

Lyapunov stability

ORDINARY DIFFERENTIAL EQUATIONS

An ordinary differential equation is a mathematical model of a continuous state continuous time system:

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

$X = \mathbb{R}^n$ \equiv state space

$f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ \equiv vector field (assigns a "velocity" vector to each x)

Given an initial value $x_0 \in X$,

an execution (solution in the sense of Caratheodory) over the time interval $[0, T)$ is a function $x: [0, T) \rightarrow \mathbb{R}^n$ such that:

- $x(0) = x_0$
- x is continuous and piecewise differentiable
- $x(t) = x(0) + \int_0^t f(x(\tau))d\tau, \forall t \in [0, T)$

ODE SOLUTION: WELL-POSEDNESS

$$\dot{x}(t) = f(x(t)), \quad x(0) = x_0$$

Theorem [global existence and uniqueness]

If $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is **globally Lipschitz continuous**, then $\forall x_0$ there exists a single solution with $x(0)=x_0$ defined on $[0, \infty)$.

Def. $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is **globally Lipschitz continuous**, if there exists a constant L such that

$$\|f(x_1) - f(x_2)\| \leq L \|x_1 - x_2\|, \quad \forall x_1, x_2 \in \mathbb{R}^n$$

STABILITY OF CONTINUOUS SYSTEMS

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

with $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ globally Lipschitz continuous

Definition (equilibrium):

$x_e \in \mathbb{R}^n$ for which $f(x_e)=0$

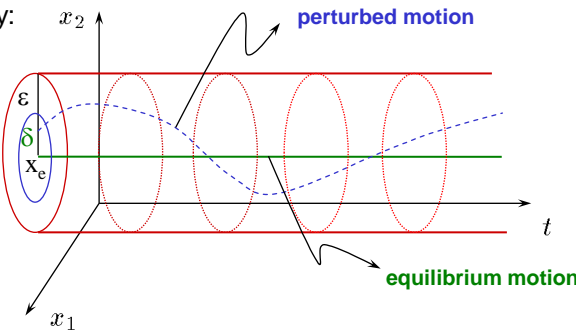
Definition (stable equilibrium):

$$\forall \varepsilon > 0, \exists \delta > 0 : \|x_0 - x_e\| < \delta \Rightarrow \|x(t) - x_e\| < \varepsilon, \forall t \geq 0$$

$$\|v\| = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$$

execution starting from $x(0)=x_0$

Graphically:



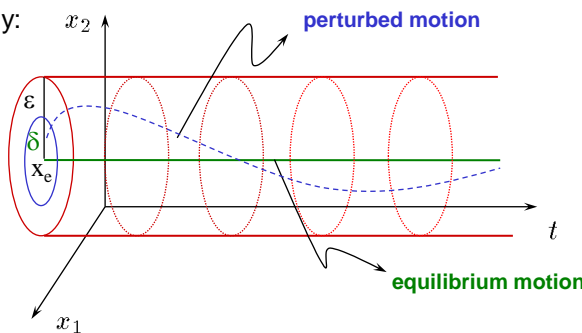
small perturbations lead to small changes in behavior

Definition (asymptotically stable equilibrium):

$$\forall \varepsilon > 0, \exists \delta > 0 : \|x_0 - x_e\| < \delta \rightarrow \|x(t) - x_e\| < \varepsilon, \forall t \geq 0$$

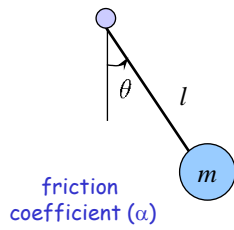
and δ can be chosen so that $\lim_{t \rightarrow \infty} (x(t) - x_e) = 0$

Graphically:



small perturbations lead to small changes in behavior and are re-absorbed, in the long run

EXAMPLE: PENDULUM



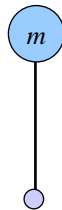
$$x_1 = \theta$$

$$x_2 = \dot{\theta}$$

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

$$\dot{x}_2 = -\frac{g}{l} \sin(x_1) - \frac{\alpha}{m} x_2$$

EXAMPLE: PENDULUM



$$x_1 = \theta$$

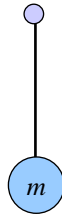
$$x_2 = \dot{\theta}$$

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

$$\dot{x}_2 = -\frac{g}{l} \sin(x_1) - \frac{\alpha}{m} x_2$$

$$x_e = \begin{bmatrix} \pi \\ 0 \end{bmatrix} \text{ unstable equilibrium}$$

EXAMPLE: PENDULUM



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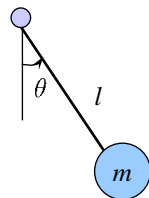
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EXAMPLE: PENDULUM

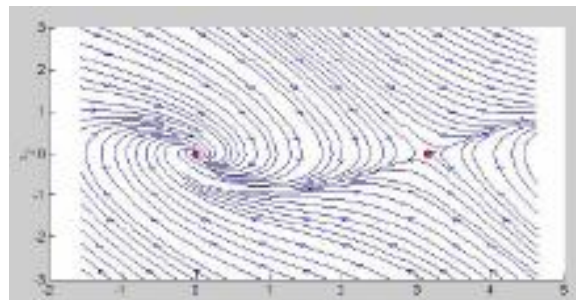


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Let x_e be asymptotically stable.

Definition (domain of attraction):

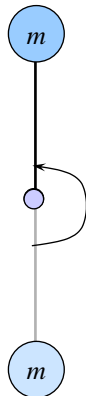
The domain of attraction of x_e is the set of x_0 such that

$$\lim_{t \rightarrow \infty} (x(t) - x_e) = 0$$

execution starting
from $x(0)=x_0$

Definition (globally asymptotically stable equilibrium):

x_e is globally asymptotically stable (GAS) if its domain of attraction is the whole state space \mathbb{R}^n



EXAMPLE: PENDULUM

$$x_1 = \theta$$

$$x_2 = \dot{\theta}$$

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

$$\dot{x}_2 = -\frac{g}{l} \sin(x_1) - \frac{\alpha}{m} x_2$$

$$x_e = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

as. stable equilibrium

small perturbations are
absorbed, not all

perturbations \rightarrow not GAS

Let x_e be asymptotically stable.

Definition (exponential stability):

x_e is exponentially stable if $\exists \alpha, \delta, \beta > 0$ such that

$$\|x_0 - x_e\| < \delta \rightarrow \|x(t) - x_e\| \leq \alpha \|x_0 - x_e\| e^{-\beta t}, \forall t \geq 0$$

STABILITY OF CONTINUOUS SYSTEMS

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

with $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ globally Lipschitz continuous

Definition (equilibrium):

$x_e \in \mathbb{R}^n$ for which $f(x_e) = 0$

Without loss of generality we suppose that

$$x_e = 0$$

if not, then $z := x - x_e \rightarrow dz/dt = g(z)$, $g(z) := f(z+x_e)$ ($g(0) = 0$)

STABILITY OF CONTINUOUS SYSTEMS

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

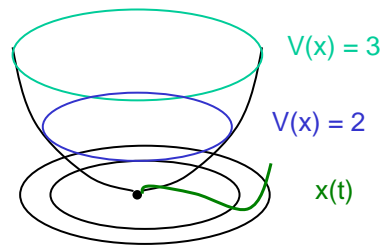
with $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ globally Lipschitz continuous

How to prove stability?

find a function $V: \mathbb{R}^n \rightarrow \mathbb{R}$ such that

$V(0) = 0$ and $V(x) > 0$, for all $x \neq 0$

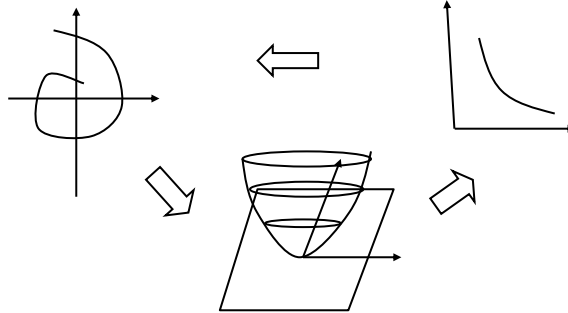
$V(x)$ is decreasing along the executions of the system



STABILITY OF CONTINUOUS SYSTEMS

execution $x(t)$

behavior of V along the execution $x(t)$: $V(t) = V(x(t))$



candidate function $V(x)$

Advantage with respect to exhaustive check of all executions?

STABILITY OF CONTINUOUS SYSTEMS

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

with $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ globally Lipschitz continuous $f(x) = \begin{bmatrix} f_1(x) \\ f_2(x) \\ \vdots \\ f_n(x) \end{bmatrix}$

$V: \mathbb{R}^n \rightarrow \mathbb{R}$ continuously differentiable (C^1) function

Rate of change of V along the execution of the ODE system:

$$\dot{V}(x) = \frac{dV(x(t))}{dt} = \sum_{i=1}^n \frac{\partial V}{\partial x_i} \dot{x}_i = \sum_{i=1}^n \frac{\partial V}{\partial x_i} f_i(x) = \frac{\partial V}{\partial x}(x) \cdot f(x)$$

(Lie derivative of V with respect to f)

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

gradient vector

STABILITY OF CONTINUOUS SYSTEMS

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Rate of change of V along the execution of the ODE system:

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(Lie derivative of V with respect to f)

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

gradient vector

No need to solve the ODE for evaluating if $V(x)$ decreases along the executions of the system

LYAPUNOV STABILITY

Theorem (Lyapunov stability Theorem):

Let $x_e = 0$ be an equilibrium for the system and $D \subset \mathbb{R}^n$ an open set containing $x_e = 0$.

If $V: D \rightarrow \mathbb{R}$ is a C^1 function such that

$$V(0) = 0$$

$$V(x) > 0, \forall x \in D \setminus \{0\}$$

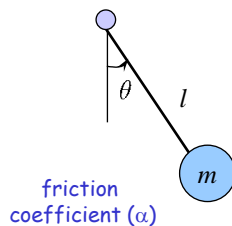
$$\dot{V}(x) \leq 0, \forall x \in D$$

V positive definite on D

V non increasing along system executions in D (negative semidefinite)

Then, x_e is **stable**.

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$$x_1 = \theta$$

$$x_2 = \dot{\theta}$$

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

$$\dot{x}_2 = -\frac{g}{l} \sin(x_1) - \frac{\alpha}{m} x_2$$

$$x_e = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$V(x) := mgl(1 - \cos(x_1)) + \frac{1}{2}mx_2^2 \geq 0 \quad \text{energy function}$$

$$\dot{V}(x) = [mgl \sin(x_1) \quad ml^2x_2] \cdot \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = -\alpha l^2 x_2^2 \leq 0$$

x_e stable

LYAPUNOV STABILITY

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If $V: D \rightarrow \mathbb{R}$ is a C^1 function such that

$$V(0) = 0$$

$$V(x) > 0, \forall x \in D \setminus \{0\}$$

$$\dot{V}(x) \leq 0, \forall x \in D$$

Then, x_e is **stable**.

If it holds also that

$$\dot{V}(x) < 0, \forall x \in D \setminus \{0\}$$

Then, x_e is **asymptotically stable (AS)**

LYAPUNOV GAS THEOREM

Theorem (Barbashin-Krasovski Theorem):

Let $x_e = 0$ be an equilibrium for the system.

If $V: \mathbb{R}^n \rightarrow \mathbb{R}$ is a C^1 function such that

$$V(0) = 0$$

$$V(x) > 0, \forall x \in \mathbb{R}^n \setminus \{0\}$$

$$\dot{V}(x) < 0, \forall x \in \mathbb{R}^n \setminus \{0\}$$

V positive definite on \mathbb{R}^n

V decreasing along system executions in \mathbb{R}^n (negative definite)

$$\lim_{\|x\| \rightarrow \infty} V(x) = \infty$$

V radially unbounded

Then, x_e is globally asymptotically stable (GAS).

STABILITY OF LINEAR CONTINUOUS SYSTEMS

$$\dot{x}(t) = Ax(t)$$

- $x_e = 0$ is an equilibrium for the system

$$x(t) = e^{At}x(0), t \geq 0$$

$$e^{At} \rightarrow 0$$

- the elements of matrix e^{At} are linear combinations of $e^{\lambda(A)t}, te^{\lambda_i(A)t}, \dots, t^k e^{\lambda(A)t}$, where $\lambda(A)$ is an eigenvalue of A

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- $x_e = 0$ is asymptotically stable if and only if A is Hurwitz (all eigenvalues with real part < 0)
- asymptotic stability \equiv GAS

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Alternative characterization...

STABILITY OF LINEAR CONTINUOUS SYSTEMS

$$\dot{x}(t) = Ax(t)$$

Theorem (necessary and sufficient condition):

The equilibrium point $x_e = 0$ is asymptotically stable if and only if for all matrices $Q = Q^T$ positive definite ($Q > 0$) the

$$A^T P + P A = -Q \quad \text{Lyapunov equation}$$

has a unique solution $P = P^T > 0$.

Remarks:

Q positive definite ($Q > 0$) iff $x^T Q x > 0$ for all $x \neq 0$

Q positive semidefinite ($Q \geq 0$) iff $x^T Q x \geq 0$ for all x and $x^T Q x = 0$ for some $x \neq 0$

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has a unique solution $P = P^T > 0$.

Proof.

(if) $V(x) = x^T P x$ is a Lyapunov function

$$\begin{aligned} \dot{V}(x) &= \dot{x}^T P x + x^T P \dot{x} \\ &= x^T (A^T P + PA)x = -x^T Q x < 0, \forall x \neq 0 \end{aligned}$$

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Proof.

(only if) Consider $P = \int_0^{\infty} e^{A^T t} Q e^{At} dt$

$$\begin{aligned} A^T P + PA &= \int_0^{\infty} A^T e^{A^T t} Q e^{At} dt + \int_0^{\infty} e^{A^T t} Q e^{At} A dt \\ &= \int_0^{\infty} \frac{d}{dt} e^{A^T t} Q e^{At} dt = -Q \end{aligned}$$

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Proof.

(only if) Consider $P = \int_0^{\infty} e^{A^T t} Q e^{At} dt$

$P = P^T$ and $P > 0$ easy to show

P unique can be proven by contradiction

STABILITY OF LINEAR CONTINUOUS SYSTEMS

$$\dot{x}(t) = Ax(t)$$

Remarks: for a linear system

- existence of a (quadratic) Lyapunov function $V(x) = x^T P x$ is a **necessary and sufficient condition for asymptotic stability**
- it is **easy to compute a Lyapunov function** since the Lyapunov equation

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

is a linear algebraic equation

STABILITY OF LINEAR CONTINUOUS SYSTEMS

$$\dot{x}(t) = Ax(t)$$

Theorem (exponential stability):

Let the equilibrium point $x_e = 0$ be asymptotically stable. Then, the rate of convergence to $x_e = 0$ is exponential:

$$\|x(t)\| \leq \mu e^{-\lambda_0 t} \|x_0\|, t \geq 0$$

for all $x(0) = x_0 \in \mathbb{R}^n$, where $\lambda_0 \in (0, \min_i |\operatorname{Re}\{\lambda_i(A)\}|)$ and $\mu > 0$ is an appropriate constant.

STABILITY OF LINEAR CONTINUOUS SYSTEMS

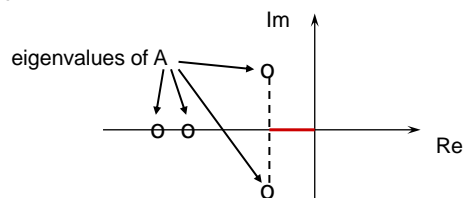
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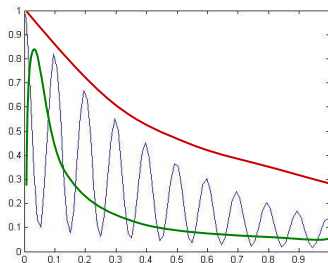
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Remark: $\|x(t)\| = \|e^{At} x_0\| \leq \mu e^{-\lambda_0 t} \|x_0\|, t \geq 0, \forall x_0$
 $\rightarrow \|e^{At}\| = \sup_{x_0 \neq 0} \frac{\|e^{At} x_0\|}{\|x_0\|} \leq \mu e^{-\lambda_0 t}, t \geq 0$

STABILITY OF LINEAR CONTINUOUS SYSTEMS

$$\dot{x}(t) = Ax(t)$$

Proof (exponential stability):

$A + \lambda_0 I$ is Hurwitz (eigenvalues are equal to $\lambda(A) + \lambda_0$)

Then, there exists $P = P^T > 0$ such that

$$(A + \lambda_0 I)^T P + P (A + \lambda_0 I) < 0$$

which leads to

$$x(t)^T [A^T P + P A] x(t) < -2 \lambda_0 x(t)^T P x(t)$$

Define $V(x) = x^T P x$, then

$$\dot{V}(x(t)) < -2 \lambda_0 V(x(t))$$

from which

$$V(x(t)) < e^{-2 \lambda_0 t} V(x_0)$$

STABILITY OF LINEAR CONTINUOUS SYSTEMS

$$\dot{x}(t) = Ax(t)$$

(cont'd) Proof (exponential stability):

$$x^T \lambda_{\min}(P) I x \leq V(x) = x^T P x \leq x^T \lambda_{\max}(P) I x$$

$$\lambda_{\min}(P) \|x(t)\|^2 \leq V(x(t)) < e^{-2 \lambda_0 t} V(x_0) \leq e^{-2 \lambda_0 t} \lambda_{\max}(P) \|x_0\|^2$$

thus finally leading to

$$\|x(t)\| \leq \sqrt{\frac{\lambda_{\max}(P)}{\lambda_{\min}(P)}} e^{-\lambda_0 t} \|x_0\|$$

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$$\dot{x}(t) = Ax(t)$$

- $x_e = 0$ is an equilibrium for the system
- $x_e = 0$ is asymptotically stable if and only if A is Hurwitz (all eigenvalues with real part < 0)
- asymptotic stability \equiv GAS \equiv exponential stability \equiv GES