**  $\dot{x} = Ax$

*Lyapunov equation:* Let  $Q = Q^T > 0$ . Solve

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

with respect to the symmetric matrix  $P$ .

*Lyapunov function:*  $V(x) = x^T P x, \Rightarrow$

$$\dot{V}(x) = x^T P \dot{x} + \dot{x}^T P x = x^T (PA + A^T P)x = -x^T Q x < 0$$

**Asymptotic Stability:** If  $P = P^T > 0$ , then the Lyapunov Stability Theorem implies (local=global) asymptotic stability, hence the eigenvalues of  $A$  must satisfy  $\operatorname{Re} \lambda_k(A) < 0, \forall k$

# Converse Theorem for Linear Systems

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If  $\operatorname{Re} \lambda_k(A) < 0 \ \forall k$ , then for every  $Q = Q^T > 0$  there exists  $P = P^T > 0$  such that  $PA + A^T P = -Q$

*Proof:* Choose  $P = \int_0^\infty e^{A^T t} Q e^{A t} dt$ . Then

$$\begin{aligned} A^T P + PA &= \lim_{t \rightarrow \infty} \int_0^t \left( A^T e^{A^T \tau} Q e^{A \tau} + e^{A^T \tau} Q A e^{A \tau} \right) d\tau \\ &= \lim_{t \rightarrow \infty} \left[ e^{A^T \tau} Q e^{A \tau} \right]_0^t \\ &= -Q \end{aligned}$$

# Interpretation

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Assume  $\dot{x} = Ax$ ,  $x(0) = z$ . Then

$$\int_0^{\infty} x^T(t) Q x(t) dt = z^T \left( \int_0^{\infty} e^{A^T t} Q e^{At} dt \right) z = z^T P z$$

Thus  $V(z) = z^T P z$  is the cost-to-go from  $z$  (with no input) and integral quadratic cost function with weighting matrix  $Q$ .



# Lyapunov's Linearization Method

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Recall from Lecture 2:

**Theorem** Consider

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

Assume that  $f(0) = 0$ . Linearization

$$\dot{x} = Ax + g(x), \quad \|g(x)\| = o(\|x\|) \text{ as } x \rightarrow 0.$$

(1)  $\operatorname{Re} \lambda_k(A) < 0, \forall k \Rightarrow x = 0$  locally asympt. stable

(2)  $\exists k : \operatorname{Re} \lambda_k(A) > 0 \Rightarrow x = 0$  unstable

# Proof of (1) in Lyapunov's Linearization Method

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Put  $V(x) := x^T P x$ . Then,  $V(0) = 0$ ,  $V(x) > 0 \forall x \neq 0$ , and

$$\begin{aligned}\dot{V}(x) &= x^T P f(x) + f^T(x) P x \\ &= x^T P [Ax + g(x)] + [x^T A^T + g^T(x)] P x \\ &= x^T (PA + A^T P)x + 2x^T P g(x) = -x^T Q x + 2x^T P g(x)\end{aligned}$$

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

and for all  $\gamma > 0$  there exists  $r > 0$  such that

$$\|g(x)\| < \gamma \|x\|, \quad \forall \|x\| < r$$

Thus, choosing  $\gamma$  sufficiently small gives

$$\dot{V}(x) \leq -(\lambda_{\min}(Q) - 2\gamma \lambda_{\max}(P)) \|x\|^2 < 0$$

# Lyapunov Theorem for Global Asymptotic Stability

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**Theorem** Let  $\dot{x} = f(x)$  and  $f(x^*) = 0$ .

If there exists a  $C^1$  function  $V : \mathbf{R}^n \rightarrow \mathbf{R}$  such that

- (1)  $V(x^*) = 0$
- (2)  $V(x) > 0$ , for all  $x \neq x^*$
- (3)  $\dot{V}(x) < 0$  for all  $x \neq x^*$
- (4)  $V(x) \rightarrow \infty$  as  $\|x\| \rightarrow \infty$

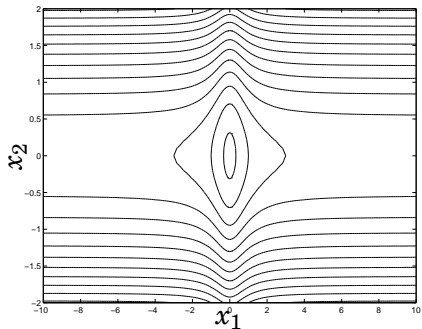
then  $x^*$  is a globally asymptotically stable equilibrium.

# Radial Unboundedness is Necessary

If the condition  $V(x) \rightarrow \infty$  as  $\|x\| \rightarrow \infty$  is not fulfilled, then global stability cannot be guaranteed.

**Example** Assume  $V(x) = x_1^2/(1 + x_1^2) + x_2^2$  is a Lyapunov function for a system. Can have  $\|x\| \rightarrow \infty$  even if  $\dot{V}(x) < 0$ .

Contour plot  $V(x) = C$ :



Example [Khalil]:

$$\dot{x}_1 = \frac{-6x_1}{(1 + x_1^2)^2} + 2x_2$$

$$\dot{x}_2 = \frac{-2(x_1 + x_2)}{(1 + x_1^2)^2}$$

## Somewhat Stronger Assumptions

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**Theorem:** Let  $\dot{x} = f(x)$  and  $f(x^*) = 0$ . If there exists a  $C^1$  function  $V : \mathbb{R}^n \rightarrow \mathbb{R}$  such that

- (1)  $V(x^*) = 0$
- (2)  $V(x) > 0$  for all  $x \neq x^*$
- (3)  $\dot{V}(x) \leq -\alpha V(x)$  for all  $x$
- (4)  $V(x) \rightarrow \infty$  as  $\|x\| \rightarrow \infty$

then  $x^*$  is globally **exponentially** stable.

## Proof Idea

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Assume  $x(t) \neq 0$  ( otherwise we have  $x(\tau) = 0$  for all  $\tau > t$ ).

Then

$$\frac{\dot{V}(x)}{V(x)} \leq -\alpha$$

Integrating from 0 to  $t$  gives

$$\log V(x(t)) - \log V(x(0)) \leq -\alpha t \Rightarrow V(x(t)) \leq e^{-\alpha t} V(x(0))$$

Hence,  $V(x(t)) \rightarrow 0, t \rightarrow \infty$ .

Using the properties of  $V$  it follows that  $x(t) \rightarrow 0, t \rightarrow \infty$ .

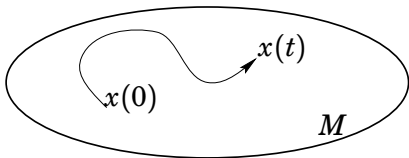
# Invariant Sets

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

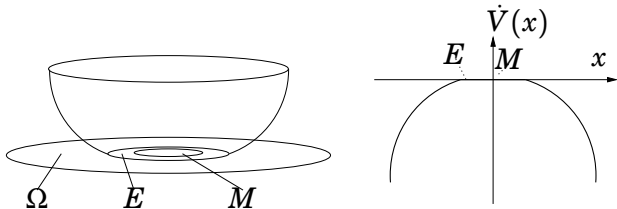
$$\dot{x} = f(x),$$

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



# LaSalle's Invariant Set Theorem

**Theorem** Let  $\Omega \subseteq \mathbf{R}^n$  compact invariant set for  $\dot{x} = f(x)$ .  
Let  $V : \Omega \rightarrow \mathbf{R}$  be a  $C^1$  function such that  $\dot{V}(x) \leq 0, \forall x \in \Omega$ ,  
 $E := \{x \in \Omega : \dot{V}(x) = 0\}$ ,  $M :=$  largest invariant subset of  $E$   
 $\Rightarrow \forall x(0) \in \Omega, x(t)$  approaches  $M$  as  $t \rightarrow +\infty$



Note that  $V$  must **not** be a positive definite function in this case.



# Special Case: Global Stability of Equilibrium

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**Theorem:** Let  $\dot{x} = f(x)$  and  $f(0) = 0$ . If there exists a  $C^1$  function  $V : \mathbb{R}^n \rightarrow \mathbb{R}$  such that

- (1)  $V(0) = 0, V(x) > 0$  for all  $x \neq 0$
- (2)  $\dot{V}(x) \leq 0$  for all  $x$
- (3)  $V(x) \rightarrow \infty$  as  $\|x\| \rightarrow \infty$
- (4) The only solution of  $\dot{x} = f(x), \dot{V}(x) = 0$  is  $x(t) = 0 \forall t$

$\implies x = 0$  is globally asymptotically stable.

## A Motivating Example (cont'd)

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$$m\ddot{x} = -b\dot{x}|\dot{x}| - k_0x - k_1x^3$$

$$V(x) = (2m\dot{x}^2 + 2k_0x^2 + k_1x^4)/4 > 0, \quad V(0,0) = 0$$

$$\dot{V}(x) = -b|\dot{x}|^3$$

Assume that there is a trajectory with  $\dot{x}(t) = 0$ ,  $x(t) \neq 0$ . Then

$$\frac{d}{dt}\dot{x}(t) = -\frac{k_0}{m}x(t) - \frac{k_1}{m}x^3(t) \neq 0,$$

which means that  $\dot{x}(t)$  can not stay constant.

Hence,  $\dot{V}(x) = 0 \iff x(t) \equiv 0$ , and LaSalle's theorem gives global asymptotic stability.

## Example—Stable Limit Cycle

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Show that  $M = \{x : \|x\| = 1\}$  is an asymptotically stable limit cycle for (almost globally, except for starting at  $x=0$ ):

$$\begin{aligned}\dot{x}_1 &= x_1 - x_2 - x_1(x_1^2 + x_2^2) \\ \dot{x}_2 &= x_1 + x_2 - x_2(x_1^2 + x_2^2)\end{aligned}$$

Let  $V(x) = (x_1^2 + x_2^2 - 1)^2$ .

$$\begin{aligned}\frac{dV}{dt} &= 2(x_1^2 + x_2^2 - 1) \frac{d}{dt}(x_1^2 + x_2^2 - 1) \\ &= -2(x_1^2 + x_2^2 - 1)^2(x_1^2 + x_2^2) \leq 0 \quad \text{for } x \in \Omega\end{aligned}$$

$\Omega = \{0 < \|x\| \leq R\}$  is invariant for  $R = 1$ .

## Example—Stable Limit Cycle

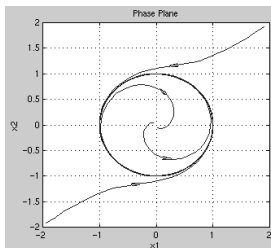
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$$E = \{x \in \Omega : \dot{V}(x) = 0\} = \{x : \|x\| = 1\}$$

$M = E$  is an invariant set, because

$$\frac{d}{dt}V = -2(x_1^2 + x_2^2 - 1)(x_1^2 + x_2^2) = 0 \quad \text{for } x \in M$$

We have shown that  $M$  is a asymptotically stable limit cycle (globally stable in  $R - \{0\}$ )



## A Motivating Example (revisited)

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$$m\ddot{x} = -b\dot{x}|\dot{x}| - k_0x - k_1x^3$$

$$V(x, \dot{x}) = (2m\dot{x}^2 + 2k_0x^2 + k_1x^4)/4 > 0, \quad V(0, 0) = 0$$

$$\dot{V}(x, \dot{x}) = -b|\dot{x}|^3 \text{ gives } E = \{(x, \dot{x}) : \dot{x} = 0\}.$$

Assume there exists  $(\bar{x}, \dot{\bar{x}}) \in M$  such that  $\bar{x}(t_0) \neq 0$ . Then

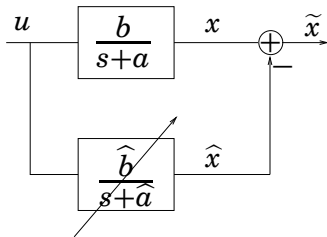
$$m\ddot{\bar{x}}(t_0) = -k_0\bar{x}(t_0) - k_1\bar{x}^3(t_0) \neq 0$$

so  $\dot{\bar{x}}(t_0+) \neq 0$  so the trajectory will immediately leave  $M$ . A contradiction to that  $M$  is invariant.

Hence,  $M = \{(0, 0)\}$  so the origin is asymptotically stable.

# Adaptive Noise Cancellation by Lyapunov Design

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$$\dot{x} + ax = bu$$

$$\dot{\hat{x}} + \hat{a}\hat{x} = \hat{b}u$$

Introduce  $\tilde{x} = x - \hat{x}$ ,  $\tilde{a} = a - \hat{a}$ ,  $\tilde{b} = b - \hat{b}$ .

Want to design adaptation law so that  $\tilde{x} \rightarrow 0$

Let us try the Lyapunov function

$$V = \frac{1}{2}(\tilde{x}^2 + \gamma_a \tilde{a}^2 + \gamma_b \tilde{b}^2)$$

$$\begin{aligned}\dot{V} &= \tilde{x}\dot{\tilde{x}} + \gamma_a \tilde{a}\dot{\tilde{a}} + \gamma_b \tilde{b}\dot{\tilde{b}} = \\ &= \tilde{x}(-a\tilde{x} - \tilde{a}\hat{x} + \tilde{b}u) + \gamma_a \tilde{a}\dot{\tilde{a}} + \gamma_b \tilde{b}\dot{\tilde{b}} = -a\tilde{x}^2\end{aligned}$$

where the last equality follows if we choose

$$\dot{\tilde{a}} = -\dot{\hat{a}} = \frac{1}{\gamma_a} \tilde{x}\hat{x} \quad \dot{\tilde{b}} = -\dot{\hat{b}} = -\frac{1}{\gamma_b} \tilde{x}u$$

Invariant set:  $\tilde{x} = 0$ .

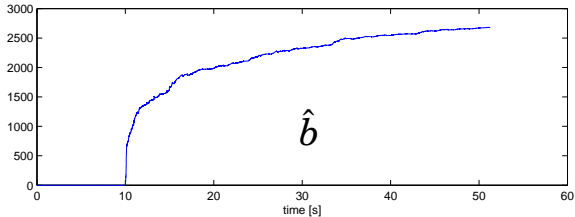
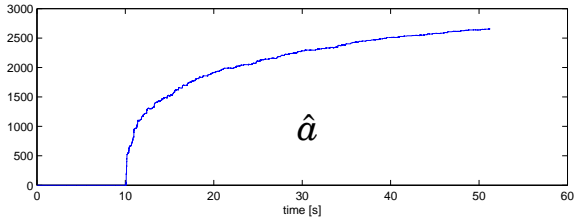
This proves that  $\tilde{x} \rightarrow 0$ .

(The parameters  $\tilde{a}$  and  $\tilde{b}$  do not necessarily converge:  $u \equiv 0$ .)

Demonstration if time permits
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# Results

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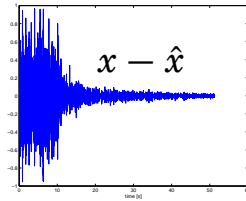


Estimation of parameters starts at  $t=10$  s.



# Results

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Estimation of parameters starts at  $t=10$  s.

## Next Lecture

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- ▶ Stability analysis using input-output (frequency) methods