
Stability

Master degree in Automation Engineering

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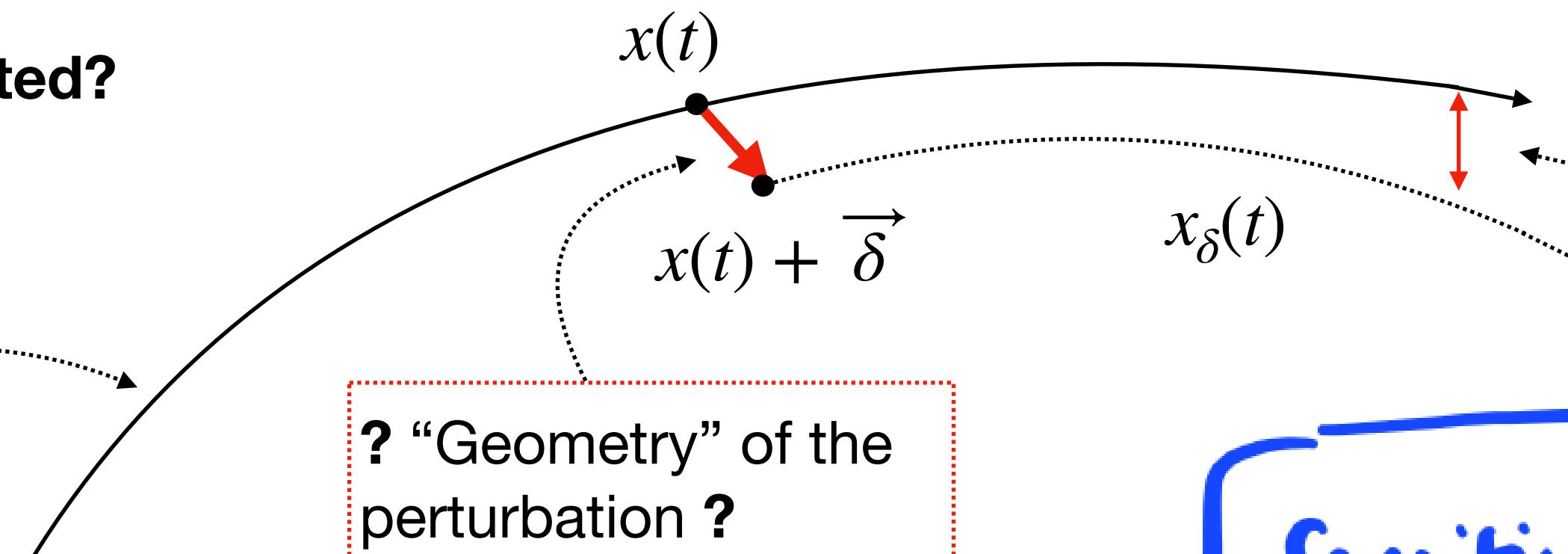
Stability \rightarrow how the system's trajectory behaves when being introduced a perturbation in its input

Very important area in Systems Theory aiming to study the behaviour of trajectories when a perturbation is applied.

How perturbed trajectories are related?

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

? Properties dependent on the "starting" "nominal trajectory"?



? "Geometry" of the perturbation ?

? remain close (same order of the magnitude of δ), diverge, asymptotically converge ?

Sensitivity (to initial conditions) analysis

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

? what about perturbation introduced by inputs?

$$\dot{\bar{x}} = \bar{f}(\bar{x}), \quad \bar{x} |_{\bar{x}(0)} = \bar{x}_0, \quad \bar{x}(t) \equiv \bar{x}$$

Starting point: autonomous systems (no inputs) and "reference" trajectory (the one that is perturbed) is an equilibrium point

$\epsilon - \delta_\epsilon$ Stability

Let's study the behaviour of the system close to an equilibrium point when the initial state is perturbed. Without loss of generality we take the equilibrium point to be the origin ($f(0) = 0$). n any (not necessarily planar or 3-D systems).

Stability: the magnitude of the perturbed trajectory is of the same order as the perturbation.

$\bar{x} \approx \text{Origin}$

The equilibrium $x = 0$ is stable if for all $\epsilon > 0$ there exists a $\delta_\epsilon > 0$ such that for all initial conditions satisfying $\|x(0)\| \leq \delta_\epsilon$ the resulting trajectory satisfies $\|x(t)\| \leq \epsilon$ for all $t \geq 0$.

ϵ is any number

Asymptotic stability: stability and the effect of the perturbation is vanishing.

stability + attractivity

The equilibrium $x = 0$ is asymptotically stable if it is stable and there exists a set $\mathcal{A} \supset \{0\}$ such that $x(0) \in \mathcal{A}$ implies $\lim_{t \rightarrow \infty} x(t) = 0$ (attractivity property). Terminology: \mathcal{A} is said to be “Domain of Attraction”.

Instability: lack of stability.

The equilibrium $x = 0$ is unstable if it is not stable (not joking!).

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

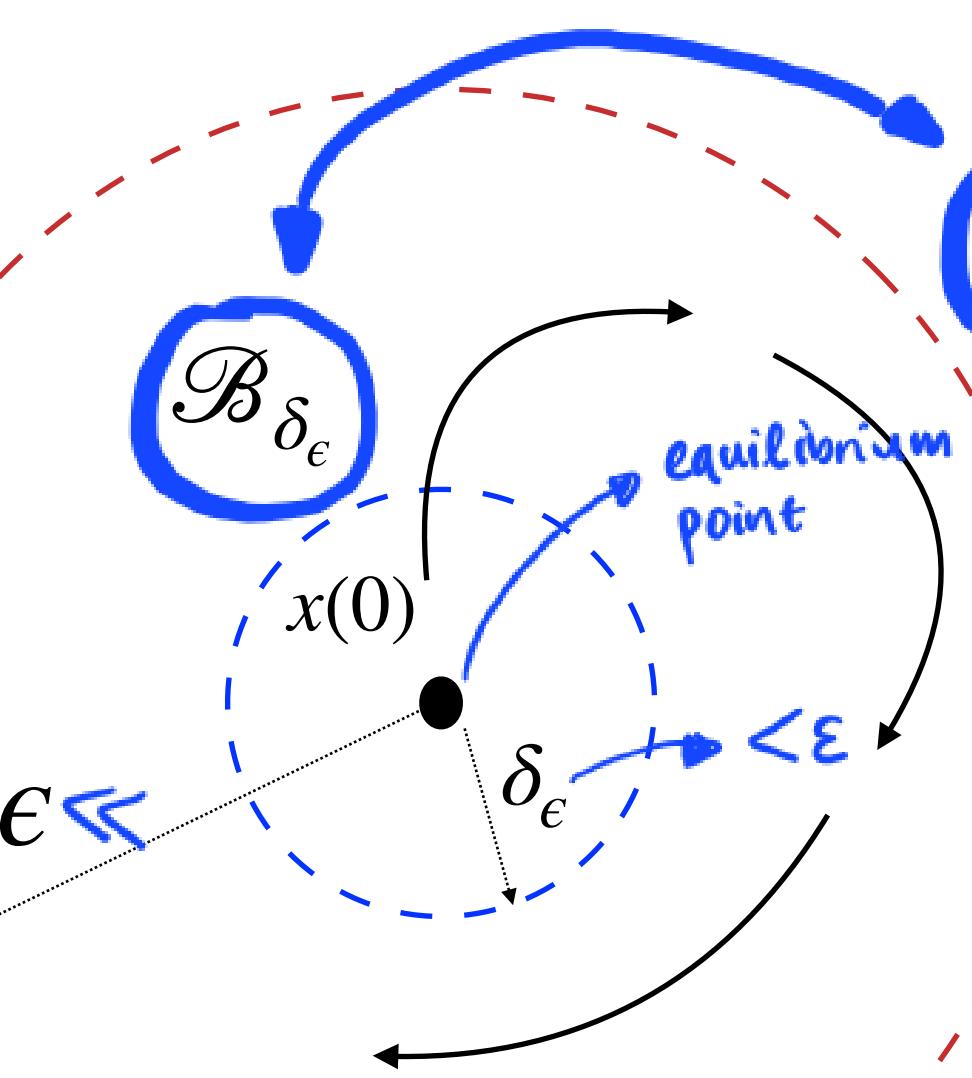
it's the Origin

$$\tilde{x} = x - \bar{x}$$

$$\dot{\tilde{x}} = f(\tilde{x} + \bar{x})$$

$\epsilon - \delta_\epsilon$ stability notion

Stability



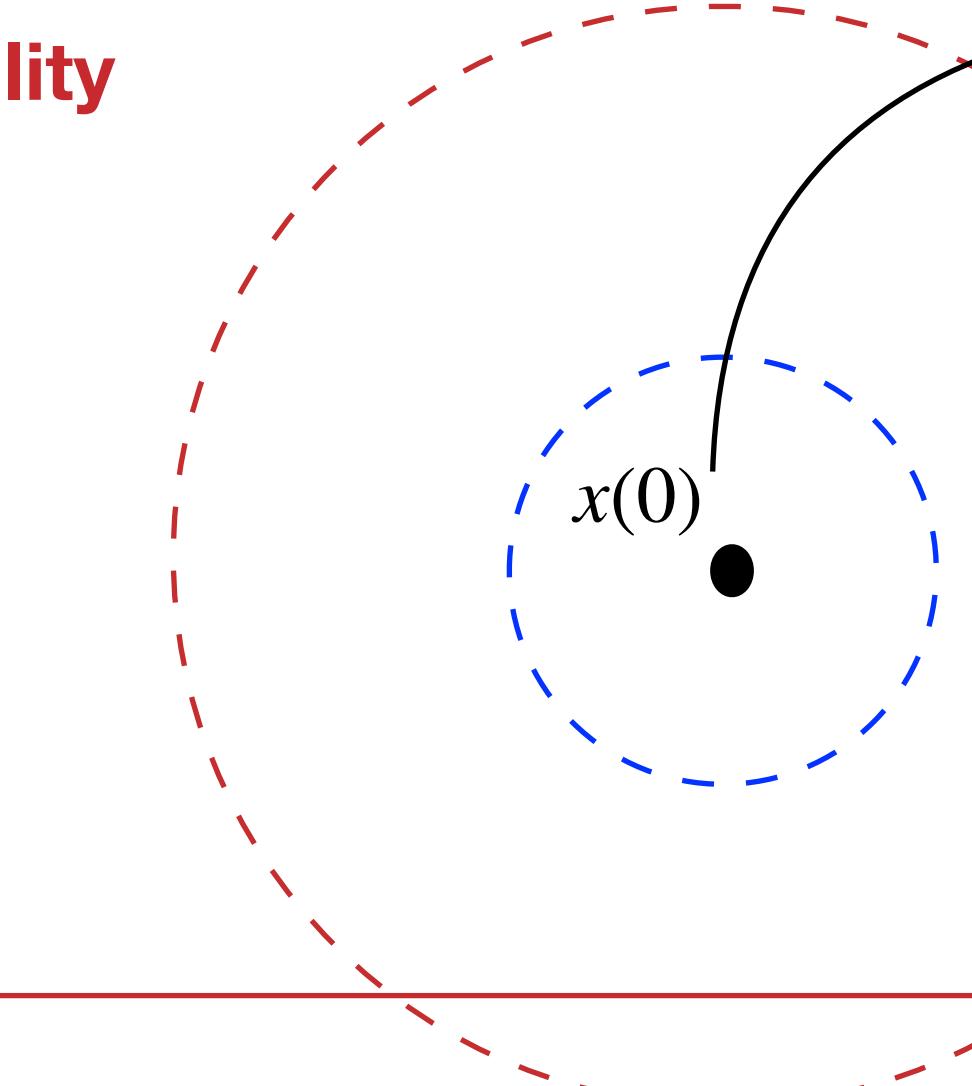
circle in the plane, sphere in the space, ...

$$\mathcal{B}_\epsilon := \{x \in \mathbb{R}^n : \|x\| \leq \epsilon\}$$

ϵ is given (any, potentially very small), there must exist a δ_ϵ (smaller..) to have stability of the equilibrium point.

$x(0)$ is not chosen (for all initial conditions in the ball of radius δ_ϵ ...)

Instability

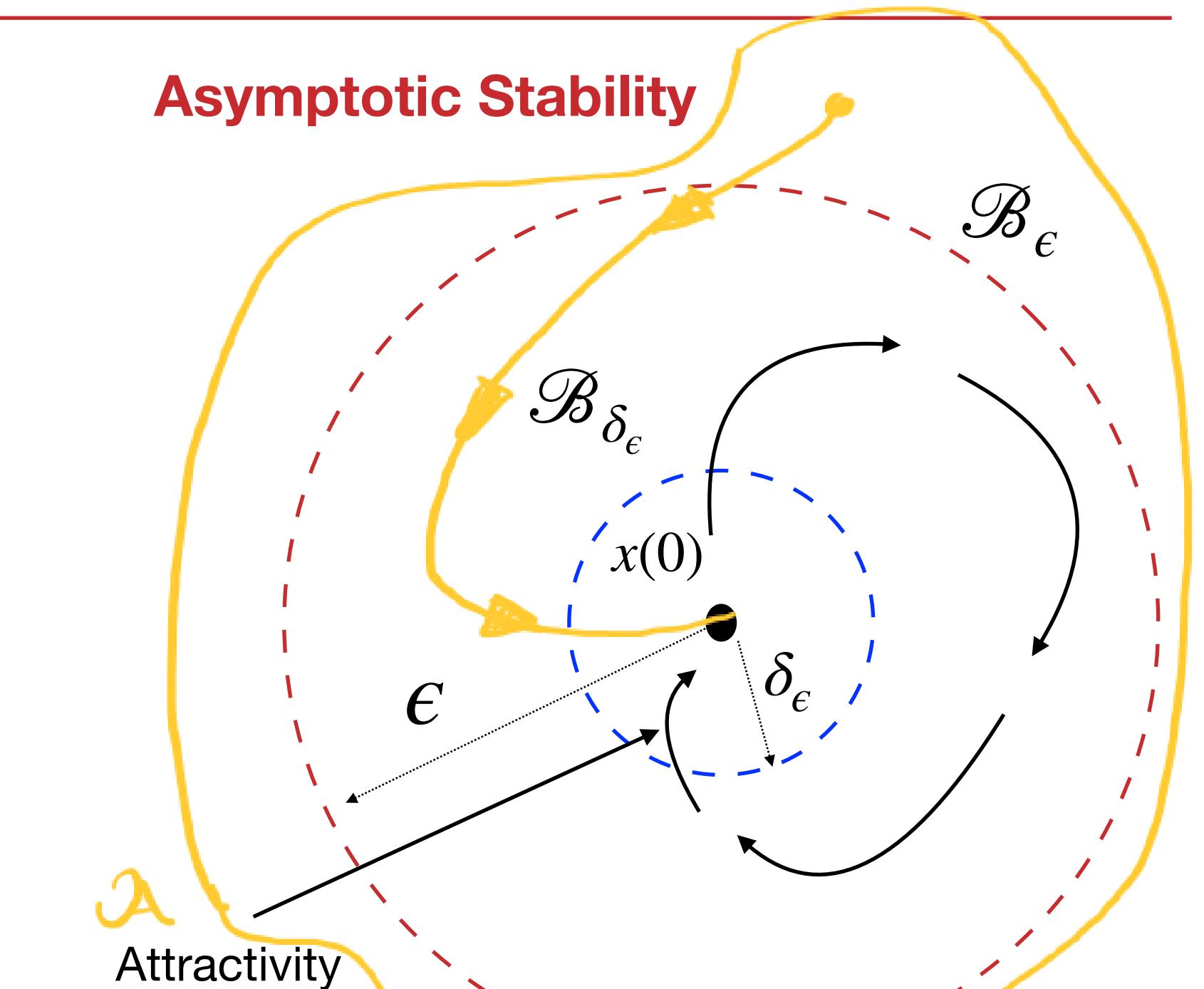


to ∞ ... but not necessarily

→ explode to infinity

Homoclinic orbit
it's unstable but asymptotically approaches the initial condition

Asymptotic Stability



Attractivity

↳ does not imply stability

Stability + attractivity

No matter which δ_ϵ is taken, at least one trajectory exits \mathcal{B}_ϵ

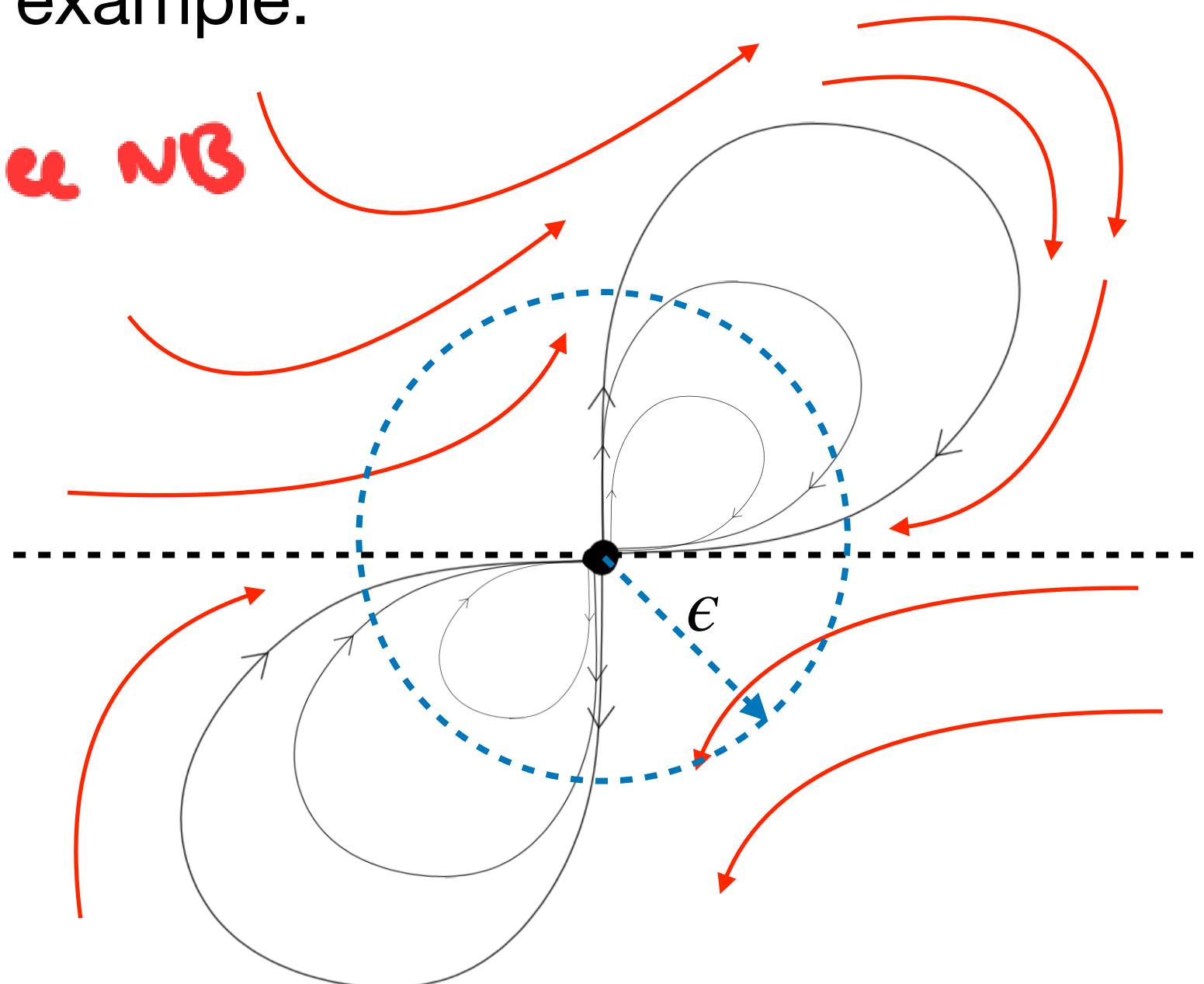
Attractivity does not imply stability
(in a general nonlinear setting)

Remarks

- The lack of stability in a nonlinear system is not necessarily associated to perturbed trajectories that diverges from the equilibrium. Homoclinic example:

$$\begin{aligned} \dot{n}_1 &= f_1(n_1, n_2) \\ \dot{n}_2 &= f_2(n_1, n_2) \end{aligned} \quad \left. \right\} \text{see NB}$$

Say that the perturbation aligns w/ the orbit,
 $\|n(z)\| \approx 10^{-40}$, and
 $\|n(z(t))\| \approx 1$, there is
 no δ_ϵ



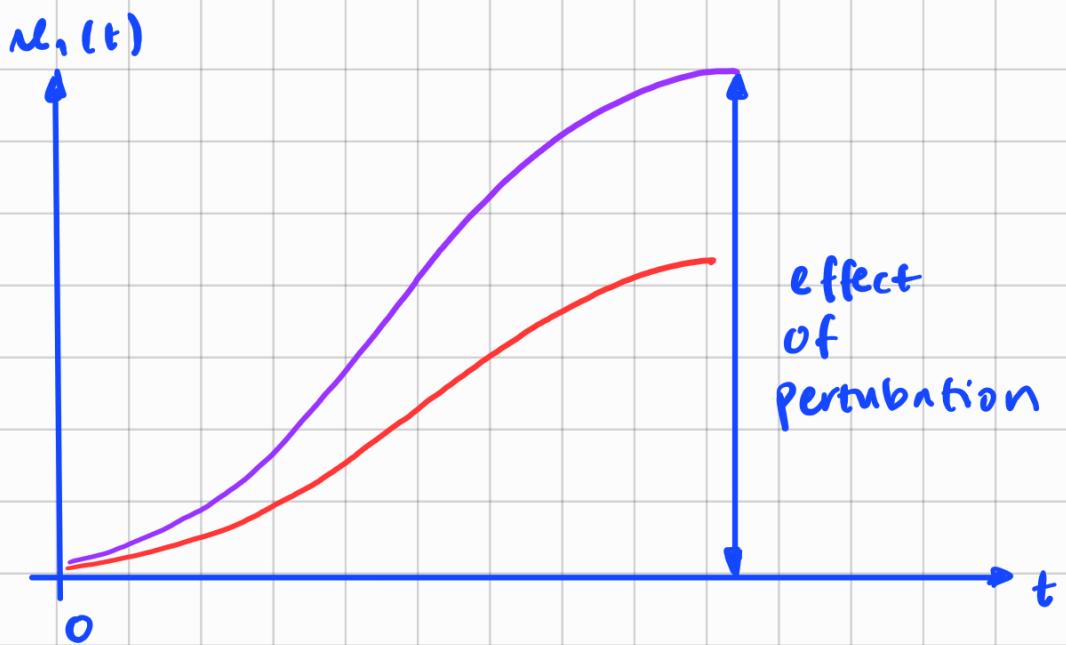
Given a circle of radius ϵ with the latter taken so that the circle intercept the homoclinic orbit, a δ_ϵ doesn't exist... and the origin is "globally" attractive

- The property of stability refers to an equilibrium point. Different equilibria of the same (nonlinear) system might have different stability properties. That is, the property of stability is not linked to the system but to a specific equilibrium
- For linear systems the notion of stability can be referred to the system and can be extended to generic trajectories (not necessarily equilibria)

NB:

$$\ddot{n}_1 = \frac{n_1^2(n_2 - n_1) + n_2^5}{(n_1^2 + n_2^2)(1 + (n_1^2 + n_2^2)^2)} \text{ (X)}$$

$$\dot{n}_2 = \frac{n_2^2(n_2 - 2n_1)}{(n_1^2 + n_2^2)^2} \text{ (X)}$$



LINEAR SYSTEM STABILITY

$$\dot{x} = Ax$$

$$x(t+1) = Ax(t), \text{ where } x(t) = \phi(t) \cdot x(0)$$

$$x_{\text{ref}}(0) = 0$$

$$\forall \epsilon > 0 \exists \delta_\epsilon > 0 : \|x(0)\|$$

$$\leq \delta_\epsilon \Rightarrow \|x(t)\| \leq \epsilon, \forall t \geq 0$$

$$\|x(t) - 0\| = \|\phi(t) \cdot x(0)\| \leq \|\phi(t)\| \cdot \|x(0)\| \leq \epsilon$$

$$\leq \delta_\epsilon ?$$

given

If $\phi(t)$ is bounded for $\forall t \geq 0$ (there exists $\exists M$:

$$\|\phi(t)\| \leq M \text{ for } \forall t \geq 0$$

$$\delta_\epsilon = \frac{\epsilon}{M}$$

$$\leq M \|x(0)\|$$

The case of linear systems

$$\begin{Bmatrix} \dot{x}(t) \\ x(t+1) \end{Bmatrix} = Ax(t) \quad x(0) = x_0$$

$$x(t) = \phi(t)x_0$$

$$\phi(t) = \begin{cases} e^{At} & \text{for C-T systems} \\ A^t & \text{for D-T systems} \end{cases}$$

For all $\epsilon > 0$ there exists a $\delta_\epsilon > 0$ such that for all x_0 fulfilling $\|x_0\| \leq \delta_\epsilon$ then $\|x(t)\| = \|\phi(t)x_0\| \leq \epsilon$ for all $t \geq 0$

Since $\|\phi(t)x_0\| \leq \|\phi(t)\| \|x_0\| \leq \|\phi(t)\| \delta_\epsilon$ it turns out that:

If $\exists M > 0 : \|\phi(t)\| \leq M \forall t \geq 0$ then the system is stable (pick $\delta_\epsilon = \frac{\epsilon}{M}$)

If $\exists M > 0 : \|\phi(t)\| \leq M \forall t \geq 0$ and $\phi(t) \rightarrow 0$ as $t \rightarrow \infty$ then the system is asymptotically stable

Theorem: A C-T (D-T) linear system is stable iff all the eigenvalues of A have a nonpositive real part (amplitude ≤ 1) and possible eigenvalues with zero real part (amplitude $= 1$) have a geometric multiplicity equal to the algebraic one. It is asymptotically stable iff the eigenvalues of A have all negative real part (amplitude < 1)

The case of linear systems

Notation:

C-T asymptotically stable systems $\iff A$ **Hurwitz**

D-T asymptotically stable systems $\iff A$ **Schur**

in negative real value
within circle

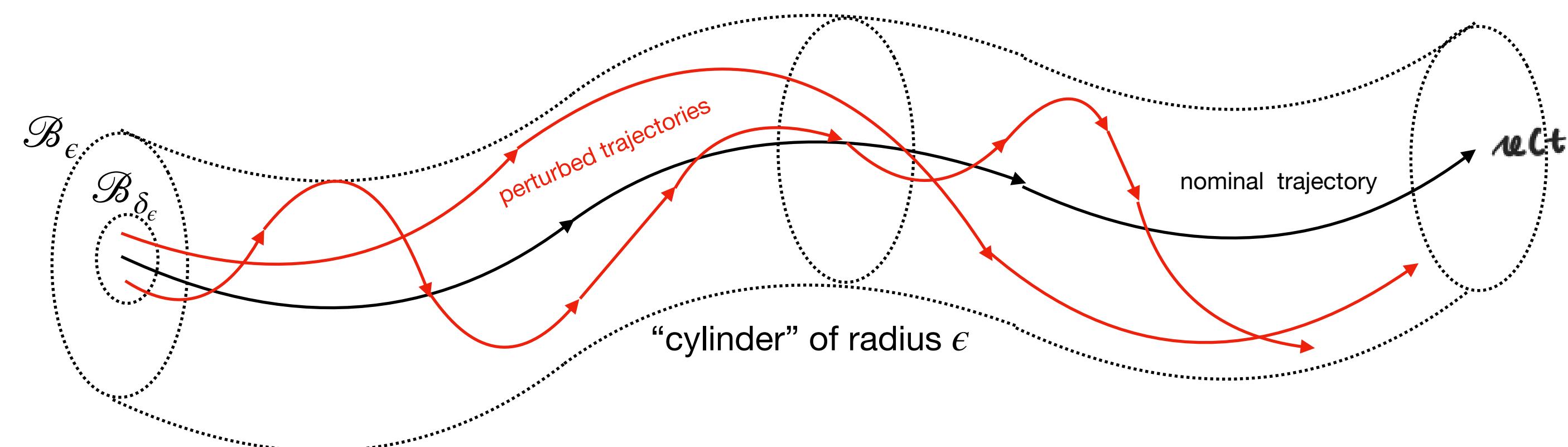
For linear systems the notion and the properties can be extended to “generic trajectories”
generic

$$x(t) = \phi(t)x_0 + \Psi(t)u([0,t])$$

$$\bar{x}(t) = \phi(t)(x_0 + \delta) + \Psi(t)u([0,t])$$

$$\|x(t) - \bar{x}(t)\| = \|\phi(t)\delta\| \leq \|\phi(t)\| \|\delta\|$$

If $\phi(t)$ is bounded then the ϵ/δ_ϵ condition is fulfilled



Furthermore, if A Hurwitz then the trajectory generated by any bounded input $u(t)$ is also bounded (BIBS)

Lyapunov Direct Theorem

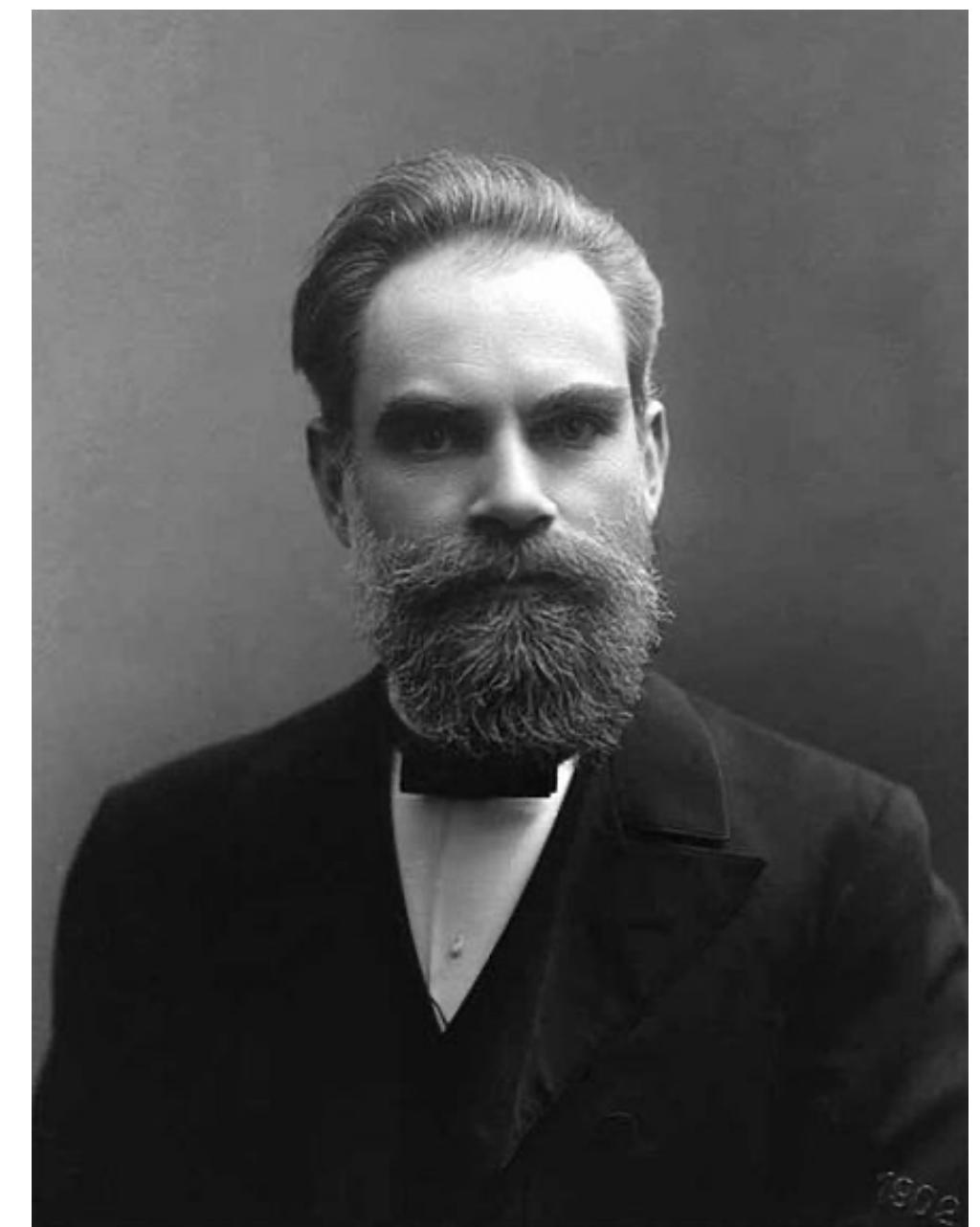
alternative
definition of ϵ - δ stability

The $\epsilon - \delta_\epsilon$ stability notion is often said also “Lyapunov Stability” because of a result, due to A. M. Lyapunov, providing sufficient and necessary conditions under which an equilibrium point of a nonlinear dynamic system is stable/asymptotically stable.

While the $\epsilon - \delta_\epsilon$ stability notion involves, in principle, trajectories $x(t)$ and, thus, would involve solving the differential equation $\dot{x} = f(x)$, the Lyapunov result is rather “geometric” in the sense that it involves only checking state-wise a real-valued function defined in the state space.

The result has a nice “energy” interpretation: if an equilibrium point is stable then there exists a form of “energy” function that is not increasing along the system trajectories. In case of an asymptotically stable equilibrium point such a energy function is also decreasing reaching a minimum at the equilibrium.

we lose the t variable, but we inspect the geometry



Aleksandr M. Lyapunov, 1857 - 1918

the idea why this theorem developed is because plotting $u(t)$ could be in \mathbb{R}^n

Lyapunov Direct Theorem

open set where
 V is defined

real-valued function

associated
with real number

Definition. A function $V : \mathcal{D} \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ is said to be **positive definite** with respect to $\bar{x} \in \mathcal{D}$ if $V(\bar{x}) = 0$ and $V(x) > 0$ for all $x \in \mathcal{D} \setminus \{\bar{x}\}$. Notation $V > 0$ (\bar{x} implicit from the context).

Definition. A function $V : \mathcal{D} \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ is said to be **positive semi-definite** with respect to $\bar{x} \in \mathcal{D}$ if $V(\bar{x}) = 0$ and $V(x) \geq 0$ for all $x \in \mathcal{D}$. Notation $V \geq 0$ (\bar{x} implicit from the context).

Definition. **Negative definite** and **semi-definite** functions can be similarly defined in a straightforward way.

Examples in \mathbb{R}^2

$$V(x) = x_1^2 + x_2^2$$

> 0 wrt the origin

$$sg(s) > 0$$

$$V(x) = x_1^2 + \int_0^{x_2} g(s) \, ds$$

> 0 wrt the origin

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

> 0 wrt the origin

$$\mathcal{D} = \{(x_1, x_2) : |x_2| < \pi/2\}$$

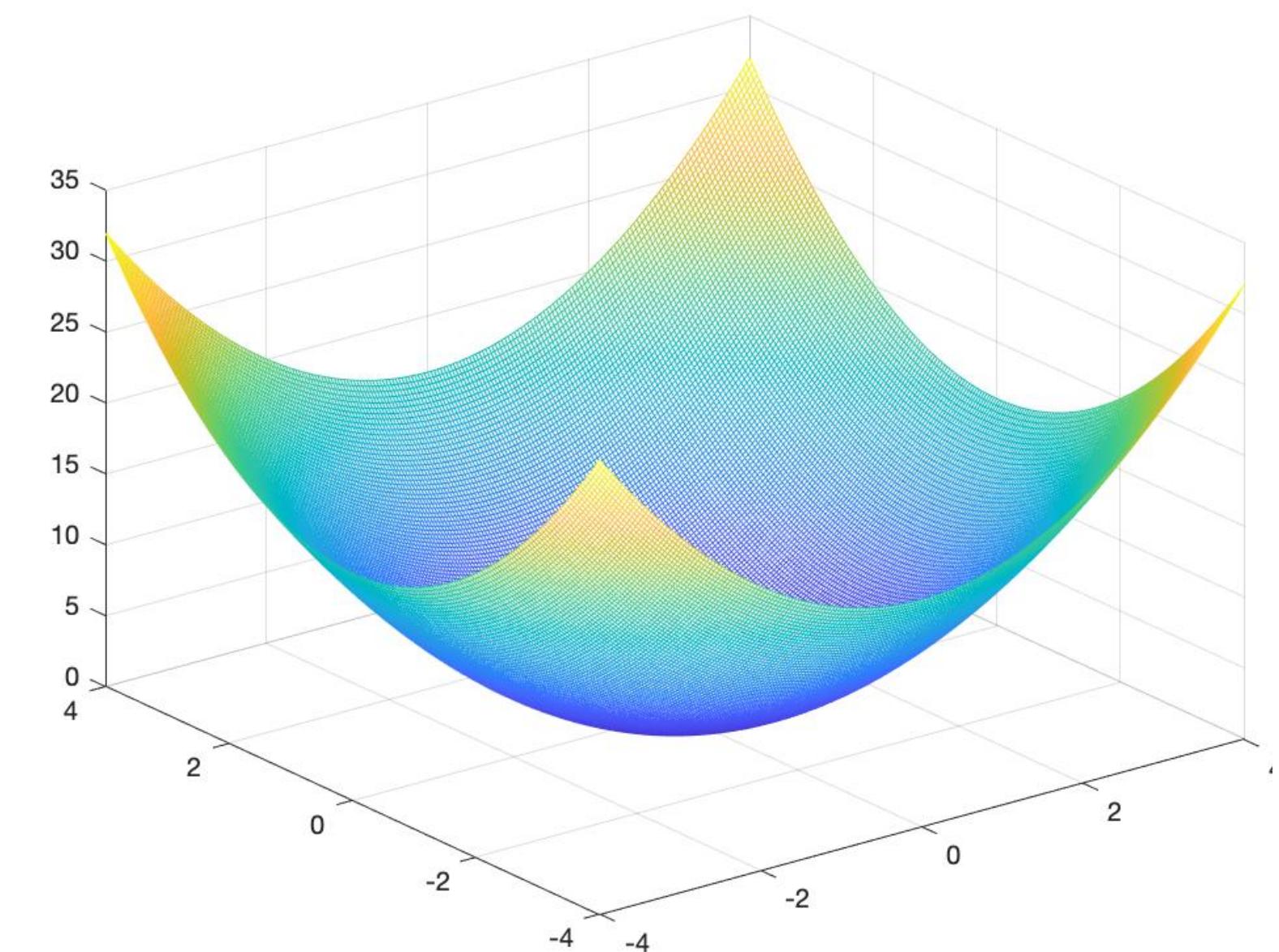
$$V(x) = x_1^4$$

≥ 0 wrt any point in
 $\{(x_1, x_2) : x_1 = 0\}$

Examples with Matlab

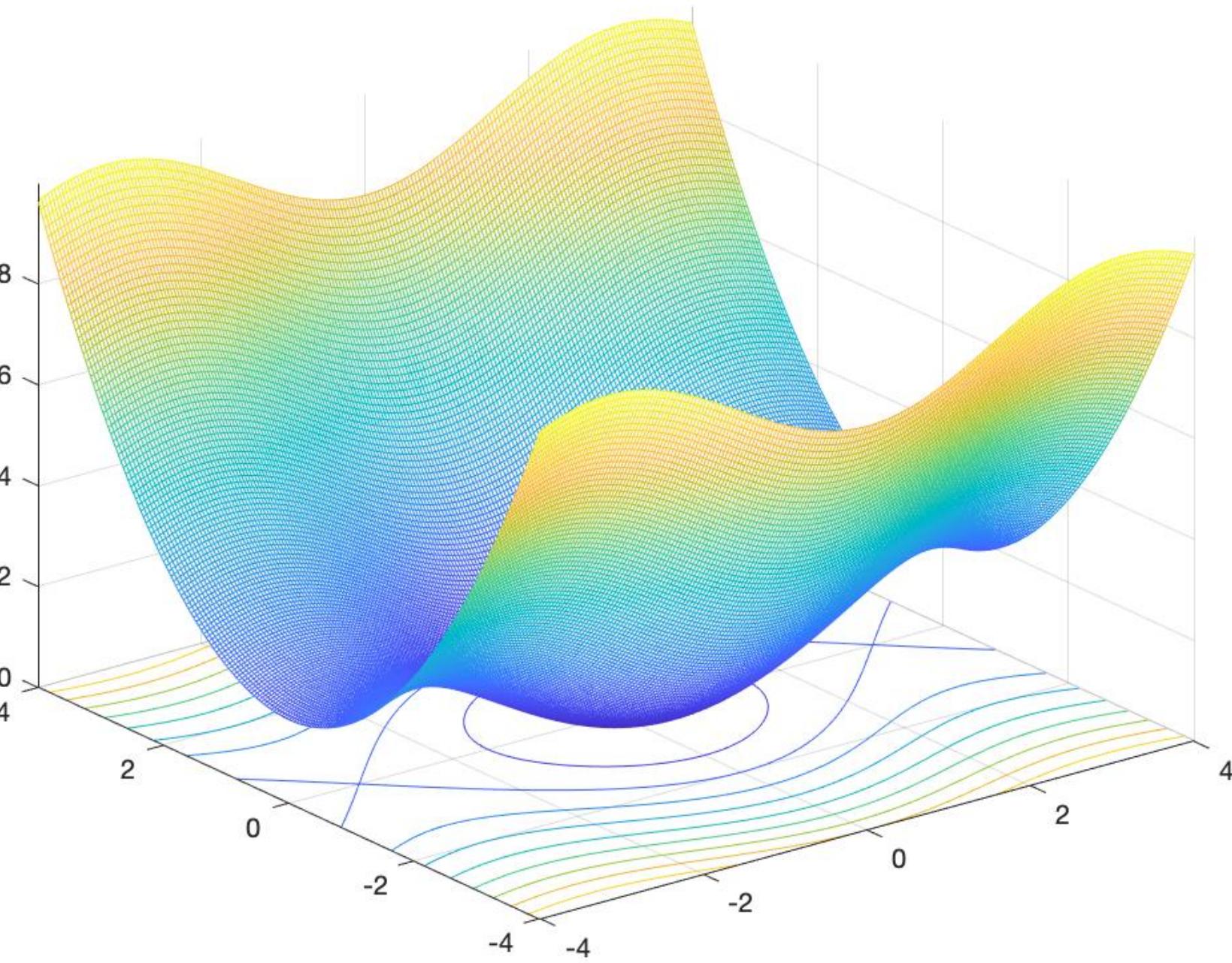
$$V(x, y) = x^2 + y^2$$

```
x=[-4:.04:4];  
y=x;  
[X,Y]=meshgrid(x,y);  
z=X.^2 + Y.^2;  
mesh(X,Y,z)
```



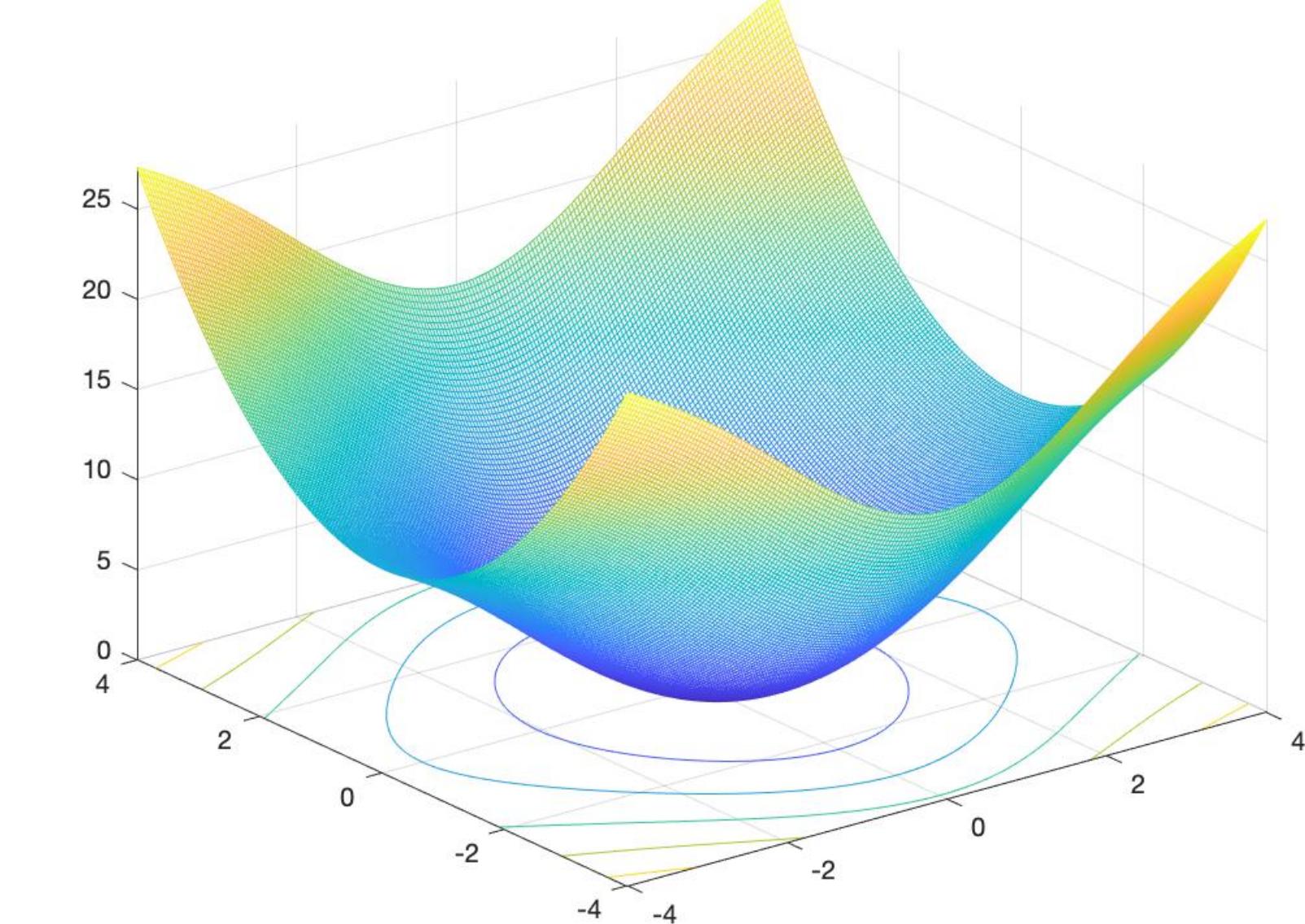
$$V(x, y) = 1 - \cos(x) + \frac{1}{2}y^2$$

```
x=[-4:.04:4];  
y=x;  
[X,Y]=meshgrid(x,y);  
z=(1-cos(X))+(Y.^2)/2;  
meshc(X,Y,z)
```



$$V(x, y) = 2(1 - \cos(x)) + \frac{1}{2}y^2 + \frac{1}{2}(x^2 + y^2)$$

```
x=[-4:.04:4];  
y=x;  
[X,Y]=meshgrid(x,y);  
z=2*(1-cos(X))+(X.^2 + Y.^2)/2 + (Y.^2)/2;  
meshc(X,Y,z)
```



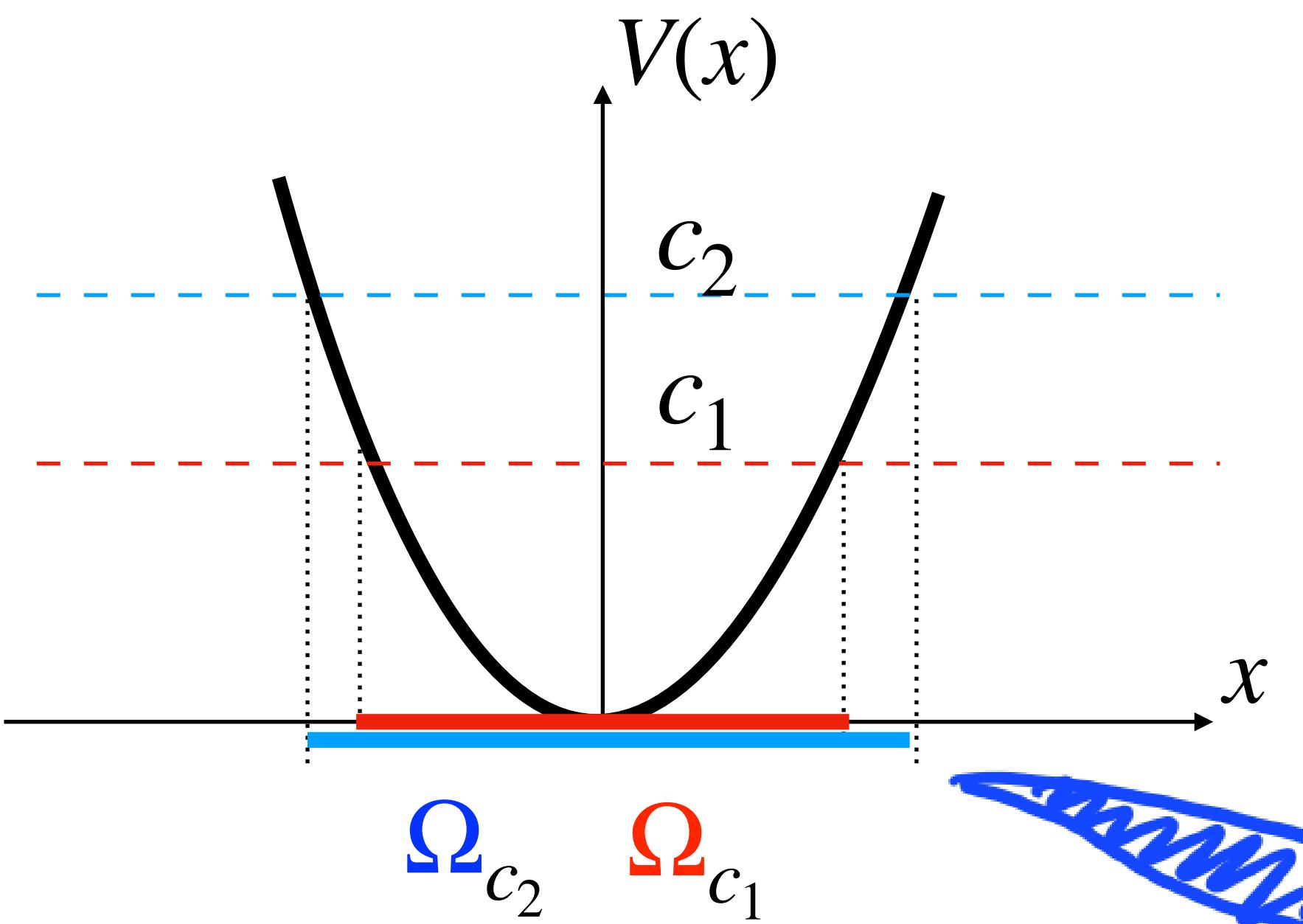
Lyapunov Direct Theorem

in \mathbb{R}^n

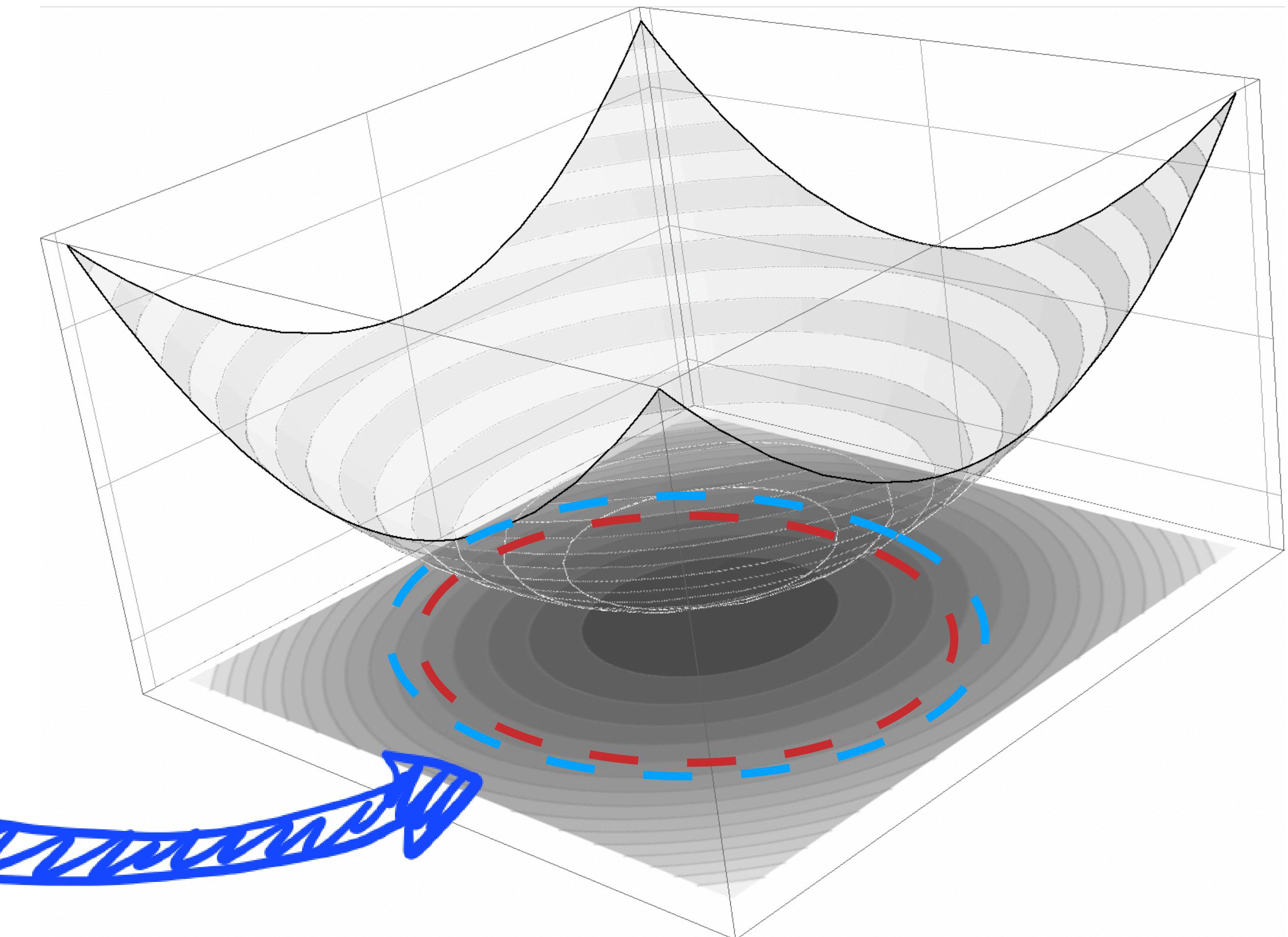
Particularly relevant in the analysis that will come is the notion of level set of $V(x)$

a set

$$\Omega_c := \{x \in \mathbb{R}^n : V(x) \leq c\}$$



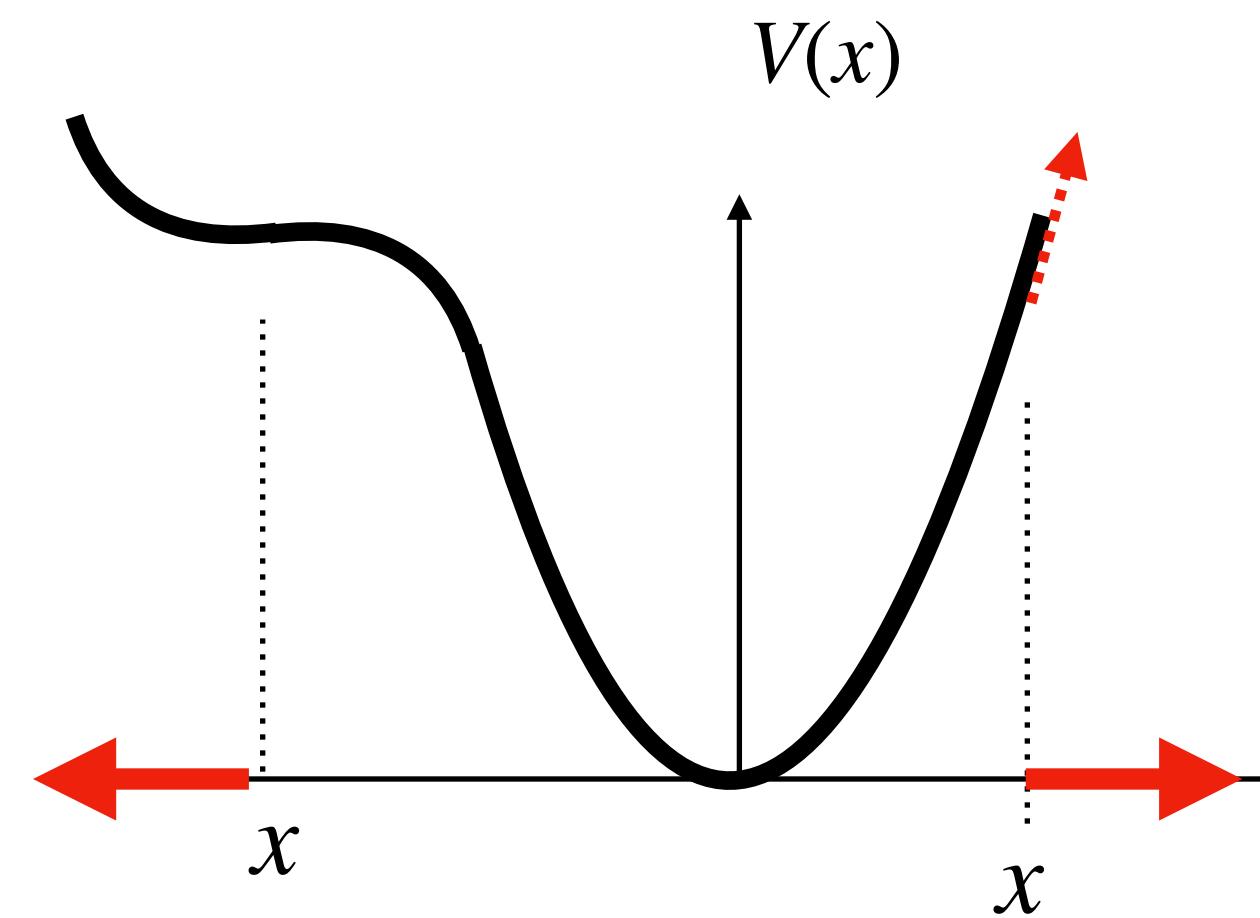
$$c_1 < c_2 \Rightarrow \Omega_{c_1} \subset \Omega_{c_2}$$



Lyapunov Direct Theorem

*vector of partial derivative
of function V*

... and the notion of divergence of $V(x)$ at the point x :

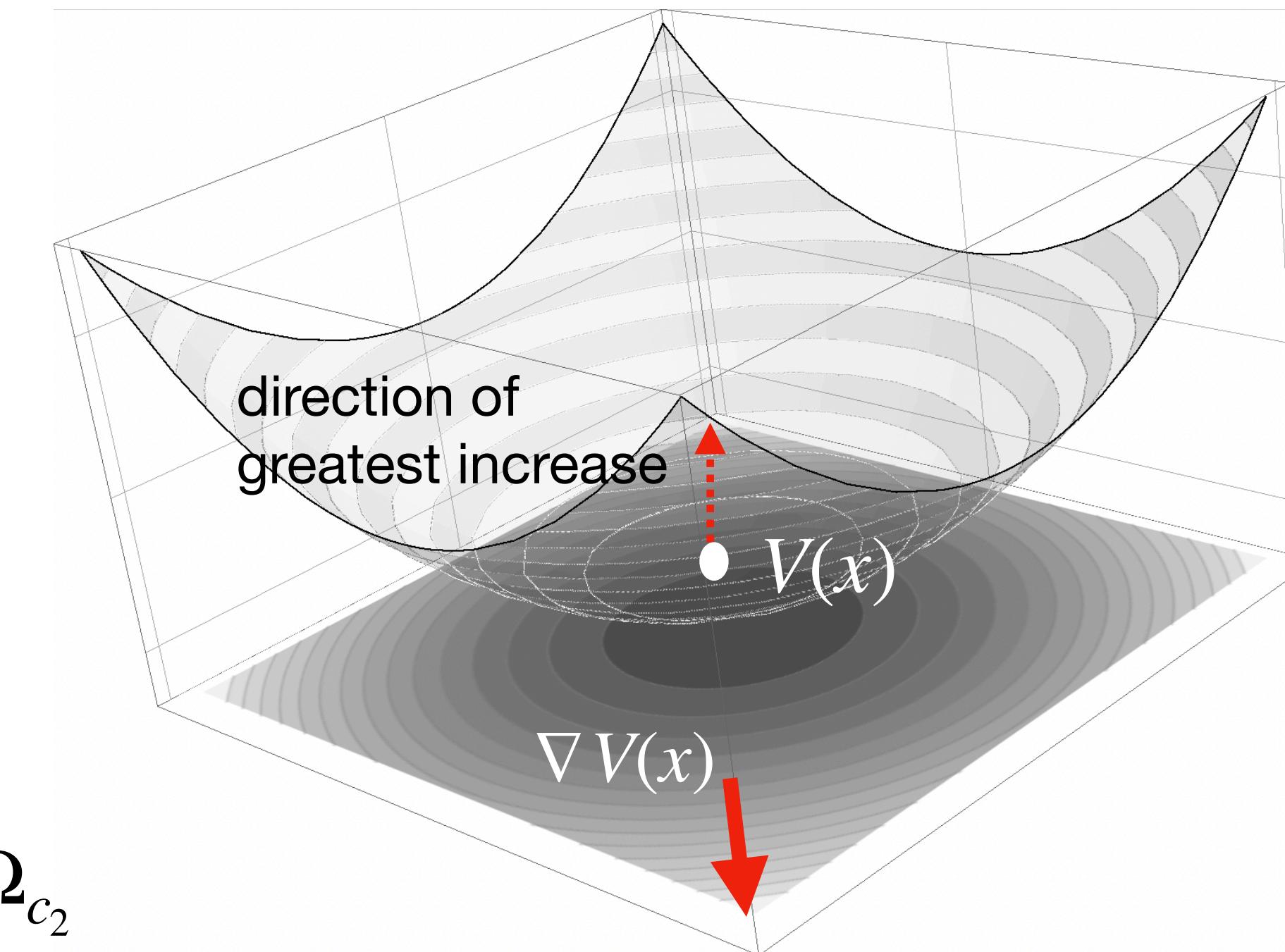
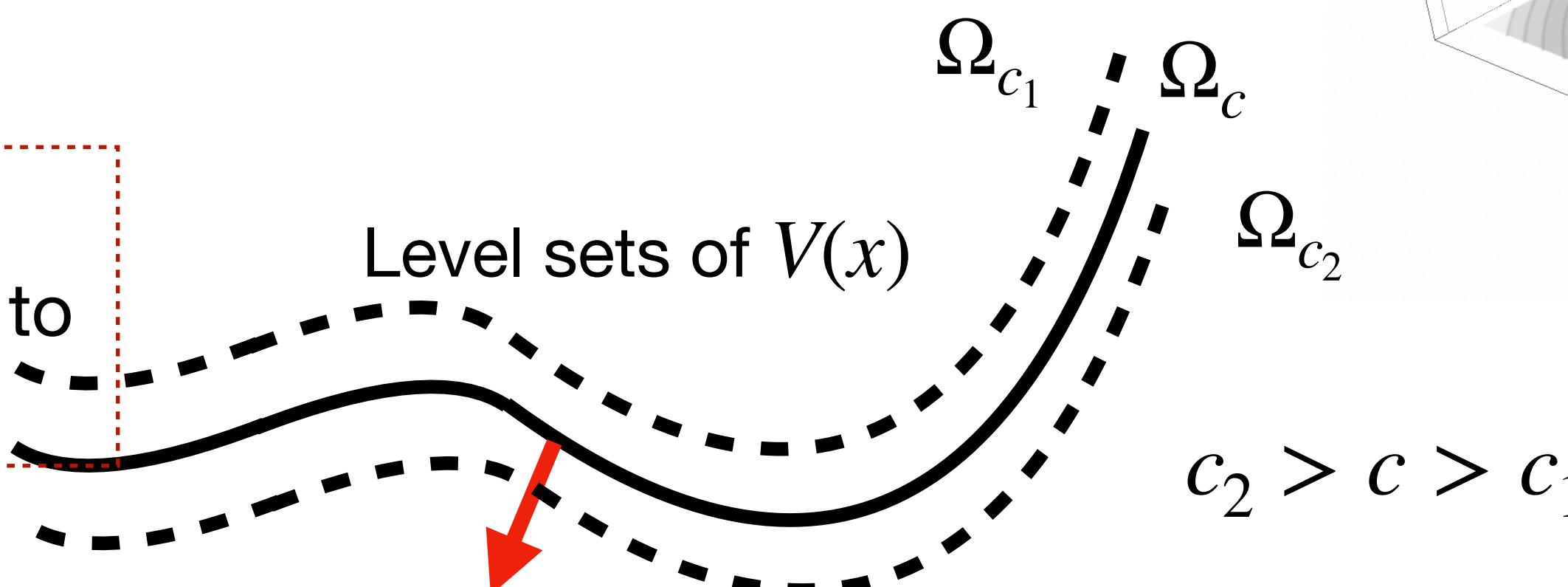


$$\nabla V(x) := \frac{dV(x)}{dx} = \left(\frac{\partial V(x)}{\partial x_1} \quad \frac{\partial V(x)}{\partial x_2} \quad \dots \quad \frac{\partial V(x)}{\partial x_n} \right)$$

↑
nabla of V Jacobian of V

$\nabla V(x)$ always directed toward
the greatest increase in $V(x)$

$\nabla V(x)$, for $x \in \Omega_c$, always
points toward level sets linked to
higher c



The special case of quadratic forms

b

$n \times n$ matrix!

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

Quadratic form

$$\begin{array}{ccc} V : \mathbb{R}^n & \longrightarrow & \mathbb{R} \\ x & \longmapsto & x^T P x \end{array}$$

Definitions

- The matrix P is **positive semi-definite** ($P \geq 0$) if $x^T P x \geq 0$ for all $x \in \mathbb{R}^n$
- The matrix P is **positive definite** ($P > 0$) if $x^T P x > 0$ for all $x \in \mathbb{R}^n \setminus \{0\}$
- The matrix P is **negative semi-definite** ($P \leq 0$) if $x^T P x \leq 0$ for all $x \in \mathbb{R}^n$
- The matrix P is **negative definite** ($P < 0$) if $x^T P x < 0$ for all $x \in \mathbb{R}^n \setminus \{0\}$

Divergence of a quadratic form:

$$\nabla V(x) = 2x^T P$$

P is necessarily
equal to P^T
or symmetric

Remark. In quadratic forms the matrix P can be taken symmetric ($P = P^T$) without loss of generality

Result. A matrix P is positive definite iff all its leading principal minors D_k are positive for all $k = 1, \dots, n$

Definition. D_k is the determinant of the matrix obtained by eliminating the last $n - k$ rows and columns from P

↗ not necessarily symmetric
 Say $P \neq P^T$, and $V = \mathbf{v}^T P \mathbf{v}$

$$P = P_S + P_A$$

$$P_S = \frac{P + P^T}{2} = P_S^T$$

$$P_A = \frac{P - P^T}{2} = -P_A^T$$

Let's embed P in quadratic form

$$V = \mathbf{v}^T P \mathbf{v} = \mathbf{v}^T (P_S + P_A) \mathbf{v} = \underbrace{\mathbf{v}^T P_S \mathbf{v}}$$
 $\underbrace{+ \mathbf{v}^T P_A \mathbf{v}}$

scalar scalar

it's always
 $=0$

$$\underbrace{\mathbf{v}^T P_A^T \mathbf{v}}_{\text{scalar}} = -\underbrace{\mathbf{v}^T P_A \mathbf{v}}_{\text{scalar}} \quad \text{the only scalar that equals to its neg. value is } 0$$

$$(w^T M v)^T = v^T M^T w^T$$

So, $P = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}$ is positive definite

$$D_1 = \det I = 1$$

$$D_2 = \det P = 1$$

$$\frac{d(\mathbf{v}^T P \mathbf{v})}{d\mathbf{v}} = (2\mathbf{v}^T P)$$

The special case of quadratic forms

What's the geometry of $\Omega_c = \{x \in \mathbb{R}^n : x^T Px \leq c\}$?

symmetric matrix not necessarily positive definite

Properties of any symmetric positive definite matrix $P = P^T > 0$

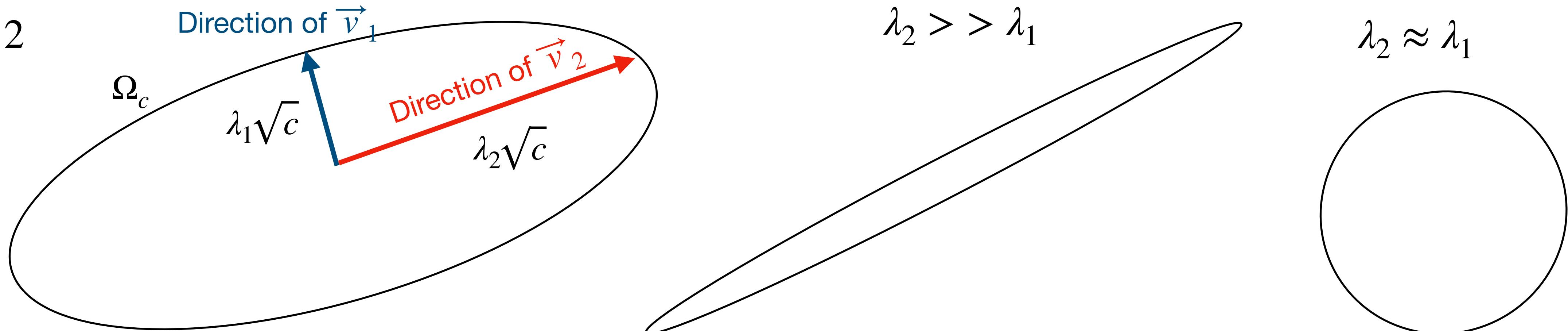
- P is diagonalisable ($a_i = g_i$)
- The eigenvalues λ_i and eigenvectors \vec{v}_i are real, $i = 1, \dots, n$
- $\lambda_i > 0, i = 1, \dots, n$, and for each pair (\vec{v}_i, \vec{v}_j) of eigenvectors $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ (orthogonal)

Cross-product is orthogonal

Eigenvalue is positive

Ω_c are ellipsoids with principal axes direct as the eigenvalues of P and amplitude proportional to the relative eigenvalues

Example $n = 2$



Lyapunov Direct Theorem (C-T systems)

eq. point (maybe at the origin)

Direct Lyapunov Theorem. Let $\bar{x} \in \mathbb{R}^n$ be an equilibrium point for $\dot{x} = f(x)$. Let $V : \mathcal{D} \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ be a positive definite function with respect to $\bar{x} \in \mathcal{D}$ and consider the real-valued function $V'(x) : \mathcal{D} \rightarrow \mathbb{R}$ defined as

$V'(x) := \nabla V(x) f(x)$. The following holds:

semi-negative definite

(a) if $V'(x) \leq 0$ for all $x \in \mathcal{D}$, then \bar{x} is a stable equilibrium point of the system;

(b) if $V'(x) < 0$ for all $x \in \mathcal{D} \setminus \{0\}$, then \bar{x} is an asymptotically stable equilibrium point of the system with a certain domain of attraction \mathcal{A}

negative definite

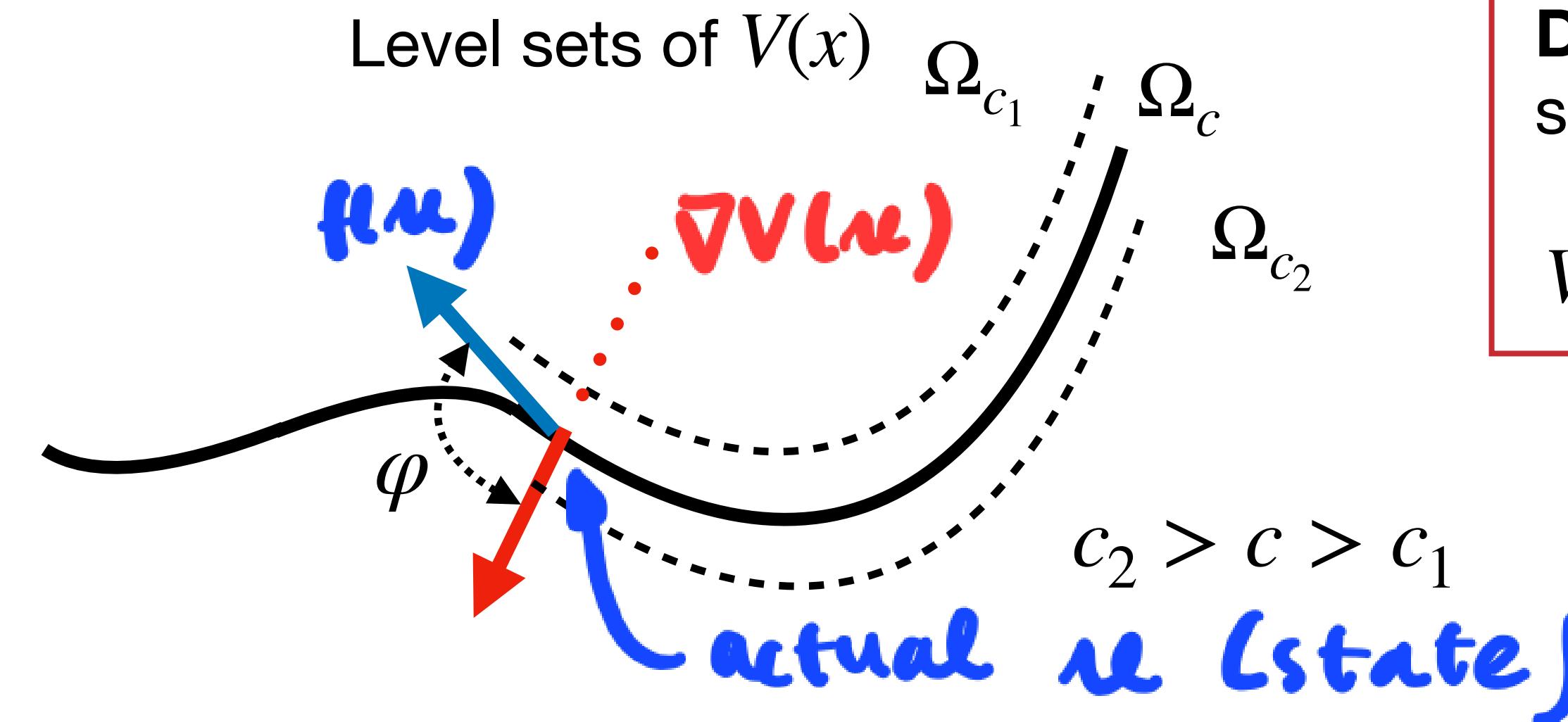
→ divergence

$V'(x) = \nabla V(x) f(x)$ scalar product between the divergence and the vector field

$$\nabla V(x) f(x) < 0 \iff |\varphi| > \frac{\pi}{2}$$

def. negative

$$\nabla V(x) f(x) = 0 \iff \varphi = \pm \frac{\pi}{2}$$

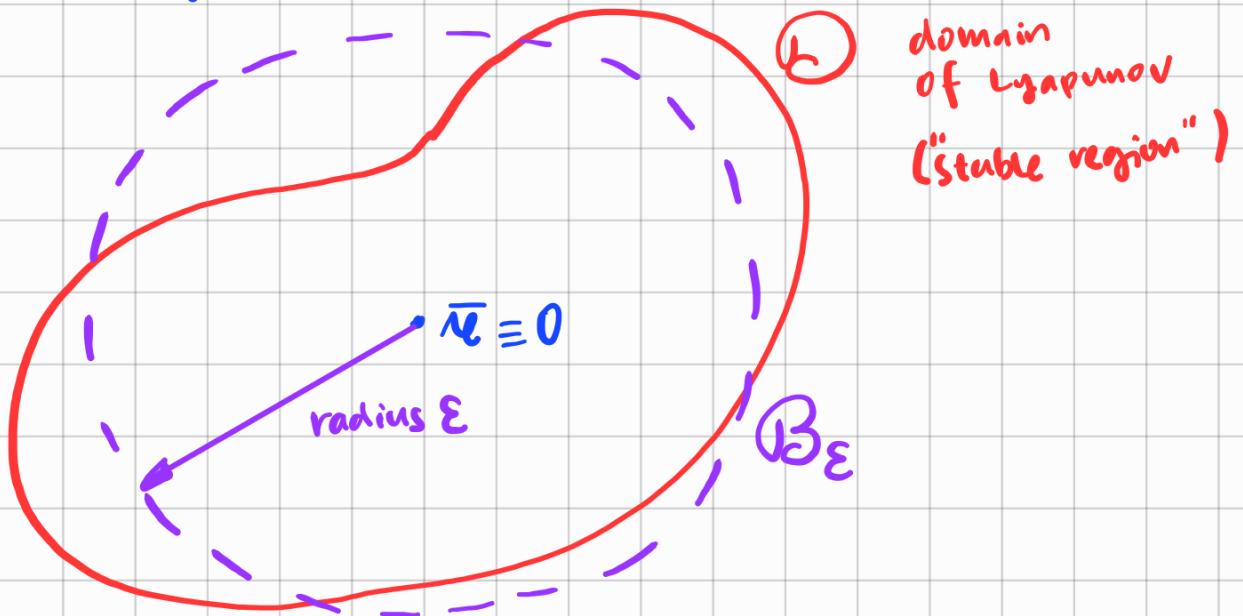


D-T systems: Same statement but with

$$V'(x) := V(f(x)) - V(x)$$

The condition is equivalent to ask that the vector field points inward (tangent) the level set

Say $\bar{x} \equiv \text{Origin}, n = 2$



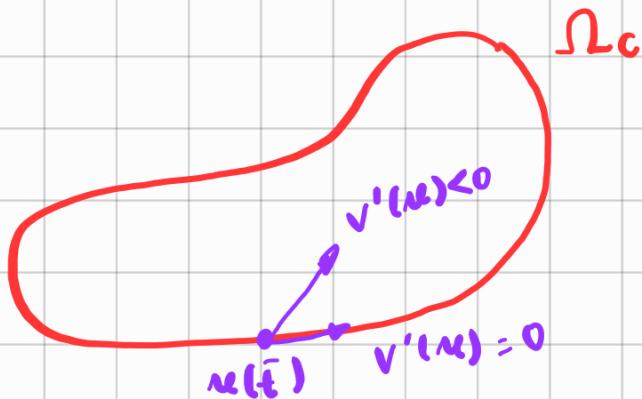
domain
of Lyapunov
(“stable region”)

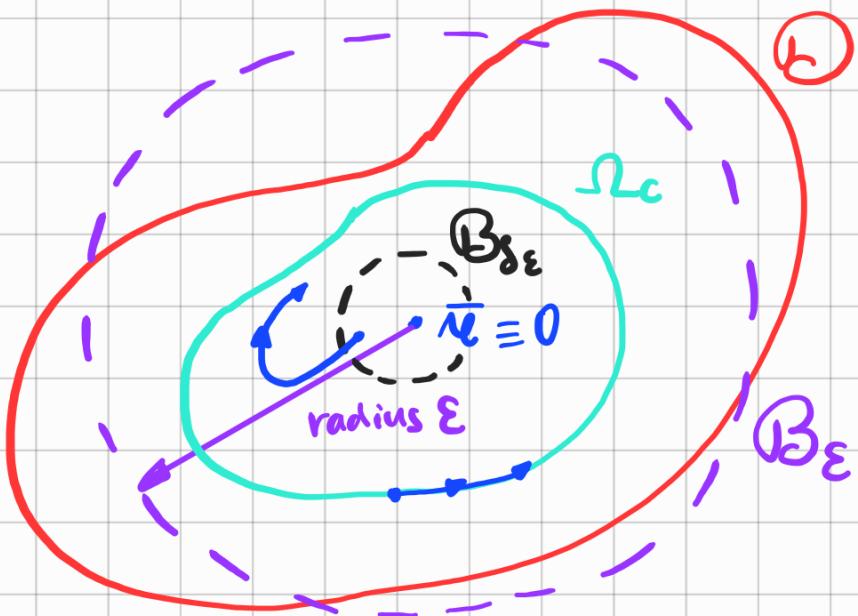
Lyapunov claims :

it implies that for $\forall x \in \Omega$
is “forward invariant” for
 $\ddot{x} = f(x)$

$$\begin{aligned} a.) \quad & V'(x) = \nabla V(x) f(x) \leq 0, \quad \forall x \in \Omega \\ & \Rightarrow \forall \varepsilon > 0 \ \exists \delta_\varepsilon > 0 : \forall \|x(0)\| \leq \delta_\varepsilon \\ & \Rightarrow \|x(t)\| \leq \varepsilon \ \forall t \geq 0. \end{aligned}$$

It implies that $V'(x) \leq 0 \Rightarrow \forall C > 0, x(\bar{t}) \in \Omega_C$,
 $x(t) \in \Omega_C \ \forall t \geq \bar{t}$





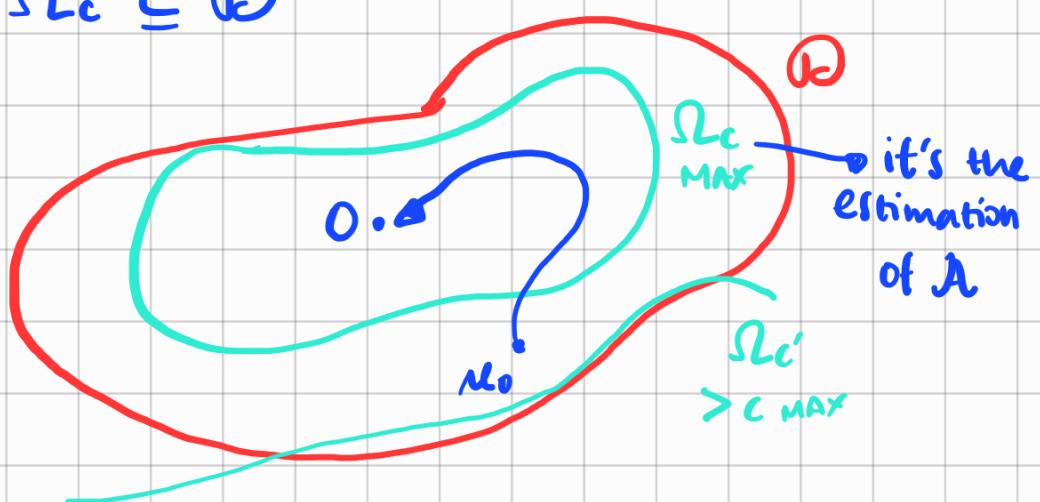
STEP 1 Pick $c > 0 : \Omega_c \subseteq \Omega ; \Omega_c \subseteq B_\varepsilon$

STEP 2 $\delta_\varepsilon > 0 : B_{\delta_\varepsilon} \subseteq \Omega_c$



$V'(x) < 0 \quad \forall x \in \Omega \setminus \{0\}$.

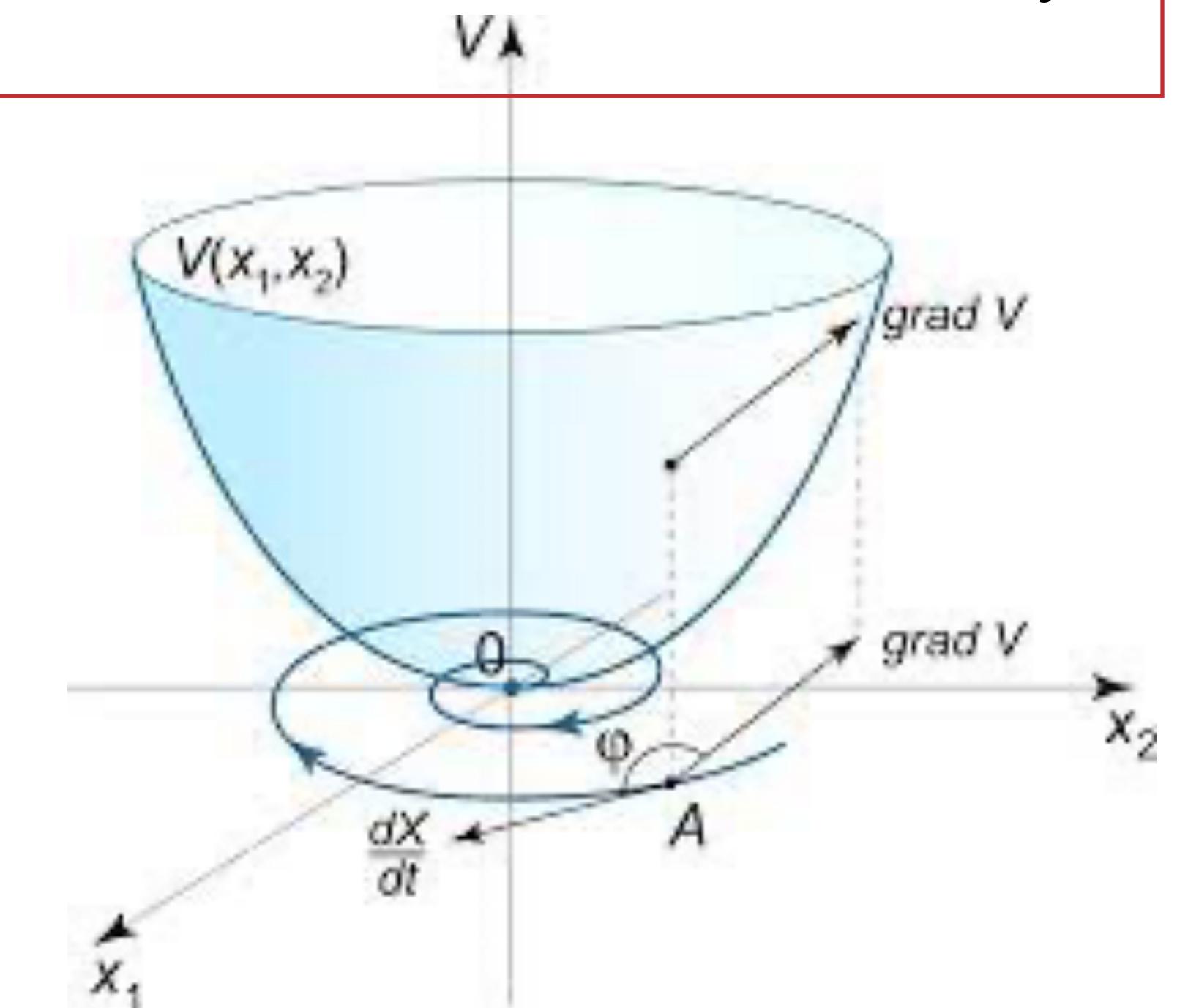
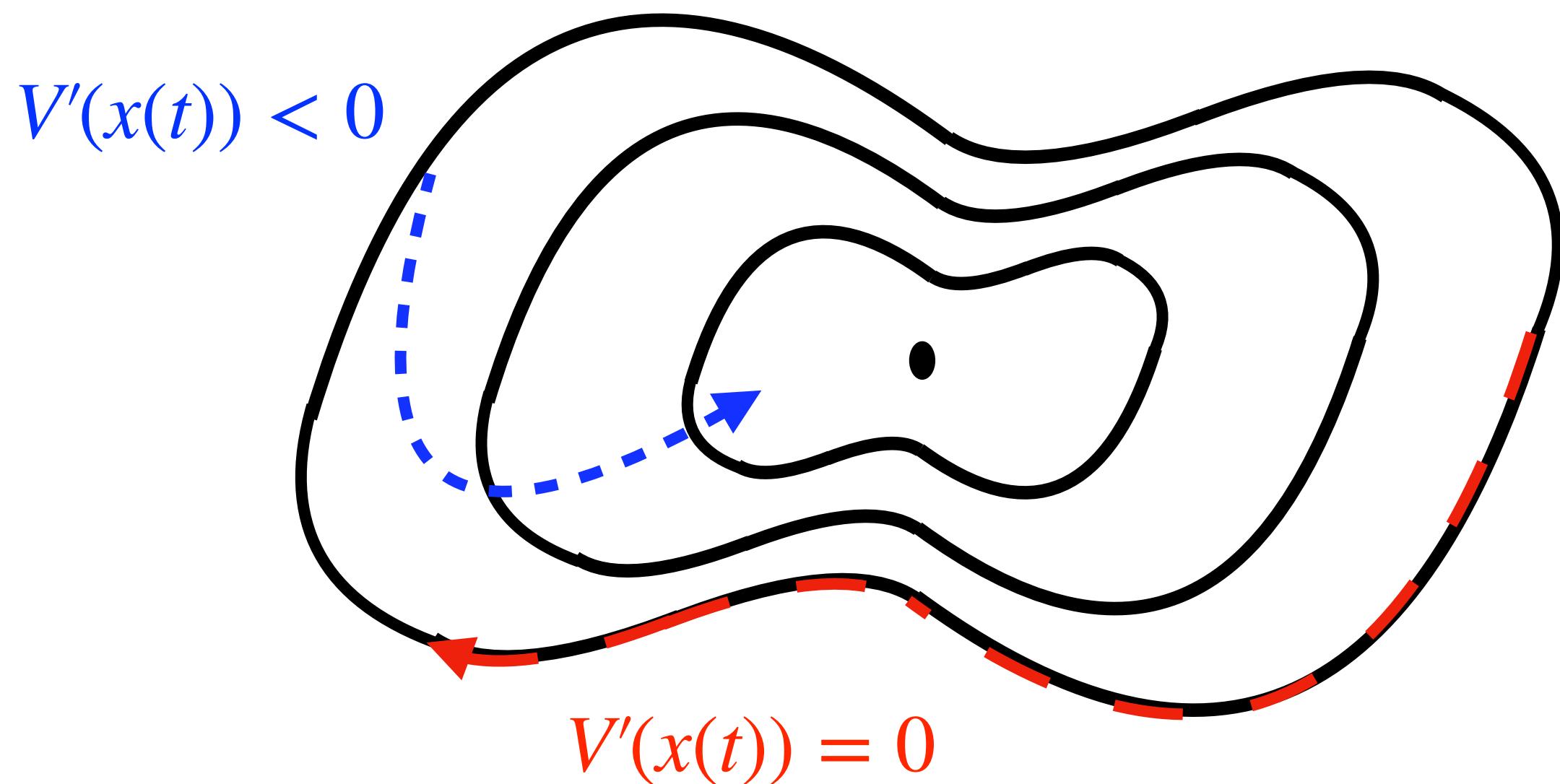
Guess the set of attraction A ? — It is the largest $\Omega_c \subseteq \Omega$



Lyapunov Direct Theorem

$$\frac{dV(x(t))}{dx} = V'(x(t)) = \nabla V(x(t)) \cdot f(x(t)) \text{ derivative of } V(x(t)) \text{ along the solution of } \dot{x}(t) = f(x(t))$$

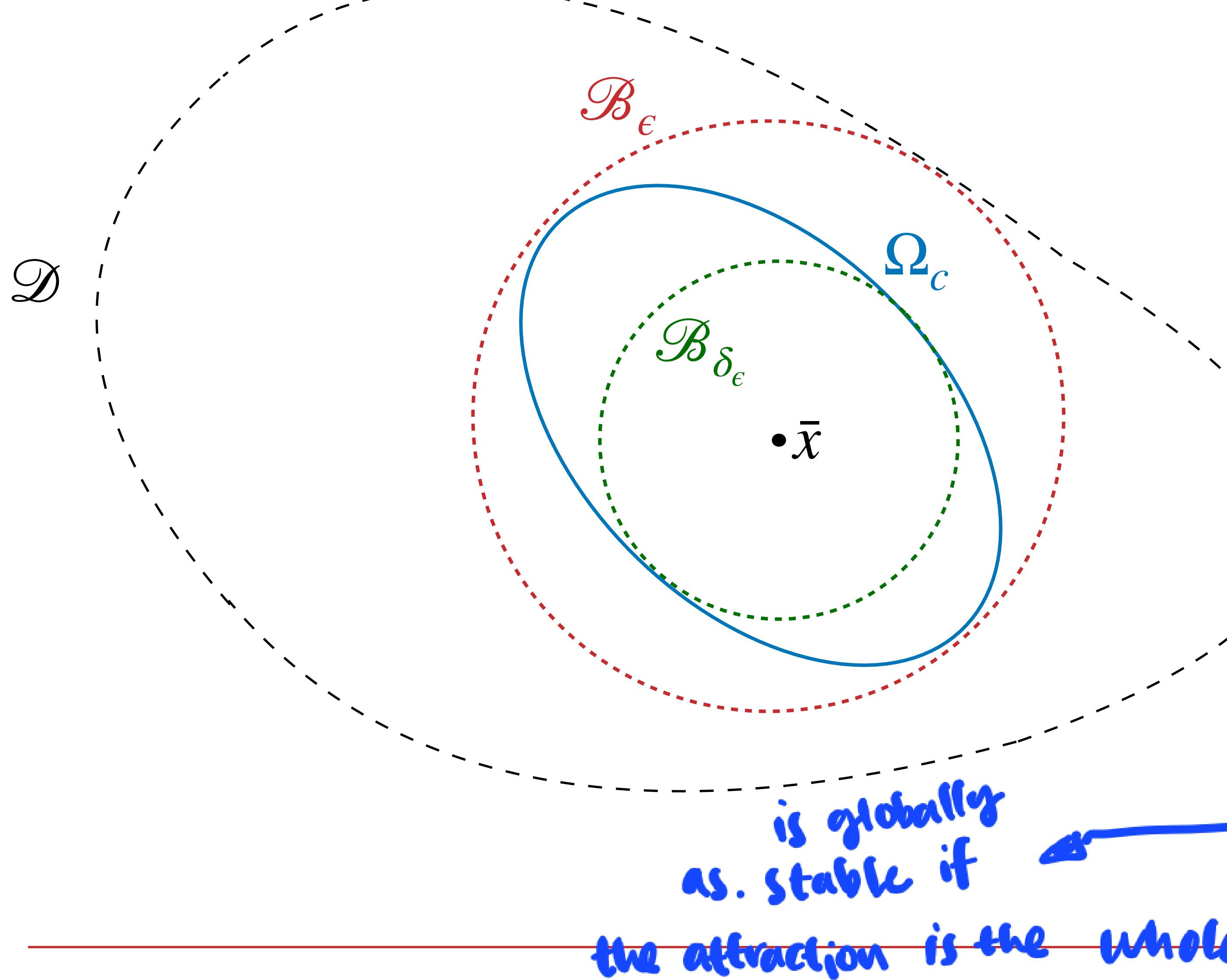
If the condition $V'(x(t)) \leq 0$ is fulfilled then the level sets are “**invariant**” (namely trajectories starting in a level set do not escape it). If $V'(x(t)) = 0$ trajectories run over the boundary of the the level set. Otherwise, they point toward inner level sets.



Same considerations if the system is D-T. The condition $V'(x) := V(f(x)) - V(x) \leq 0$ means that the trajectory jumps into inner level sets (or remain on the actual one if $V'(x) = 0$).

Lyapunov Direct Theorem

Proof.



Remark. In case of asymptotic stability the largest level set of V included in \mathcal{D} is a (conservative) estimation of the domain of attraction \mathcal{A}

Terminology

LAS (Local Asymptotic Stability) if $\mathcal{A} \subset \mathbb{R}^n$

GAS (Global Asymptotic Stability) if $\mathcal{A} = \mathbb{R}^n$

Converse Lyapunov Theorem

The direct Lyapunov theorem presents a sufficient condition for the (asymptotic) stability of an equilibrium point. If a candidate Lyapunov function V (i.e. a smooth function that is positive definite wrt the equilibrium) fails to fulfil $V' \leq 0$ the equilibrium point is not necessarily unstable (simply the V is not a Lyapunov function). However, if an equilibrium point is stable (asymptotically stable) a Lyapunov function fulfilling the above conditions necessarily exists.

Converse Lyapunov Theorem. Let $\bar{x} \in \mathbb{R}^n$ be a stable (asymptotically stable) equilibrium point of $\dot{x} = f(x)$ (with domain of attraction \mathcal{A}). Then there exists a differentiable $V : \mathcal{D} \rightarrow \mathbb{R}$ that is positive definite wrt \bar{x} (with $\mathcal{D} \supset \mathcal{A}$) and fulfilling $V'(x) \leq 0$ ($V'(x) < 0$) for all $x \in \mathcal{D}$

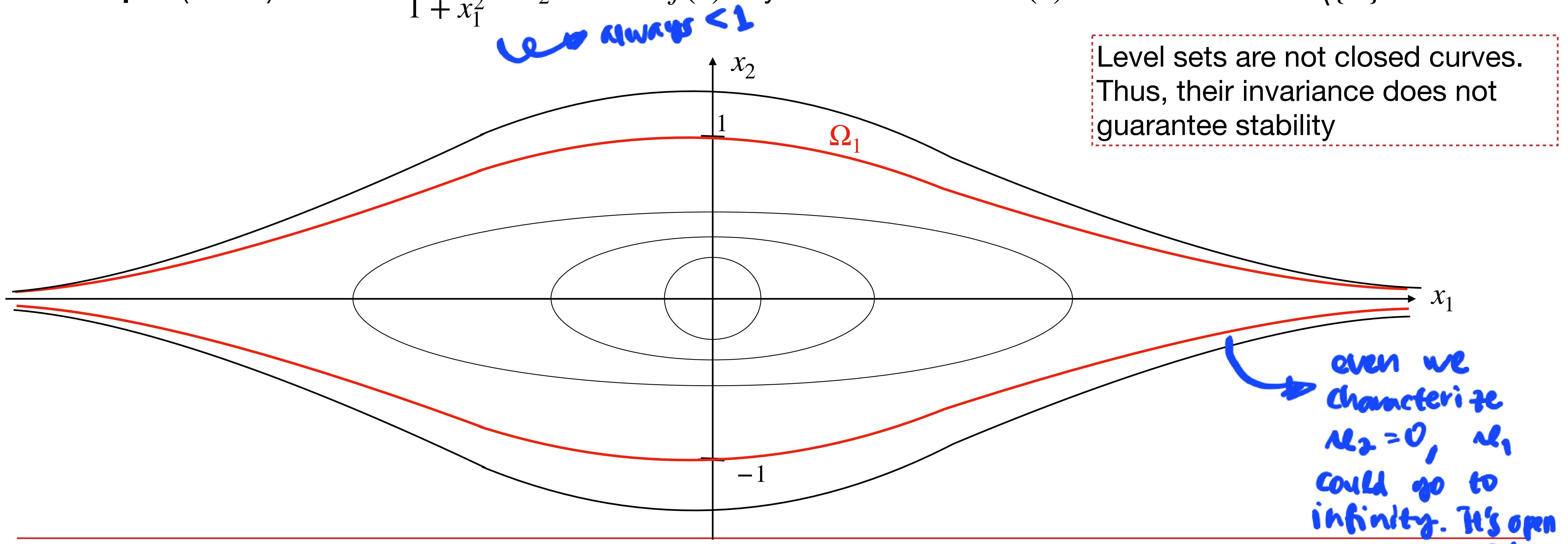
Unfortunately the proof of the theorem is not constructive...

Searching for a Lyapunov function is not easy. Energy-related considerations are good starting points in general.
When you fail to prove that the derivative of a certain V is negative could be either that the equilibrium under study is unstable or simply that the chosen V is wrong

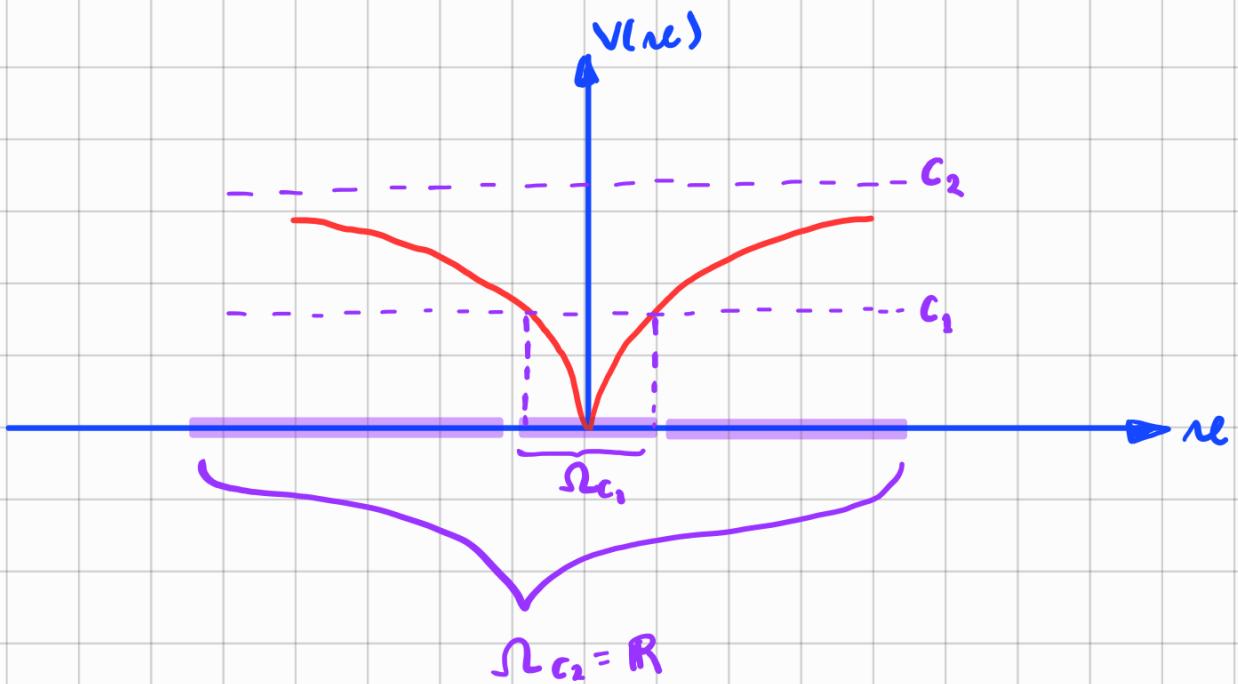
Global Asymptotic Stability

If $\mathcal{D} = \mathbb{R}^n$ and $V'(x) < 0$ for all $x \in \mathbb{R}^n \setminus \{0\}$ $\Rightarrow x = 0$ GAS ? **NO !!!**

Example ($n = 2$) $V = \frac{x_1^2}{1 + x_1^2} + x_2^2$ and $f(x)$ any function so that $V'(x) < 0$ for all $x \in \mathbb{R}^2 \setminus \{0\}$



Suppose we take, $n=1 \rightarrow V(n\epsilon) = \arctan(n\epsilon)$



Global Asymptotic Stability

Direct Lyapunov Theorem (GAS). Let $\bar{x} \in \mathbb{R}^n$ be an equilibrium point for $\dot{x} = f(x)$. Let $V : \mathbb{R}^n \rightarrow \mathbb{R}$ be a positive definite function with respect to \bar{x} which is radially unbounded ($\|x\| \rightarrow \infty \implies V(x) \rightarrow \infty$) and consider the real-valued function $V'(x) : \mathbb{R}^n \rightarrow \mathbb{R}$ defined as $V'(x) := \nabla V(x)f(x)$. Then if $V'(x) < 0$ for all $x \in \mathbb{R}^n \setminus \{0\}$, then \bar{x} is **GAS**

if we push the $\|x\|$ to infinity, $V(\|x\|)$ also goes to infinity

negative definite

$$\begin{Bmatrix} \dot{x}(t) \\ x(t+1) \end{Bmatrix} = Ax(t) \quad x(0) = x_0$$

$n \times n$
 $V(x) = x^T Px$ **Quadratic Lyapunov Function**

Theorem

- The system is **Hurwitz (Schur)** iff there exist $P = P^T > 0$ and $Q = Q^T > 0$ solution of the **Lyapunov matrix equation**

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

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

- If there exists a solution (P, Q) then there exists an infinite number of other solutions one for each $Q = Q^T > 0$. That is, Q is arbitrary
- If the Lyapunov matrix equations are fulfilled with $Q = Q^T \geq 0$ then the system is stable

Proof...

Hurwitz System:

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

quadratic

form

$$V(\mathbf{x}) = \mathbf{x}^T \mathbf{P} \mathbf{x}$$

$$V'(\mathbf{x}) = \underbrace{(2\mathbf{x}^T \mathbf{P})}_{\nabla V(\mathbf{x})} \underbrace{\mathbf{A} \mathbf{x}}_{\mathbf{f}(\mathbf{x})}$$

$$\mathbf{P} = \mathbf{P}^T > 0$$

?

$$= \mathbf{x}^T (2\mathbf{P}\mathbf{A})\mathbf{x} = \mathbf{x}^T (\mathbf{P}\mathbf{A} + \mathbf{A}^T \mathbf{P})\mathbf{x}$$

not necessarily
symmetric

symmetric part

$\frac{2\mathbf{P}\mathbf{A} + 2\mathbf{A}^T \mathbf{P}^T}{2} = \mathbf{P}$, because it's symmetric

$$\frac{2\mathbf{P}\mathbf{A} + 2\mathbf{A}^T \mathbf{P}^T}{2}$$

Suppose that there exists $\exists \mathbf{Q} \geq 0$

$$\mathbf{x}^T (2\mathbf{P}\mathbf{A})\mathbf{x} = \mathbf{x}^T (\mathbf{P}\mathbf{A} + \mathbf{A}^T \mathbf{P})\mathbf{x} = -\mathbf{x}^T \mathbf{Q} \mathbf{x}, \forall \mathbf{x}$$

$$\mathbf{P}\mathbf{A} + \mathbf{A}^T \mathbf{P} = -\mathbf{Q} \quad \begin{matrix} > 0 \\ > 0 \\ \geq 0 \end{matrix}$$

this implies $\mathbf{x} = 0$ is stable

If $\exists \mathbf{Q} > 0$,

$\begin{cases} -\mathbf{x}^T \mathbf{Q} \mathbf{x} < 0, \forall \mathbf{x} \in \mathbb{R}^n \setminus \{0\}, \\ \mathbf{Q} > 0 \end{cases} \Rightarrow \mathbf{x} = 0$ is asymptotically stable

Schur System:

$$\lambda e(t+1) = \lambda e(t)$$

$$v'(e) = v(f(e)) - v(e)$$

$$v(e) = e^T P e$$

$$e^T (A^T P A - P) e = v'(e) = \underbrace{(Ae)^T P (Ae)}_{v(f(e))} - e^T P e$$

④ symmetric

$\Rightarrow = -Q$



If $\exists P = P^T > 0$, $Q = Q^T > 0$, $PA + A^T P = -Q$

↑ ↓ (by Lyapunov)
proof?

A is Hurwitz (all λ_i : Real(λ_i) < 0)

pick $P = \int_0^\infty e^{At} Q e^{As} ds$, Q is any

well-defined

(because A Hurwitz)

Lyapunov Matrix Eq. (LME):

$$\int_0^\infty A^T e^{At} (Q e^{As} + e^{At} Q e^{As}) ds = \int_0^\infty \frac{d}{ds} e^{At} Q e^{As} ds$$

$$\star = \left[e^{A^T s} Q e^{As} \right]_0^\infty = 0 \boxed{-Q}$$

First claim:

P solves the LME for every AQ

$$P^T = \int_0^\infty (e^{As})^T Q^+ (e^{A^T s})^T ds = P, \text{ if } Q \text{ is symmetric}$$

Second claim:

If $Q = Q^T > 0$, then $P = P^T > 0$

$$\begin{aligned} A^T P x &= \int_0^\infty A^T e^T e^{A^T s} Q e^{As} A e ds \\ &\quad \stackrel{= z(s)}{\circledast} \\ &= \int_0^\infty \underbrace{z^T(s) Q z(s) ds}_{>0 \forall s} > 0 \end{aligned}$$

$$\boxed{PA + A^T P = -I}$$

Indirect Lyapunov Theorem

$$\begin{aligned} \dot{x} &= f(x) = f(\bar{x}) + A(x - \bar{x}) + g(x - \bar{x}) \\ &= 0 \text{ if } \bar{x} \text{ is an equilibrium point of } f(x) \end{aligned}$$

Taylor expansion around \bar{x}

Higher order terms in $x - \bar{x}$

$$A = f(x) = \left. \frac{df(x)}{dx} \right|_{x=\bar{x}} = \begin{pmatrix} \left. \frac{\partial f_1(x)}{\partial x_1} \right|_{x=\bar{x}} & \cdots & \left. \frac{\partial f_1(x)}{\partial x_n} \right|_{x=\bar{x}} \\ \vdots & \ddots & \vdots \\ \left. \frac{\partial f_n(x)}{\partial x_1} \right|_{x=\bar{x}} & \cdots & \left. \frac{\partial f_n(x)}{\partial x_n} \right|_{x=\bar{x}} \end{pmatrix}$$

If x ranges in a small neighbourhood of \bar{x} the higher order term are negligible and $\dot{x} = f(x) \approx A(x - \bar{x})$

Question: if A is Hurwitz can we conclude that \bar{x} is LAS for $\dot{x} = f(x)$? Yes!

Similar reasonings if the system is D-T

Indirect Lyapunov Theorem

- Suppose that there exists a $P = P^T > 0$ and $Q = Q^T > 0$ solution of the Lyapunov matrix equation.
Then $x = \bar{x}$ is LAS for $\dot{x} = f(x)$ with a certain domain of attraction ($V = (x - \bar{x})^T P (x - \bar{x})$ is a possible Lyapunov function);
- Suppose that A has at least one eigenvalue with positive real part. Then \bar{x} is unstable for $\dot{x} = f(x)$.

Proof...

$$\dot{x} = f(x) = f(0) + \left[\frac{\partial f}{\partial x} \Big|_{x=0} x + \text{higher order term} \right]$$

\downarrow

$$f(0) = 0$$

\uparrow

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} \Big|_{x=0} & \dots & \frac{\partial f_1}{\partial x_n} \Big|_{x=0} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} \Big|_{x=0} & \dots & \frac{\partial f_n}{\partial x_n} \Big|_{x=0} \end{bmatrix}$$

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

$$\approx x \approx 0$$

$\Rightarrow x = 0$ is locally asymptotically stable
for $\dot{x} = f(x)$

\hookrightarrow is A Hurwitz?

A is Hurwitz $\Rightarrow \exists P = P^T > 0, \begin{cases} PA + A^T P = -Q \\ Q = Q^T > 0 \end{cases}$

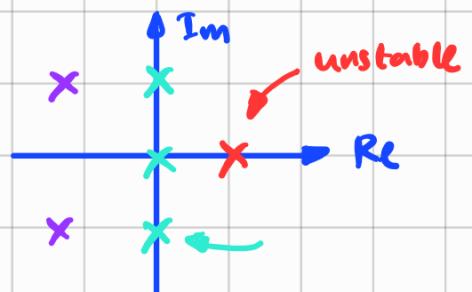
$$V(x) = x^T P x, \quad V'(x) < 0$$

$$\nabla V(x) \cdot Ax$$

\uparrow

$$\frac{\partial V}{\partial x}$$

$\Rightarrow \bar{x} = 0$ is Lyapunov
Asympt. Stable
for $\dot{x} = f(x)$



$$\dot{n} = f(n) = \alpha n^3$$

$$\bar{n} = 0, \quad A = \frac{\partial f}{\partial n} = 3\alpha n^2 \Big|_0 = 0$$

Scenario a.) $\alpha < 0 \Rightarrow \bar{n} = 0$ is globally
as. stable $V(n) = \frac{1}{2}n^2$

$$V'(n) = n \dot{n} = \alpha n^4 < 0$$

b.) $\alpha > 0 \Rightarrow \exists n_0: n(t) \rightarrow \infty$,
unstable

Let's compute $V'(x) = \nabla V(x) f(x)$

$$= \frac{\partial V}{\partial x} (Ax + g(x)) = 2x^T PAx + 2x^T Pg(x)$$

higher order term (HOT)

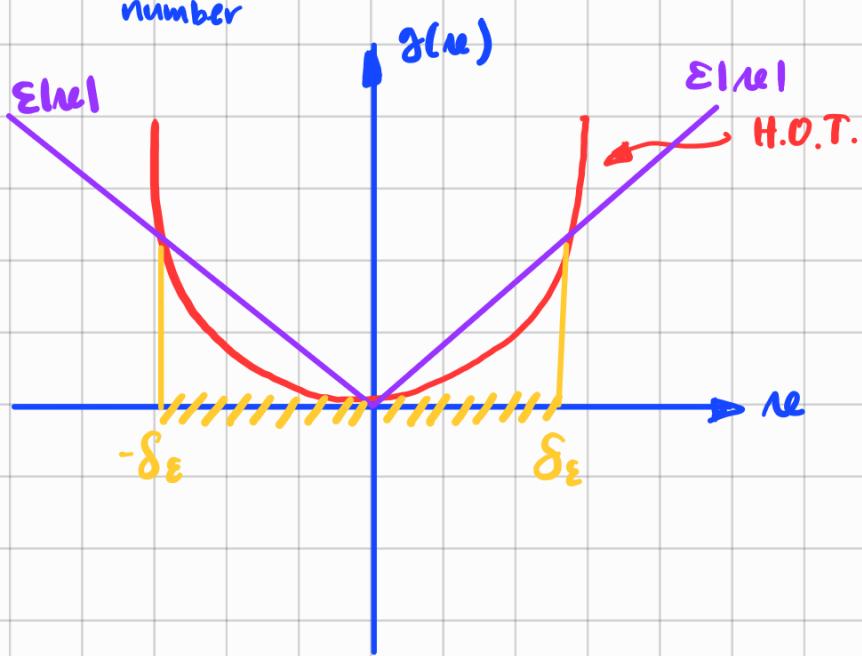
$-x^T Q x < 0$
 $\forall x \in \mathbb{R}^n \setminus \{0\}$

what happens inside HOT?

(?) $g(x)$ H.O.T.

$$\forall \epsilon > 0, \exists \delta_\epsilon > 0, \|g(x)\| \leq \epsilon \|x\| \quad \forall x : \|x\| \leq \delta_\epsilon$$

small number



$$V'(x) \leq -x^T Q x + 2 \|x\| \|P\| \|g(x)\|$$

Pick $\epsilon, \delta_\epsilon$. If $\|x\| \leq \delta_\epsilon$

$$V'(x) \leq -x^T Q x + 2\|P\| \varepsilon \|x\|^2$$

$$\underline{\lambda} \|x\|^2 \leq x^T Q x \leq \bar{\lambda} \|x\|^2$$

min.
eigen
of Q

max
eigen
of Q

$$\text{It implies then, } -x^T Q x \leq -\underline{\lambda} \|x\|^2:$$

$$V(x) \leq -(\underline{\lambda} - 2\|P\| \varepsilon) \|x\|^2$$

Pick $\varepsilon = \frac{\underline{\lambda}}{4\|P\|}$ and δ_ε accordingly :

$$\|x\| \leq \delta_\varepsilon, V'(x) \leq -\frac{\underline{\lambda}}{4} \|x\|^2$$

by
direct Lyapunov
theory

$\bar{x} = 0$ is Lyapunov
assympt. stable for $\dot{x} = f(x)$

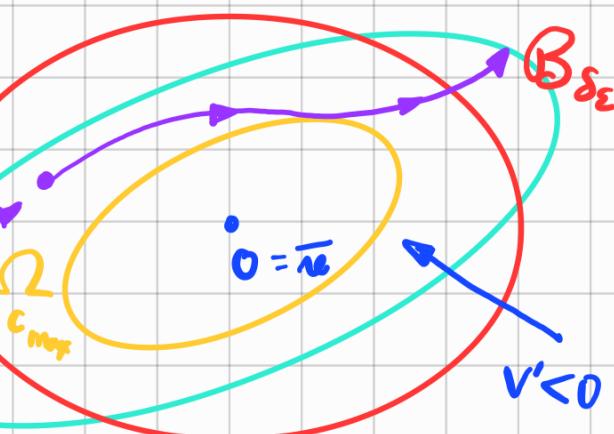
What is the attraction domain A ?

$\rightarrow A$ is
the largest t
level set
of $x^T P x$
contained
in B_{δ_ε}

$$\Omega \in \mathbb{R}^n$$

say it's
the init.
condition

$$\Omega_0$$



Indirect Lyapunov Theorem

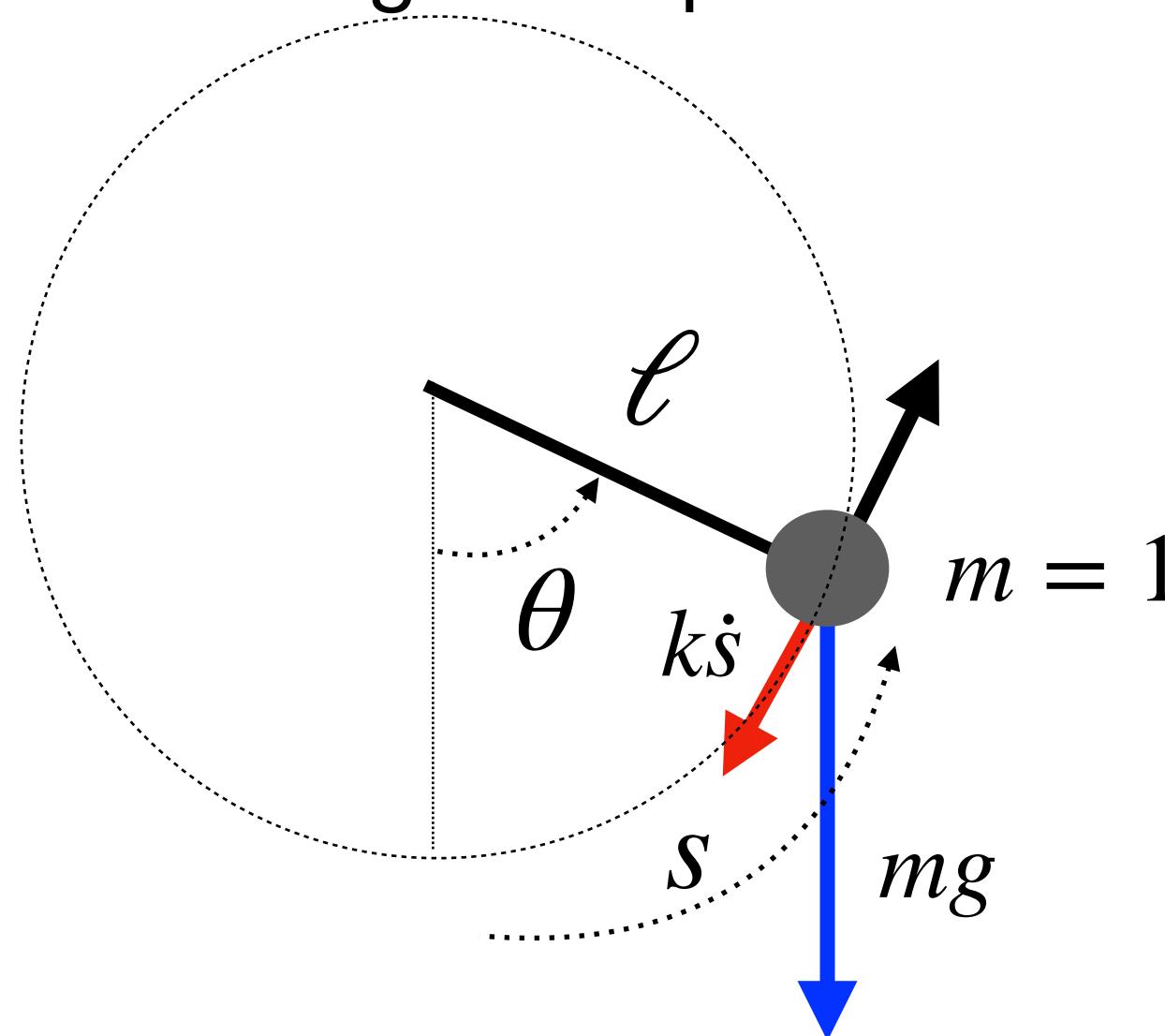
The theorem is not conclusive when A has eigenvalues on the imaginary axis. In this case the nonlinear system is said to be **non hyperbolic** at \bar{x}

In the non-hyperbolic case the equilibrium could be either stable, asymptotically stable or unstable for the nonlinear dynamics. The high-order terms play a role to determine the stability properties of the system

Example. $\dot{x} = \alpha x^3$ with α an arbitrary constant.

Krasovski-La Salle Criterion

Motivating example



$$x_1 := \theta, x_2 := \dot{\theta}$$

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= -\frac{g}{\ell} \sin x_1 - k\ell x_2\end{aligned}$$

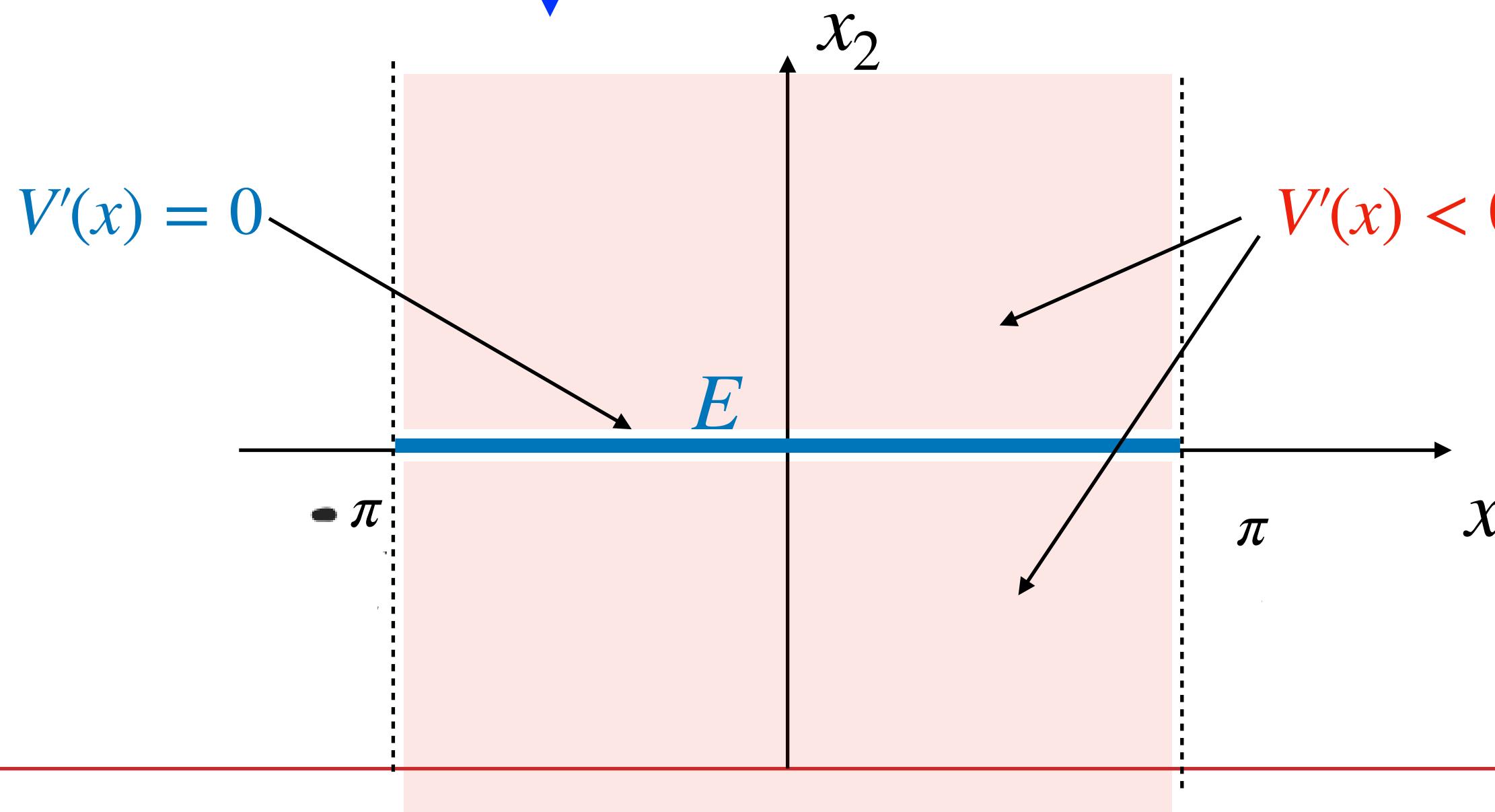
We know that the origin is LAS.
Let's try to prove it by Lyapunov

Candidate Lyapunov Function: Energy

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

Kinetic Energy Potential Energy

Positive definite with respect to $x_1 = x_2 = 0$ with
 $\mathcal{D} = \{(x_1, x_2) : |x_1| < \pi/2\}$



The energy is always decreasing except when $x \in E := \{(x_1, x_2) \in \mathcal{D} : x_2 = 0\}$. Thus $V'(x)$ is only semidefinite negative and we cannot conclude asymptotic stability (just stability).
The set E , however, is only a “transition” set....

$$V' = \frac{\partial V}{\partial \mathbf{r}} \quad f(\mathbf{r}) = -k l \mathbf{r} e_2^2 \leq 0$$

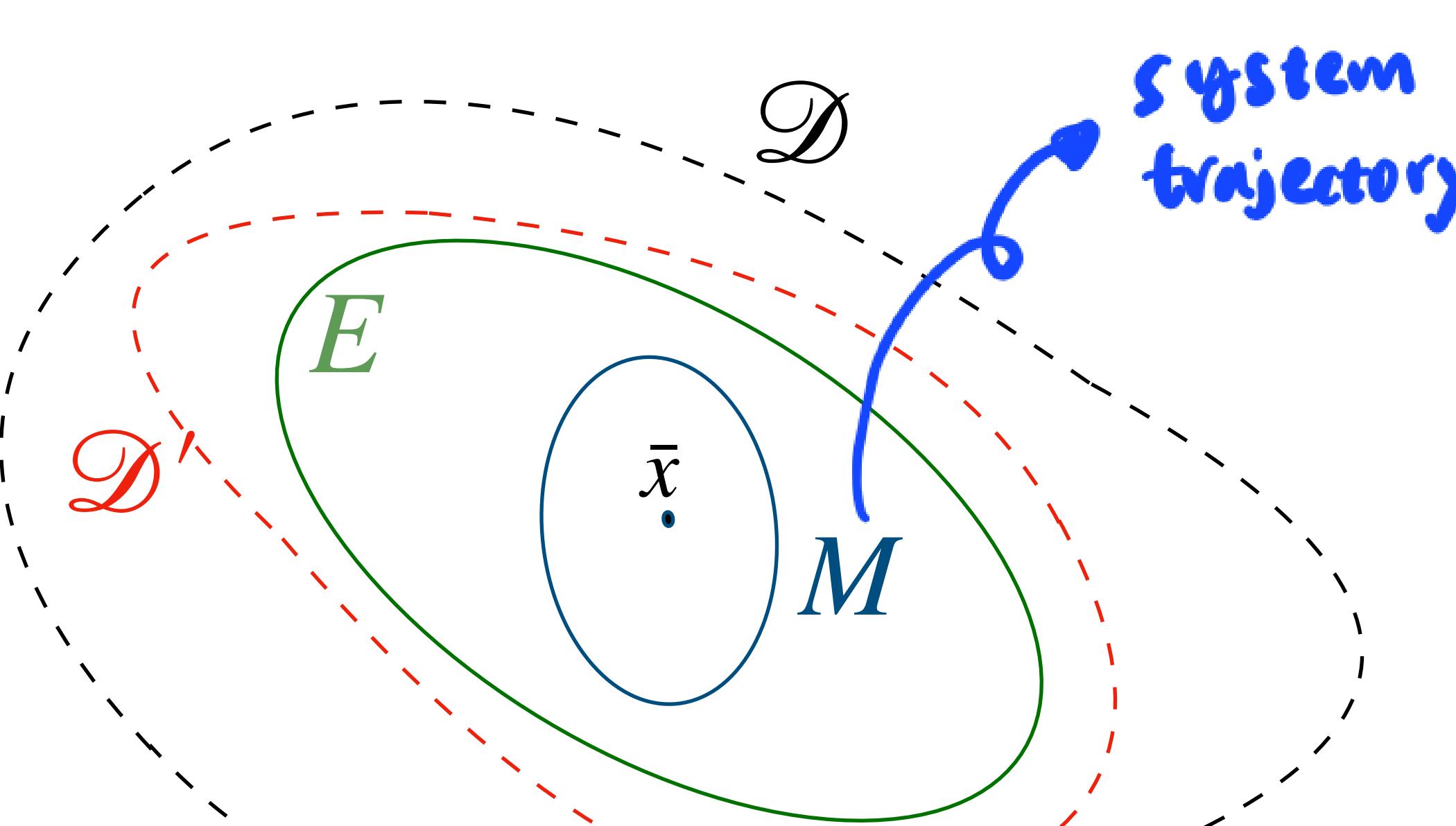
$$\begin{bmatrix} -\frac{\partial}{\partial r} \sin r e_1 & r e_2 \end{bmatrix} \begin{bmatrix} r e_2 \\ -\frac{\partial}{\partial r} \sin r e_1 - k l r e_2 \end{bmatrix}$$

Krasovski-La Salle Criterion

Let $V : \mathcal{D} \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ be a positive definite function relative to \bar{x} , an equilibrium point of the system $\dot{x} = f(x)$.

Suppose that $V'(x) \leq 0$ for all $x \in \mathcal{D}' \subseteq \mathcal{D}$. Let $E \subseteq \mathcal{D}'$ be a set where $V'(x) = 0$ and let M the largest set contained in E which is invariant for the trajectories of the system. Then:

- the equilibrium \bar{x} is stable (classical Direct Lyapunov result)
- the set M is attractive for the trajectories of the system, namely $\lim_{t \rightarrow \infty} \text{dist}(x(t), M) = 0$



Definition. A set M is invariant for $\dot{x} = f(x)$ if, having denoted by $x(t, x_0)$ the trajectory at time t with initial condition x_0 at $t = 0$, then

$$x_0 \in M \implies x(t, x_0) \in M \quad \forall t \in \mathbb{R}$$

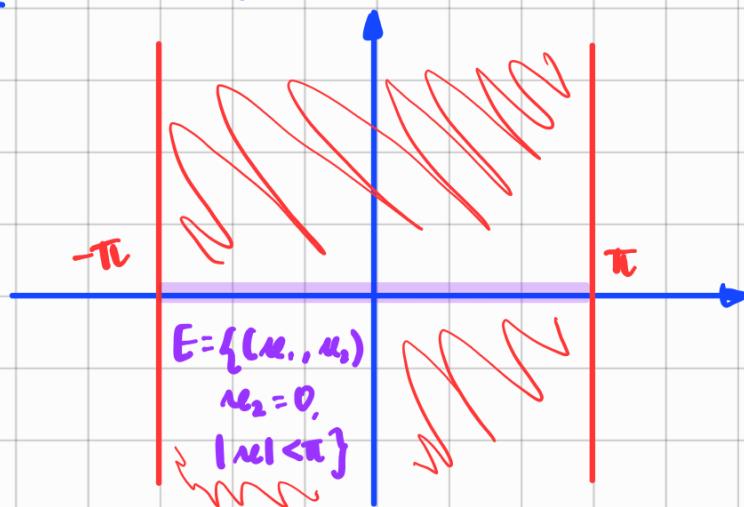
Special case. if $M = \bar{x}$ then the equilibrium point is asymptotically stable

$$V(\alpha) = \frac{1}{2} \alpha_2^2 + \frac{2}{\lambda} (1 - \cos \alpha_1)$$

$$\mathbb{D} = \{(\alpha_1, \alpha_2) : |\alpha_1| < \pi\}$$

$$\mathbb{D}' = \mathbb{D}$$

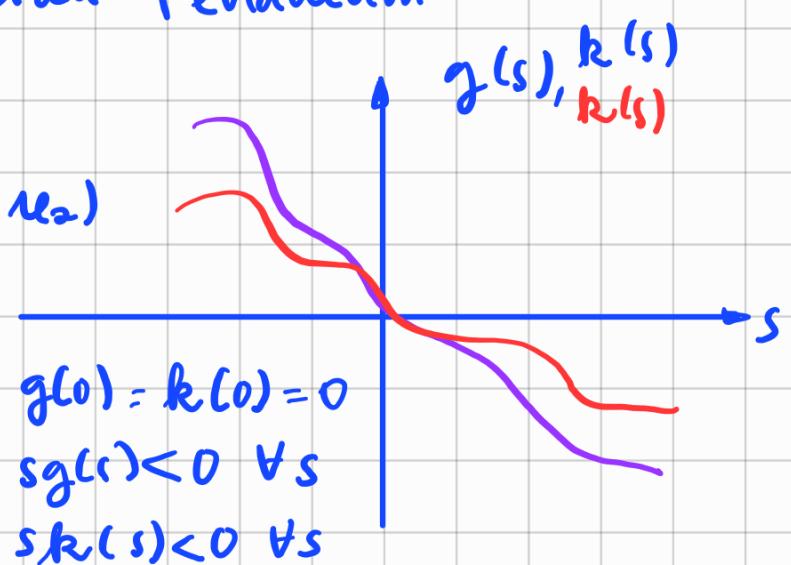
What is $M? \subseteq G$



- Obs # 1 \rightarrow All trajectory of the system evolving in E have $\dot{\alpha}_2(t) = 0$, $\ddot{\alpha}_2(t) = 0$.
- Obs # 2 \rightarrow $\sin(\alpha_1(t)) = 0$
- Obs # 3 \rightarrow The only trajectory of the sys. in E consistent with obs # 1 and # 2 is $(\alpha_1, \alpha_2) \equiv 0$

Example : "Generalized Pendulum"

$$\begin{cases} \dot{\alpha}_1 = \alpha_2 \\ \dot{\alpha}_2 = g(\alpha_1) + k(\alpha_2) \end{cases}$$

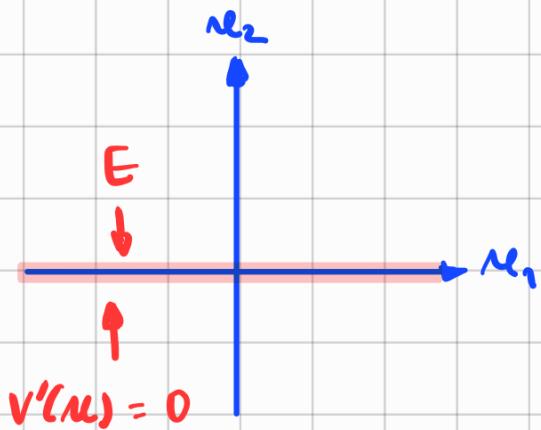


Can we conclude $\bar{u} = (u_1, u_2) = (0, 0)$ is global as. stable?

$$V(u) = \frac{1}{2} u_2^2 - \int_0^{u_1} g(s) ds \quad \xrightarrow{\text{positive def.}} \text{stable}$$

$$V'(u) = \underbrace{(g(u_1) \quad u_2)}_{\nabla V} \begin{pmatrix} u_2 \\ g(u_1) + k u_2 \end{pmatrix} \quad f(u)$$

$$= u_2 k(u_2) \leq 0$$



Examples

- Pendulum case with “variable-gradient method”
- Pendulum case with La Salle
- Generalised Pendulum
- Adaptive (to control direction) control systems

EXAMPLE: Adaptive Control

State equation
 $\dot{y} = ay + u$, $a \in \mathbb{R}$ unknown,

The goal: design $u(y)$: $y \xrightarrow[t \rightarrow \infty]{} 0$

Trivial Scenario:

$\bar{a} > 0$ known

$|a| \leq \bar{a}$

$$u = -ky \quad \longrightarrow \quad \dot{u} = -ky$$

$$k > \bar{a} \quad \dot{k} = \gamma y^2$$

$$\dot{y} = -\underbrace{(k-a)y}_{>0}$$

$$\begin{aligned} \mathbf{x}_1 &= y \\ \mathbf{x}_2 &= k \end{aligned} \quad \mathbf{x} = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{bmatrix}$$

$$\begin{aligned} \dot{\mathbf{x}}_1 &= a\mathbf{x}_1 - \mathbf{x}_1\mathbf{x}_2 \\ \dot{\mathbf{x}}_2 &= \gamma \mathbf{x}_1^2 \end{aligned}$$

$$\nabla V = \begin{bmatrix} 2\mathbf{x}_1 & \frac{2}{\gamma}(\mathbf{x}_2 - b) \end{bmatrix}$$

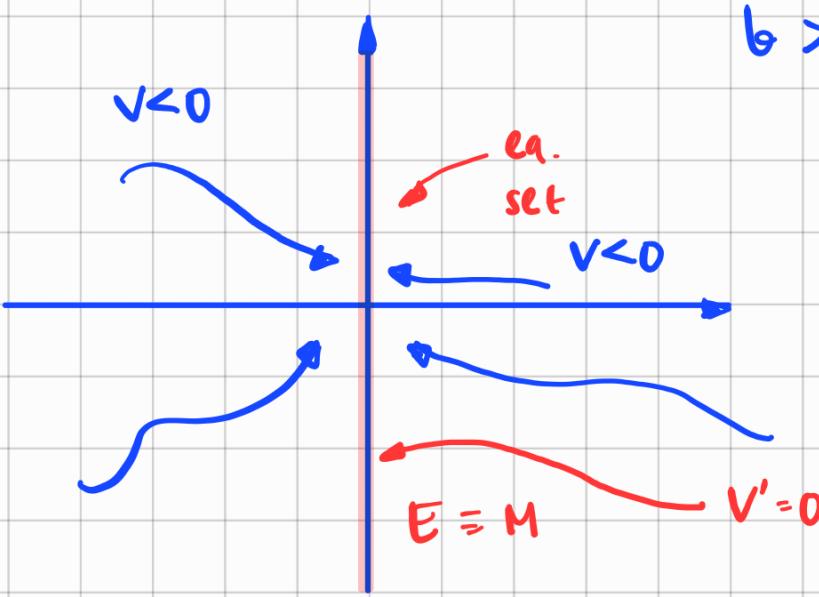
$$V' = \nabla V \cdot f = 2a\mathbf{x}_1^2 - \cancel{2\mathbf{x}_1^2\mathbf{x}_2} + 2(\mathbf{x}_2 - b)\mathbf{x}_1^2$$

$$= \underbrace{-2(b-a)}_{>0} \mathbf{x}_1^2 \leq 0$$

$$V(\mathbf{u}) = u_1^2 + \frac{1}{\delta} (u_2 - b)^2$$

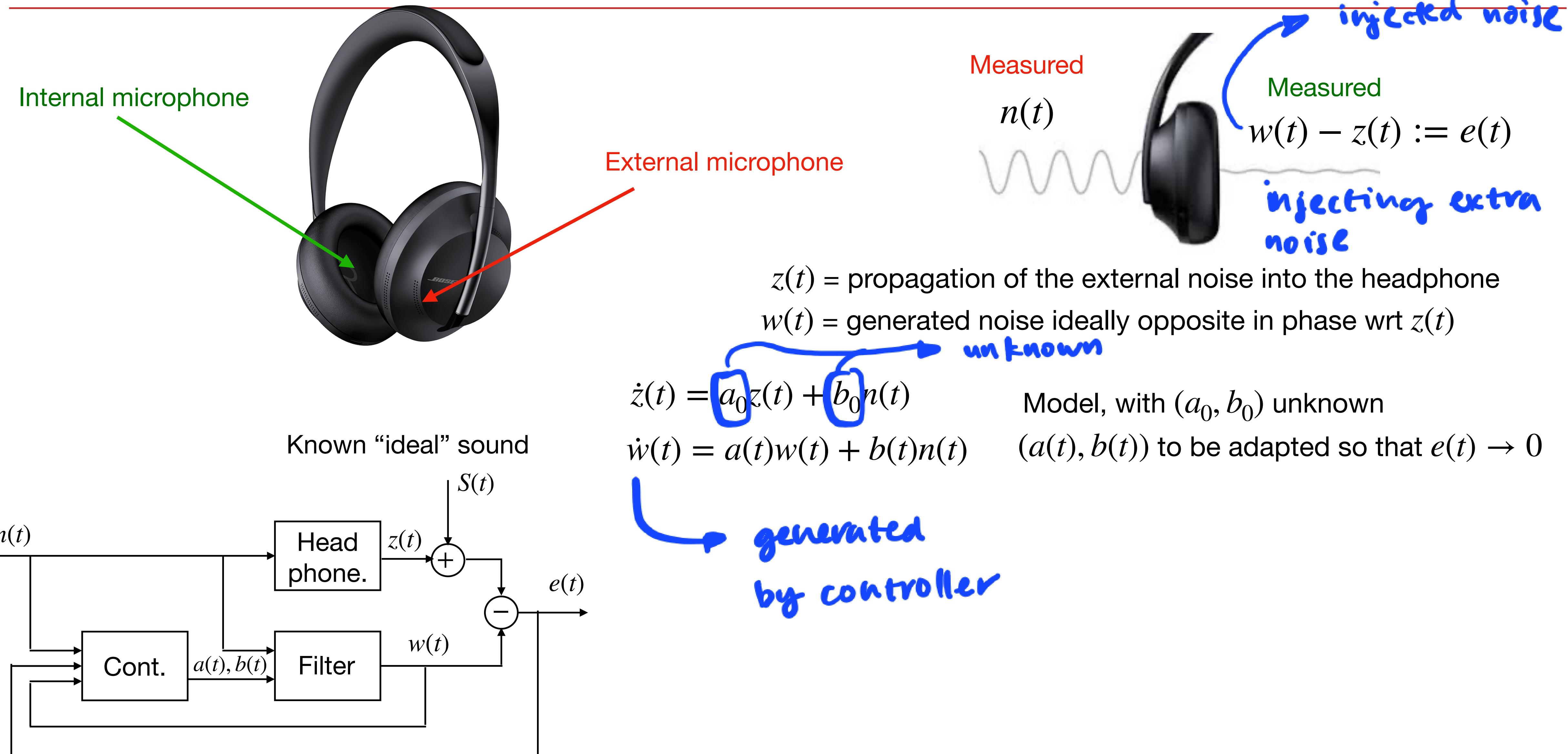
$V(\mathbf{u}) > 0$ with respect to $(u_1, u_2) = (0, b)$

$$b > |a|$$



$$\lim_{t \rightarrow \infty} (u_1, u_2) \xrightarrow{E = M}$$

Example: active noise cancellation



$$e = w - z \quad a_0 < 0$$

$$\dot{e} = aw + bn - az - bn$$

$\downarrow \pm a w \rightarrow$ add and subtract
e

$$(a - a_0)w + a_0(w - z) + (b - b_0)n$$

$$\begin{aligned} \dot{a} &= V_a \\ \dot{b} &= V_b \end{aligned} \quad \left. \begin{aligned} &\text{to be designed} \\ &\text{ } \end{aligned} \right\}$$

$$\mathbf{n} = \begin{bmatrix} n_1 \\ n_2 \\ n_3 \end{bmatrix} = \begin{bmatrix} e \\ a - a_0 \\ b - b_0 \end{bmatrix}$$

$$\dot{n}_1 = n_2 \circled{w(t)} + a_0 \circled{n_1} + n_3 \circled{n(t)}$$

known

error

known
from $e(t)$

$$n_2 = V_a$$

$$n_3 = V_b$$

Lyaپunov Equation:

$$V(n) = \frac{1}{2} (n_1^2 + \frac{1}{\delta a} n_2^2 + \frac{1}{\delta b} n_3^2)$$

good term
since n_1 is always > 0
and $a_0 < 0$

$$\nabla V \cdot f(n) = \cancel{n_1 n_2 w} + (a_0 n_1^2) + n_1 n_3 N + \cancel{\frac{1}{\delta a} n_2 V_a} + \cancel{\frac{1}{\delta b} n_3 V_b} \leq 0$$

$$V_a = -\gamma_a n_1 w$$

$$V_b = -\gamma_b n_3 n$$



$$\dot{V}_a = n_2$$

$$\dot{V}_b = \dot{n}_3$$

M?



E

n₂

n₃

$$\dot{v} = 0$$

n₁

$$0 = n_2 w(t) + n_3 v(t)$$

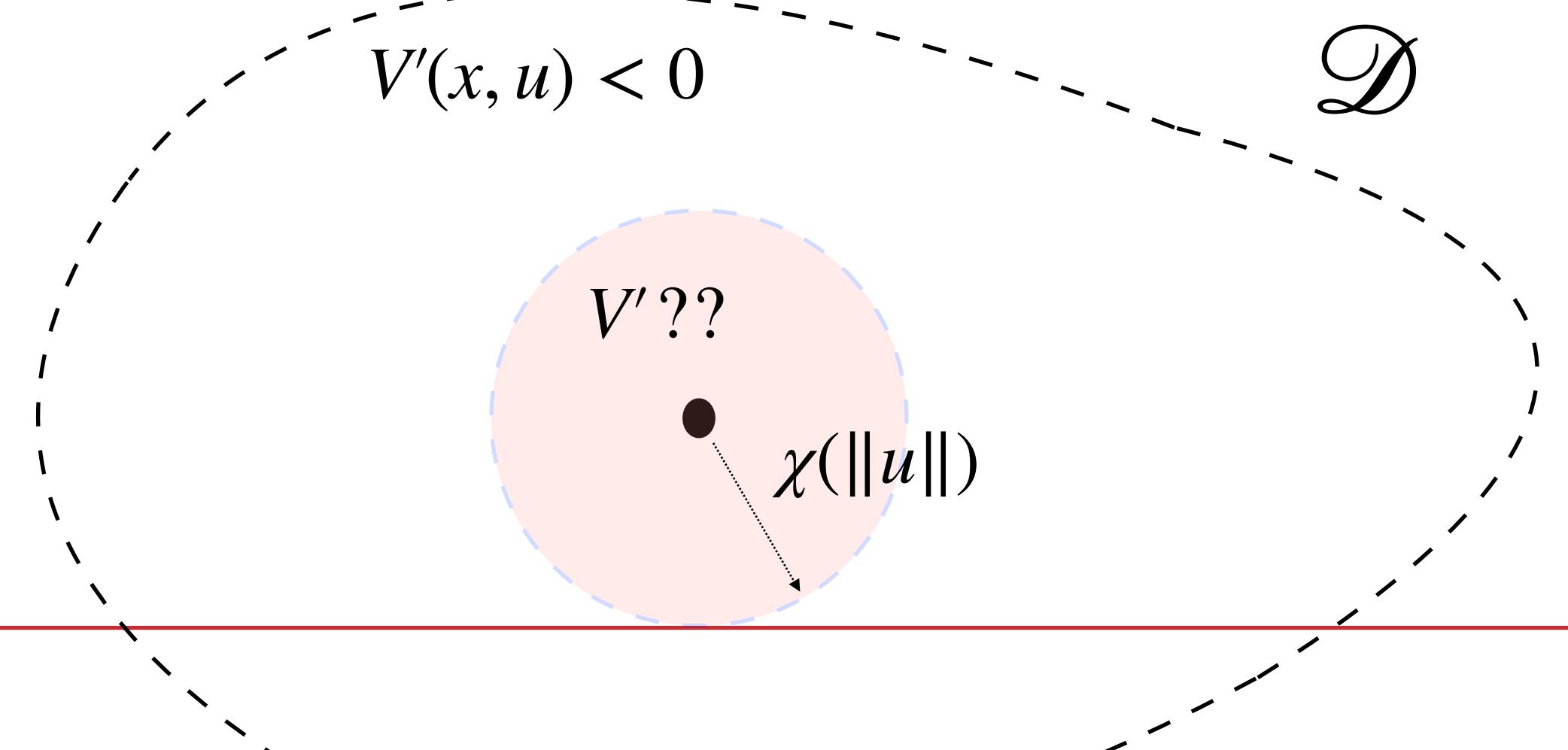
$$\dot{n}_2 = 0, \dot{n}_3 = 0$$

Systems with inputs (and outputs)

When the system has (generic) inputs, the notion of stability gets more involved. As seen, an input can substantially change properties of the input-free system. A well known notion in literature to express a “stability” property in presence of inputs is the one called “**Input-to-State Stability**” (**ISS**). It still refers to an equilibrium point and, in Lyapunov terms, can be expressed as follow.

A system $\dot{x} = f(x, u)$ fulfilling $f(0,0) = 0$ is ISS with respect to $\bar{x} = 0$ if there exist a positive definite real valued function $V : \mathcal{D} \rightarrow \mathbb{R}$ (**ISS Lyapunov function**) and a **strictly increasing** function $\chi : \mathbb{R} \rightarrow \mathbb{R}$ fulfilling

$$\|x\| \geq \chi(\|u\|) \implies V'(x, u) = \nabla V(x) f(x, u) < 0$$

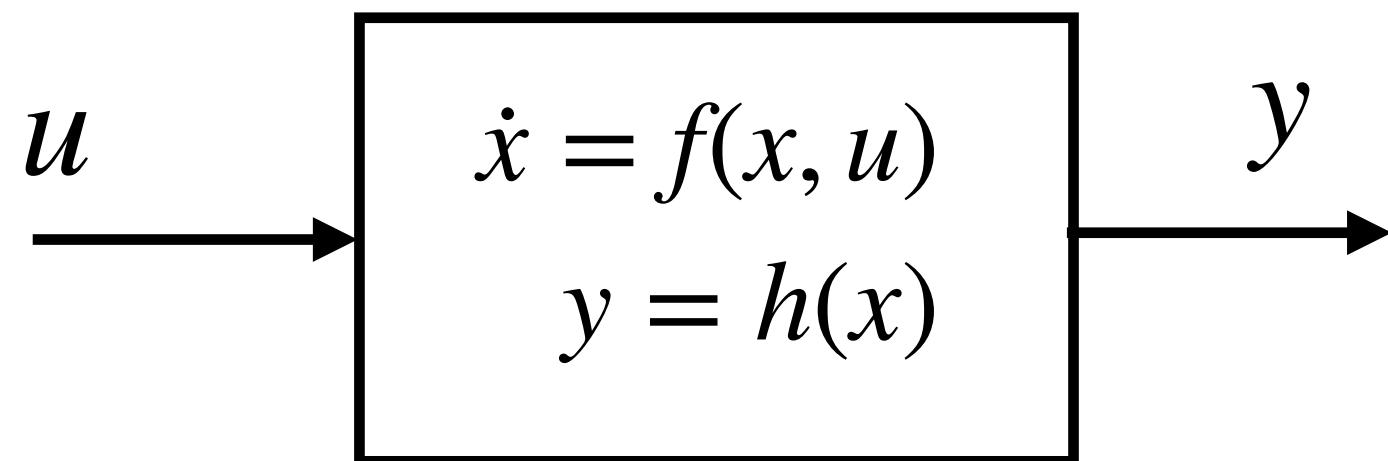


In a nutshell: when the state is “large” with respect to the input then derivative of V is decreasing

When $u = 0$ we obtain the classical stability notion

Systems with inputs (and outputs)

For systems with input and output an important (in the field of engineering) property related to stability and Lyapunov analysis is the one of “**passivity**”



The system aside is passive if there exist a positive definite real valued function $V : \mathcal{D} \rightarrow \mathbb{R}$ (**storage function**) fulfilling

$$\dot{V}(x(t)) = V'(x, u) = \nabla V(x)f(x, u) \leq u^T y$$

Interpretation (circuit perspective):

$$\int_0^T \dot{V}(x(t)) dt = V(x(T)) - V(x(0))$$

Energy stored in the system
in the interval $[0, T]$

$$\leq \int_0^T u(t)^T y(t) dt$$

Energy supplied to the
system in the interval $[0, T]$

The system dissipates energy

