

Department of Electrical Engineering (ISY)



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Nonlinear Control Systems

PhD course, Spring 2017

Lecturer and examiner:

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Aim:

The course aims at giving an overview of the main control problems and of some of the mathematical tools required in the analysis and synthesis of nonlinear control systems.

Organization

12 lectures (2h each, once a week).

Schedule: Tuesdays at 13:15 in Algoritmen (B house, A-corridor). First class: March 7th.

Prerequisites:

Knowledge in Linear Algebra, Automatic Control, Linear System Theory (basic or advanced) is assumed.

Exam:

Hand-in exercises during the course, plus final take home exam. Credits: 6+3 p

Topics (tentative):

- ❑ Lyapunov stability
 - ❑ stability by linearization
 - ❑ Lyapunov theorems
 - ❑ La Salle invariance principle;
- ❑ Nonlinear controllability
 - ❑ Differential geometric tools: manifolds, vector fields, Lie brackets, Frobenius Theorem;
 - ❑ controllability and Chow Theorem;
 - ❑ drift versus driftless systems, accessibility versus controllability;
- ❑ Geometric methods for control synthesis
 - ❑ system inversion and differential flatness
 - ❑ feedback linearization;
- ❑ Feedback stabilization
 - ❑ Brockett necessary condition
 - ❑ control Lyapunov functions

Course material:

There is no single book covering all material that will be treated in the course. Some parts can be found in the following:

- ❑ H. Khalil. *Nonlinear Systems*, 3rd ed. Prentice Hall. 2002. (stability and stabilization)
- ❑ Murray-Li-Sastry. *A mathematical Introduction to Robotic Manipulation* CRC press 1994. (Controllability; nonholonomic systems; Lie algebras)
- ❑ F. Ticozzi. *Nonlinear Systems and Control*, Lecture Notes, 2016.

Page responsible: **Claudio Altafini**

Last updated: 2017-03-19

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Linear vs. Nonlinear

	L	N
frequency analysis	yes	no
modal analysis	yes	no
superposition principle	yes	no
multiple isolated equil	no	yes
finite escape time	no	yes
nonuniqueness of solution	no	yes
limit cycle (isolated periodic orbits)	no	yes
bifurcations	no	yes
chaos	no	yes

Multiple isolated equilibria

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In a linear system $\dot{x} = Ax$

- $\bar{x} = 0$ is always an equil point : $\dot{x} = 0 = A\bar{x}$
- if $\bar{x} \neq 0$ is an equil point, then $\text{span}\{\bar{x}\}$ is a subspace of equilibria : $A\bar{x} = 0 \Leftrightarrow A\bar{k}\bar{x} = 0$
 $\rightarrow \bar{x} \in \ker(A) \Rightarrow \det(A) = 0$ $k \in \mathbb{R}$
- if $\det(A) \neq 0$ then $\bar{x} = 0$ is the only equil. point
- if $\det(A) = 0$ then \exists subspace of equil points
- cannot have multiple isolated equil points

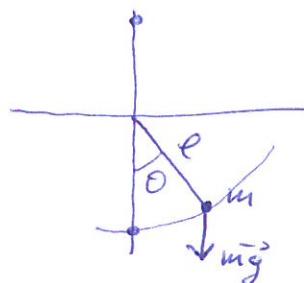
In a nonlin. system situation is different

example pendulum eq.

Newton law:

$$ml\ddot{\theta} + mg \sin \theta + ke\dot{\theta} = 0$$

friction term



state space form: $x_1 = \theta$, $x_2 = \dot{\theta}$

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -\frac{g}{l} \sin x_1 - \frac{k}{m} x_2 \end{cases}$$

nonlinear

equil: $x_2 = 0$

$$\sin x_1 = 0$$

$$\bar{x} = (0, 0)$$

$$\bar{x} = (\pi, 0)$$

isolated eq.
points

n3

Finite escape time

For the linear system $\dot{x} = Ax$, if A is stable (i.e. all eig of A have $\text{Re}[\lambda_i] \leq 0$) then the trajectories are bounded. If $\exists \lambda_i$ s.t. $\text{Re}[\lambda_i] > 0$ then $\exists x_0$ s.t. the solution diverges

$$\lim_{t \rightarrow \infty} \|x(t)\| = \lim_{t \rightarrow \infty} \|e^{At}x(0)\| = +\infty$$

for each finite t , however, the solution is finite i.e. $\|x(t)\| < +\infty \quad \forall t \text{ finite}$

For nonlinear systems, the solution may become ∞ even when t is finite

example : $\dot{x} = -x + x^2$

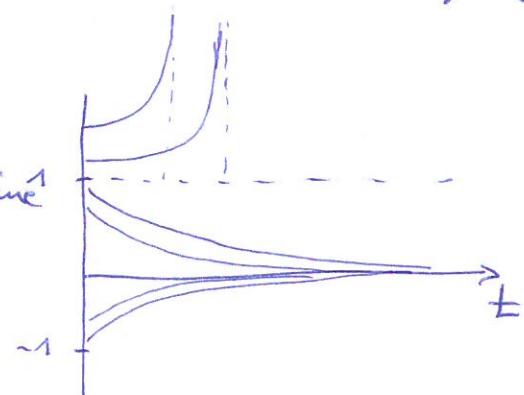
$x=0$ equil point, if we look at the linearized around 0, $\dot{\tilde{x}} = -\tilde{x}$, then the solution near 0 is $\tilde{x}(t) = \tilde{e}^{-t}\tilde{x}(0)$, stable and converging to $0 = \bar{x}$.

But when $x_0 > 1$ the situation is different: divergence in finite time

Integrating the ODE

$$x(t) = \frac{\tilde{e}^t x_0}{1 - x_0 + x_0 \tilde{e}^t}$$

when $x_0 > 1$: denominator at $t=0$ is $1 - x_0 + x_0 = 1$ then it decreases until 0 when $x_0 \tilde{e}^t = x_0 - 1$ \Rightarrow solution diverges to $+\infty$

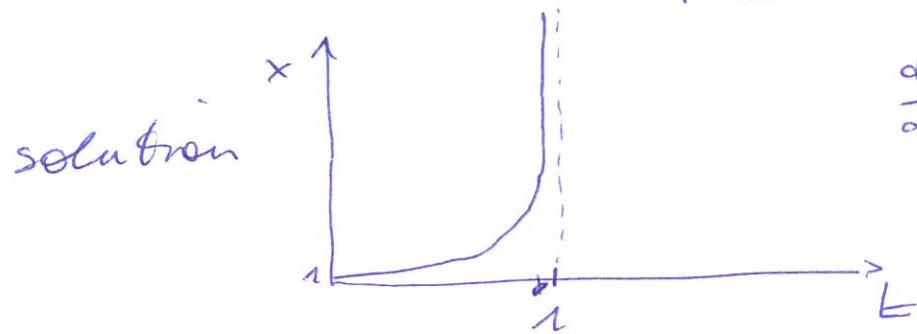


example (simpler)

$$\begin{cases} \dot{x} = x^2 \\ x(0) = 1 \end{cases}$$

the solution is $x(t) = \frac{1}{1-t}$

(verify by differentiation)



$$\frac{dx}{dt} = -\frac{(-1)}{(1-t)^2} = \frac{1}{(1-t)^2} = x^2$$

$x(t) \rightarrow \infty$ as $t \rightarrow 1^-$ \Rightarrow finite escape time

\Rightarrow solution exists only in $0 \leq t < 1$

more general solution

$$\begin{cases} \dot{x} = x^2 \\ x(0) = x_0 \end{cases} \Rightarrow x(t) = \frac{x_0}{1-x_0 t}$$

$$\frac{dx}{x^2} = dt \Rightarrow -\frac{1}{x(t)} + \frac{1}{x_0} = t$$

- if $x_0 < 0 \Rightarrow x(t)$ exists in $[0, \infty)$

- if $x_0 > 0 \Rightarrow x(t)$ exists in $[0, \frac{1}{x_0})$

I can compute the maximal solution only because I could integrate explicitly the ODE. In general I cannot do it!

Nonuniqueness of solutions

- linear systems: $\ddot{x} = Ax$ has always a unique sol.
- nonlinear system need not have it!

example $\dot{x} = 2\sqrt{x}$

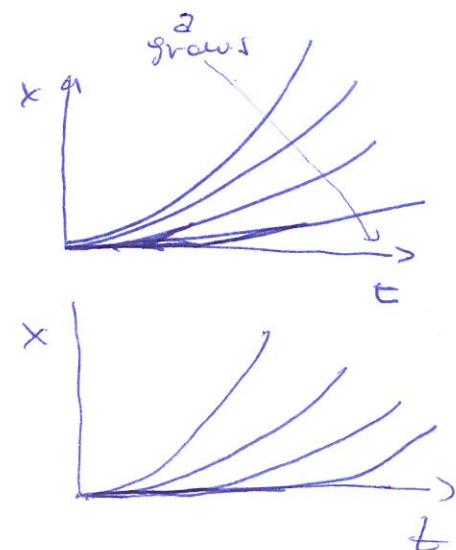
$$x(0) = 0$$

For each $a > 0$ the following is a solution

$$x(t) = \begin{cases} 0 & \text{if } 0 \leq t \leq a \\ (t-a)^2 & \text{if } t > a \end{cases}$$

In fact, differentiating:

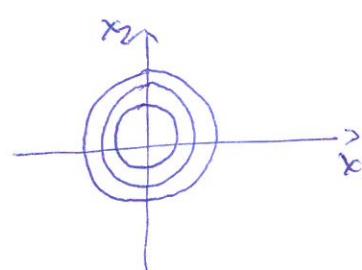
$$\begin{aligned} \frac{dx}{dt} &= \begin{cases} 0 & 0 \leq t \leq a \\ 2(t-a) & t > a \end{cases} \\ &= 2\sqrt{x} \end{aligned}$$



(stable) limit cycle

Linear system $\dot{x} = Ax$ can have cyclical trajectories, but they can only be stable, never attractors of nearby trajectories (they are not isolated)

$$A = \begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix}$$



phase portrait

$$\det(sI - A) = s^2 + \omega^2 = 0 \\ \Rightarrow \lambda_{1,2} = \pm i\omega$$

$$x(t) = e^{At} x(0) = \begin{bmatrix} \cos \omega t & \sin \omega t \\ -\sin \omega t & \cos \omega t \end{bmatrix} x(0)$$

- Nonlinear systems can have isolated periodic trajectories that attract nearby trajectories

example basic oscillator circuit

$$i = h(v)$$

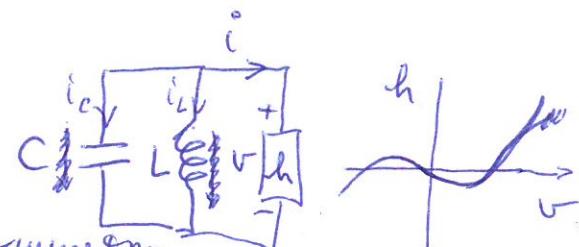
nonlinear

$$L, C > 0$$

resistance of odd symmetry

$$h(0) = 0$$

$$\frac{dh}{dv}\Big|_0 = h'(0) < 0$$



Kirchoff law

$$i_C + i_L + i = 0$$

$$C \frac{dV_C}{dt} = i_C$$

$$L \frac{di_L}{dt} = V_L$$

$$C \frac{dV_C}{dt} + \frac{1}{L} \int_{-\infty}^t v(s) ds + h(v) = 0$$

(h7)

differentiating w.r.t. time and multiply by L

$$L C \frac{d^2 v}{dt^2} + v + L h'(v) \frac{dv}{dt} = 0$$

special case of Lienard's equations

- rescale time : $\tau = \frac{t}{\sqrt{CL}} \Rightarrow \frac{dv}{d\tau} = \sqrt{CL} \frac{dv}{dt}$

$$\frac{d^2 v}{d\tau^2} = CL \frac{d^2 v}{dt^2}$$

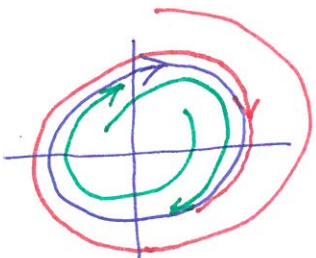
$$\Rightarrow \frac{d^2 v}{d\tau^2} + \underbrace{\sqrt{\frac{L}{C}}}_{=\epsilon} h'(v) \frac{dv}{d\tau} + v = 0$$

Special case: Van der Pol oscillator: $h(v) = -v + \frac{1}{3} v^3$

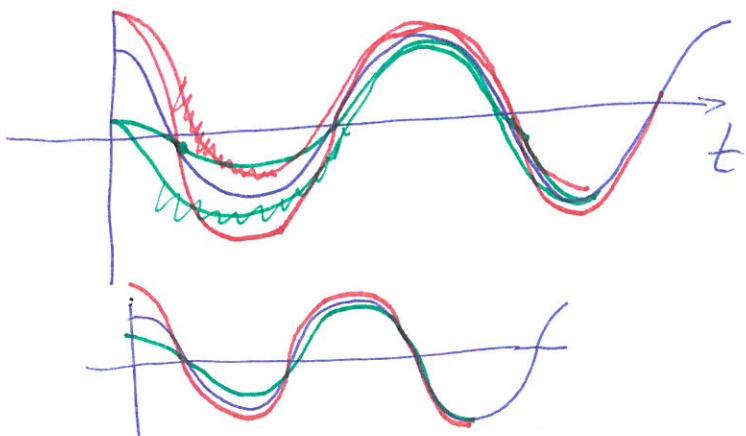
$$\ddot{v} + \epsilon(1-v^2)\dot{v} + v = 0 \quad \epsilon = \sqrt{\frac{L}{C}}$$

state space model : $x_1 = v, \quad x_2 = \dot{v}$

$$\begin{cases} \overset{\circ}{x}_1 = x_2 \\ \overset{\circ}{x}_2 = -x_1 + \epsilon h'(x)x_2 \end{cases}$$



stable limit cycle



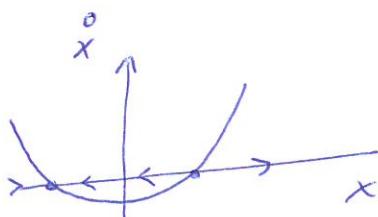
Bifurcation

= qualitative change in solution of ODE as we change a parameter

- change in the number of equil. points
- change in the stability characteristics of the equil. points

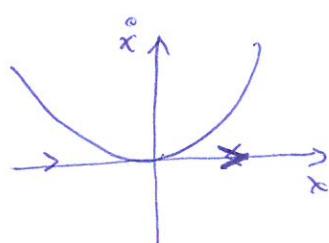
example: saddle-node bifurcation ("blue-sky" bifurc.)

$$\ddot{x} = r x + x^2$$



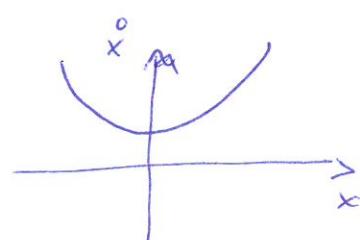
$$r < 0$$

2 equil
1 stable
1 unstable



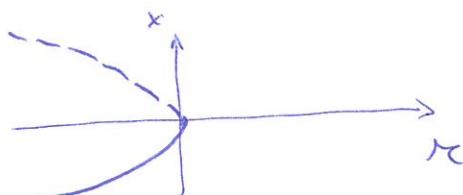
$$r = 0$$

2 equil
coalesce
half-stable



$$r > 0$$

no equil
point.



bifurcation diagram.

Terminology

General form of a nonlinear system

$$(x) \quad \dot{x} = f(t, x, u) \quad t = \text{time} \in \mathbb{R}_+$$

x = state vector $\in \mathcal{D} \subset \mathbb{R}^n$

u = control input $\in \mathcal{U} \subset \mathbb{R}^m$

$f: \mathbb{R}_+ \times \mathcal{D} \times \mathcal{U} \rightarrow \mathbb{R}^n$ vector field

System is autonomous (or time-invariant)

if t does not appear ~~not~~ explicitly on the r.h.s.

i.e. $\dot{x} = f(x, u)$

more precisely: $x(t_0, t, t_0) = x(t - t_0, x_0)$

examples:

- linear system

$$\dot{x} = Ax + Bu$$

- bi-linear system

$$\dot{x} = Ax + Bxu$$

bi-linear: linear in both x and u

- control-affine

$$\dot{x} = \tilde{f}(x) + \sum_{i=1}^m g_i(x) u_i \quad u_i: \text{scalar}$$

drift

- driftless if $f(x) = 0$

$$\dot{x} = \sum_{i=1}^m g_i(x) u_i$$

- system without control

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

Existence and uniqueness of solutions

(n1)

Initial value problem for ODE

$$\begin{cases} \dot{x} = f(t, x) \\ x(t_0) = x_0 \end{cases}$$

- when \exists a solution ?

- when is the solution unique ?

$f: \mathbb{R}_+ \times D \rightarrow \mathbb{R}^n$ we assume $D \subset \mathbb{R}^n$ open

def open set if any $x_0 \in D$ has a neighborhood inside D

i.e. $\exists \epsilon > 0$ s.t. $\|x - x_0\| < \epsilon \Rightarrow x \in D$

what norm? any norm!

def norm $\|\cdot\|$ is an \mathbb{R} -valued function s.t.

1) $\|x\| \geq 0 \quad \forall x \in \mathbb{R}^n, \|x\| = 0 \iff x = 0$

2) $\|x+y\| \leq \|x\| + \|y\| \quad \forall x, y \in \mathbb{R}^n$
(triangle inequality)

3) $\|\alpha x\| = |\alpha| \|x\| \quad \forall \alpha \in \mathbb{R}, x \in \mathbb{R}^n$

example : p-norm

$$\|x\|_p = \left(|x_1|^p + \dots + |x_n|^p \right)^{1/p} \quad 1 \leq p < \infty$$

$$\|x\|_\infty = \max_i |x_i|$$

$p=2$: Euclidean norm

Prop all p-norms are equivalent: given α, β

$\exists c_1, c_2 \in \mathbb{R}_+$ s.t.

$$c_1 \|x\|_\alpha \leq \|x\|_\beta \leq c_2 \|x\|_\alpha \quad \forall x \in \mathbb{R}^n$$

Back to the initial value problem

- for existence of solution, continuity of $f(t, x)$ in both t and x is enough
 $f \in C^0(\mathbb{R}_+ \times D)$
- for existence and uniqueness: one can ask
 - continuity in x of f
 - piecewise-continuity in t of f (so that to include for instance cases in which the control $u(t)$ is discontinuous ex: step response)

(Cauchy-Lipschitz; Picard-Lindelöf theorem ...)

-2 local Lipschitz condition

$$\|f(t, x) - f(t, y)\| \leq L \|x - y\| \quad L > 0$$

$\forall x, y \in B(x_0, r) = \{x \in \mathbb{R}^n \text{ s.t. } \|x - x_0\| < r\}$

$\forall t \in [t_0, t_1]$

Thm the system $\begin{cases} \dot{x} = f(t, x) \\ x(t_0) = x_0 \end{cases}$ with the local Lipschitz condition has a unique solution over the interval $[t_0, t_0 + \delta]$ for some $\delta > 0$
 Prop. maximal interval in which the solution can be extended must be open $[t_0, t_0 + \delta_{\max}]$, perhaps ∞
 L = Lipschitz constant (independent of x, y, t)

Global Lipschitz condition: when Lipschitz in \mathbb{R}^n

measuring: in \mathbb{R}^1 , for $f: \mathbb{R} \rightarrow \mathbb{R}$ (autonomous)
 the Lipschitz cond is $\frac{|f(y) - f(x)|}{|y - x|} \leq L$

\Rightarrow "increment" in ODE is bounded even if f is not differentiable

- differentiability is stronger than Lipschitz
- Lipschitz is stronger than continuous

Lemma Let $f(t, x) \in C^0([t_0, t_1] \times D)$.

Assume $\frac{\partial f}{\partial x} \exists$ continuous on $[t_0, t_1] \times D$

If t_1 is convex $\subset D$ \exists const $L > 0$ s.t.

$$\left\| \frac{\partial f}{\partial x}(t, x) \right\| \leq L \quad (\text{norm of the Jacobian})$$

then $\|f(t, x) - f(t, y)\| \leq L \|x - y\|$

meaning: f has bounded partial derivatives $\quad //$

Here $\|A\| = \text{matrix norm} = \text{induced } p\text{-norm}$

$$\|A\|_p = \sup_{x \neq 0} \frac{\|Ax\|_p}{\|x\|_p} = \max_{\|x\|_p=1} \|Ax\|_p$$

examples: Euclidean norm $\|A\|_2 = \left(\max_{i,j} (A^T A) \right)^{1/2}$

$$\|A\|_\infty = \max_i \sum_{j=1}^n |a_{ij}| \quad \text{column sum} \quad \begin{matrix} \text{largest} \\ \text{single} \\ \text{value} \\ \text{of } A \end{matrix}$$

$$\|A\|_1 = \max_j \sum_{i=1}^n |a_{ij}| \quad \text{row sum}$$

example (of non-uniqueness)

$$\dot{x} = z\sqrt{x} \quad \frac{\partial f}{\partial x} = \frac{z}{2\sqrt{x}} \quad \text{not bounded at } x=0$$

\Rightarrow ODE is not locally Lipschitz at $x=0$

\Rightarrow multiple solutions exist

Extension to global uniqueness existence and uniqueness

then $f: \mathbb{R}_+ \times \mathbb{R}^n$ cont in x , piecewise cont in t and satisfies a global Lipschitz cond

$$\|f(t, x) - f(t, y)\| \leq L \|x - y\| \quad \forall x, y \in \mathbb{R}^n$$

then the system $\dot{x} = f(t, x)$
 $x(t_0) = x_0$ has a
unique solution over $[t_0, t_1]$

$$\forall t \in [t_0, t_1]$$

// (L can be arbitrarily large)

- global Lipschitz cond. is difficult to satisfy
 \Rightarrow conservative suff. cond.
- often you can have it only on compact sets

def: compact set = closed and bounded set

def closed set: if its complement is open.

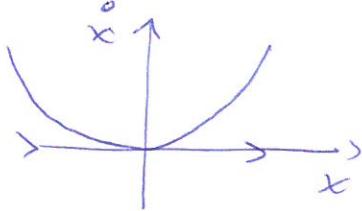
- for autonomous systems: stronger suff. cond.

then If $f \in C^1(\mathcal{D})$ i.e. cont with cont first derivative in x and autonomous, then the solution exists and is unique $\forall t \in \mathbb{R}$ and $\forall t \in [t_0, t_0 + \delta_{\max}]$

example (finite escape time)

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$$\begin{cases} \dot{x} = x^2 \\ x(0) = 1 \end{cases}$$

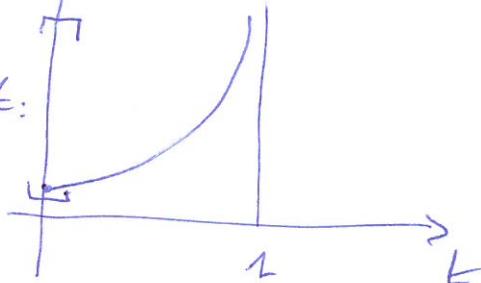


$$\frac{\partial f}{\partial x} = 2x$$

- f is locally Lipschitz (even differentiable)
- Lipschitz in any compact set
- not globally Lipschitz ($\frac{\partial f}{\partial x} \rightarrow \infty$ as $x \rightarrow \infty$)
- solution can be extended only

to $[0, 1]$, not beyond

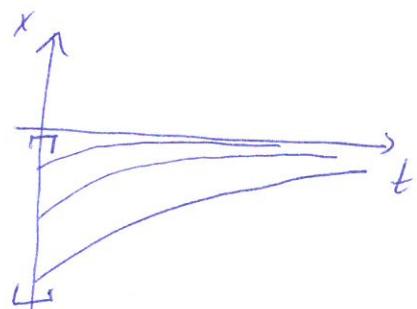
can take
any compact:
+ always
Lipschitz in
any compact



as $t \rightarrow \infty$, $x(t)$
escapes any compact
set

changing initial cond:

$$\begin{cases} \dot{x} = x^2 \\ x(0) = x_0 \end{cases} \Rightarrow x(t) = \frac{x_0}{1 - x_0 t}$$



• if $x_0 > 0$ still finite escape time

• if $x_0 < 0$ solution can be extended to $t \rightarrow \infty$

reason: $x(t) \in W$ compact $W \subset \mathbb{R}$

W compact is also invariant

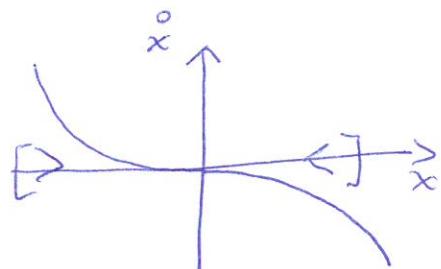
→ how much can the solution be extended
depends on x_0 ! (and I need to know the
solution is unfeasible in general).

existence and uniqueness in compact sets

Theorem If $f(t, x)$ cont. in $x \in D$, piecewise cont. in $t \in [t_0, t_1]$, locally Lipschitz in $[t_0, t_1] \times D$ and if $\exists W \subset D$ compact and invariant for the system, then $x_0 \in W \Rightarrow$ the solution \exists unique on $[t_0, t_1]$

def W is an invariant set for $\dot{x} = f(t, x)$ if $x_0 \in W \Leftrightarrow x(t) \in W \quad \forall t$

example $\begin{cases} \dot{x} = -x^3 \\ x(0) = x_0 \end{cases}$



$f(x)$ is Locally Lipschitz on \mathbb{R}
solution never leaves the
compact set $\{x \in \mathbb{R} \text{ s.t. } |x| \leq |x_0|\}$
 \Rightarrow solution can be extended
till $t \rightarrow \infty$

Lemma If $f(t, x)$ and $\frac{\partial f}{\partial x}(t, x)$ cont. in $[t_0, t_1] \times \mathbb{R}^n$
then f is globally Lipschitz $\Leftrightarrow \frac{\partial f}{\partial x}$ is
uniformly bounded on $[t_0, t_1] \times \mathbb{R}^n$

STABILITY ANALYSIS

Consider an autonomous system (without control)

$$\dot{x} = f(x) \quad x \in D \subset \mathbb{R}^n \quad D \text{ open}$$

equilibrium point can always be considered
the origin $\bar{x} = 0$

In fact, if $\bar{x} \neq 0$, call $z = x - \bar{x}$

$$\dot{z} = \dot{x} = f(z - \bar{x}) = \bar{f}(z)$$

with $\bar{z} = 0$ as equil point

def $\bar{x} = 0$ equil point is

- stable if for each $\epsilon > 0 \exists \delta > 0$ s.t.

$$\|x(0)\| < \delta \Rightarrow \|x(t)\| < \epsilon \quad \forall t \geq 0$$

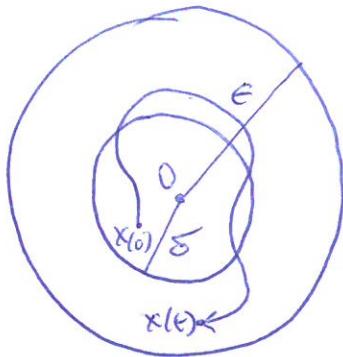
- unstable if it is not stable

- (locally) asymptotically stable if it is stable and δ can be chosen such that

$$\|x(0)\| < \delta \Rightarrow \lim_{t \rightarrow \infty} x(t) = 0$$

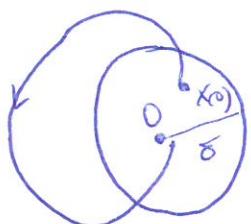
meaning

- stability



provided that you start in $B(0, \delta)$
you never exit
the ball $B(0, \epsilon)$

- asymptotic stability : stability + convergence



convergence: provided you start inside the ball $B(0, \delta)$
you eventually converge to 0

Lyapunov direct method

Main idea: find \rightarrow given $\begin{cases} \dot{x} = f(x) \\ x(0) = 0 \end{cases}$ find a function

V for which $\bar{x} = 0$ is a local minimum
(V is called Lyapunov function)

For asymptotic stability we would like to have

- $V : D \rightarrow \mathbb{R}$ which is $C^1(D)$
- $V(0) = 0$
- $V(x) > 0$ in $D \setminus \{0\}$ \rightarrow positive definite function
- derivatives of V along the trajectories of the system decrease (i.e. solutions of the system move towards the local min of V)

$$\dot{V}(x(t)) = \underbrace{\frac{\partial V}{\partial x} \dot{x}}_{\text{gradient}} = \frac{\partial V}{\partial x} f(x) = \sum_{i=1}^n \frac{\partial V}{\partial x_i} f_i(x) \leq 0$$

if $f(x) = \begin{bmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{bmatrix}$

$\dot{V}(x) \leq 0$ negative semidefinite function
 $(\dot{V}(0) = 0)$

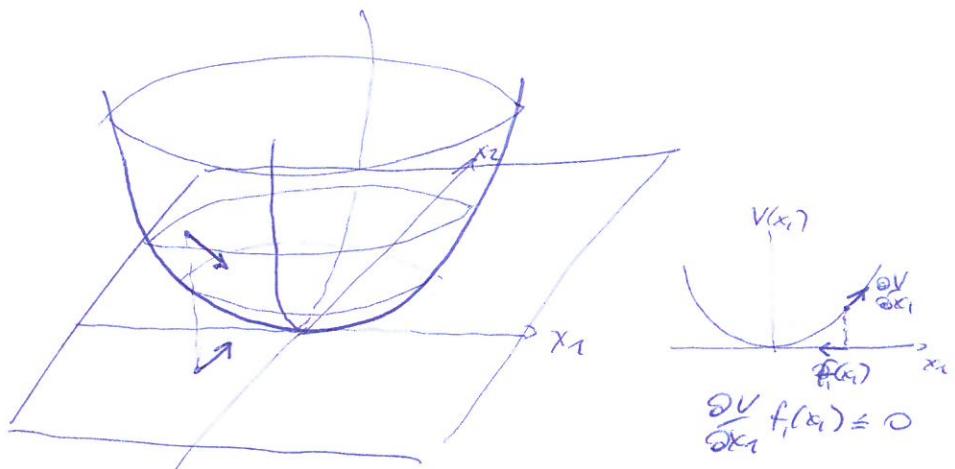
classical examples:

- energy function or potential energy of a mechanical system
- quadratic functions of linear systems

$$V(x) = x^T P x \quad P = P^T > 0 \quad \text{posit. def.}$$

- picture in \mathbb{R}^2

$\dot{V}(x)$ decreases
iff x "gets closer"
to $\tilde{x}=0$



- key point: avoid solving the ODE ($\dot{V}(x)$) is computed along the trajectories but since

$\dot{V}(x) = \frac{\partial V}{\partial x} f(x)$, I do not need the explicit solution to do so)

- $\dot{V}(x) \leq 0 \Rightarrow x(t)$ stays in a compact + Lipschitz
 \Rightarrow solution can be prolonged $\forall t > 0$
 \Rightarrow solution that stays in an arbitrarily small
compact around $\tilde{x}=0$ is stable (informally)