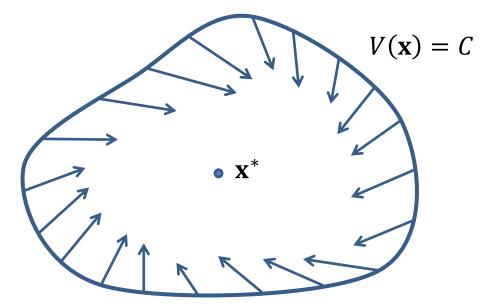
# Lecture 4: Lyapunov Functions

• Suppose there exists a connected orientable region (meaning there is an inside and outside) defined by  $\{x: V(x) \le C\}$  surrounding an equilibrium point  $x^*$  so that all flows crossing the boundary point remain inside the region.

Once inside the region,
 the flow cannot escape



#### Nested boundaries

Consider a nested sequence of surfaces defined by a reducing set of constants. The increasing normal to each surface is given by  $\nabla V$ 

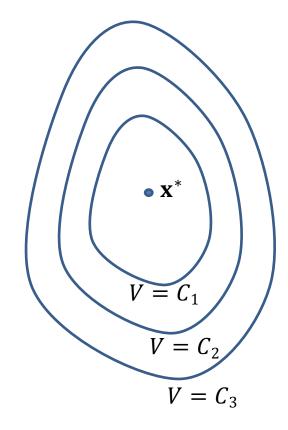
We require the flow to point inwards, i.e.

$$\nabla V \cdot \dot{\mathbf{x}} \leq 0$$
 But  $\dot{\mathbf{x}} = f(\mathbf{x})$ . We thus require  $\nabla V \cdot f(\mathbf{x}) \leq 0$ 

But

$$\frac{dV}{dt} = \sum \frac{\delta V}{\delta x_i} \frac{dx_i}{dt} = \underline{\nabla} V. \dot{\mathbf{x}} \le 0$$

so flows go downhill and end up at the bottom



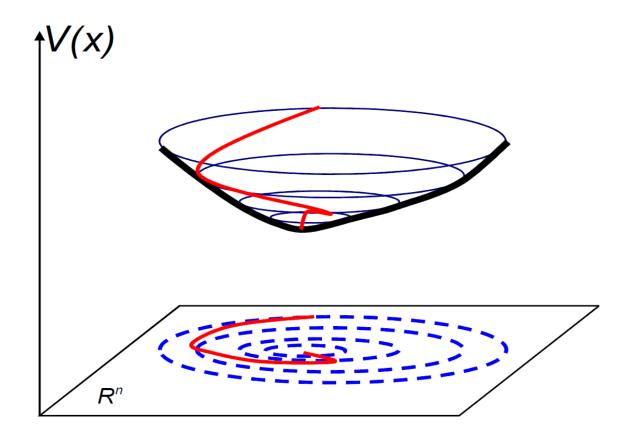
# Lyapunov's Theorem

- Let  $\mathbf{x}^*$  be an equilibrium point of  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$ , i.e.  $\mathbf{f}(\mathbf{x}^*) = 0$ . Let D be an open set surrounding  $\mathbf{x}^*$  and let  $V(\mathbf{x}): D \to \mathbb{R}$  be a continuously differentiable function on D such that
  - 1.  $V(\mathbf{x}^*) = 0$  and  $V(\mathbf{x}) > 0$  for all  $\mathbf{x} \neq \mathbf{x}^*$
  - 2.  $\dot{V}(\mathbf{x}) = \nabla V \cdot \mathbf{f}(\mathbf{x}) \leq 0$

then  $\mathbf{x}^*$  is **stable**. If, in addition

- 3.  $\dot{V}(\mathbf{x}) < 0$  for all  $\mathbf{x} \neq \mathbf{x}^*$
- then  $\mathbf{x}^*$  is **asymptotically stable**
- $V(\mathbf{x})$  is called a Lyapunov function
- If  $\lim_{\|\mathbf{x}\|\to\infty}V(\mathbf{x})=\infty$  and  $D=\mathbb{R}^n$  then  $\mathbf{x}^*$  is globally asymptotically stable

#### Illustration



 $V(\mathbf{x})$  decreases along solution trajectories

### Example 1

Consider the dynamical system

$$\dot{x} = y$$
  
$$\dot{y} = -x + \epsilon x^2 y$$

Equilibrium: (0,0) has Jacobian:  $\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$  with eigenvalues:  $\lambda = \pm j$ 

so Hartman-Grobman doesn't apply

Let 
$$V(x,y) = \frac{1}{2}(x^2 + y^2)$$
,
$$\frac{dV}{dt} = \underline{\nabla}V \cdot \dot{\mathbf{x}} = x\dot{x} + y\dot{y} = xy - xy + \epsilon x^2y^2 = \epsilon x^2y^2$$

If  $\epsilon < 0$  then (0,0) is stable

#### Example 2

$$\dot{x}_1 = -2x_2 + x_2x_3 
\dot{x}_2 = x_1 - x_1x_3 
\dot{x}_3 = x_1x_2$$

• Equilibrium point: (0,0,0) is a linear centre. Let

$$V(\mathbf{x}) = \frac{1}{2}(c_1x_1^2 + c_2x_2^2 + c_3x_3^2)$$

Then

$$\dot{V} = \nabla V \cdot \dot{\mathbf{x}} = c_1 x_1 (-2x_2 + x_2 x_3) + c_2 x_2 (x_1 - x_1 x_3) + c_3 x_3 x_1 x_2$$
$$= (c_1 - c_2 + c_3) x_1 x_2 x_3 + (-2c_1 + c_2) x_1 x_2$$

- Choose  $c_2=2c_1>0$  and  $c_3=c_1$ , then  $\dot{V}=0$ , so equilibrium is stable
- $\dot{V} = 0$  on  $V(\mathbf{x}) = \frac{1}{2}(x_1^2 + 2x_2^2 + x_3^2)$  so  $\mathbf{x}(t)$  lies on  $V(\mathbf{x}) = \text{const}$

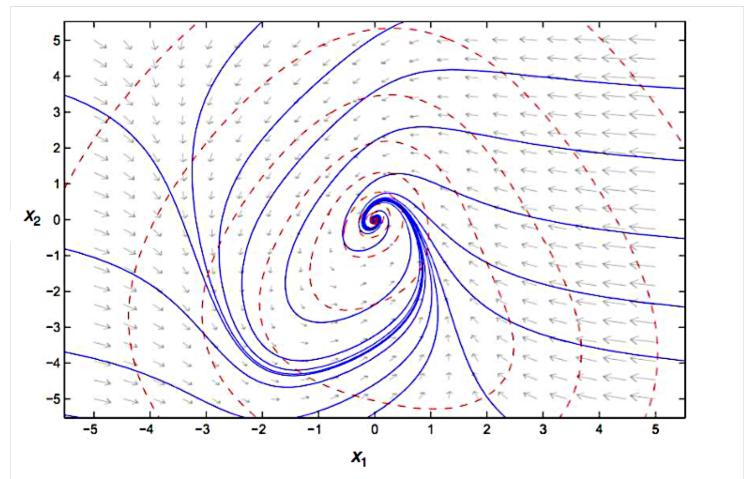
# Jet Engine Example

$$\dot{x}_1 = -x_2 + 1.5x_1^2 - 0.5x_1^3$$

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

- Equilibrium point: (0,0) has Jacobian  $\begin{bmatrix} 0 & -1 \\ 3 & -1 \end{bmatrix}$  with  $\lambda = \frac{-1 \pm j\sqrt{11}}{2}$  so is a linear stable focus.
- Hartman-Grobman theorem states that the non-linear system is stable (but only close to the origin)
- Lyapunov functions can extend this result globally using specially constructed functions – see the lecture notes

# Jet Engine Example



Level curves of the Lyapunov function showing global stability of the Jet engine model

# Vector fields possessing an integral

- Consider the flow associated with the solution of  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$  as a vector field
- This is said to have an integral  $I(\mathbf{x})$  (a scalar function) if

$$\frac{dI(\mathbf{x})}{dt} = \frac{\partial I(\mathbf{x})}{\partial \mathbf{x}} \dot{\mathbf{x}} = \frac{\partial I(\mathbf{x})}{\partial \mathbf{x}} \mathbf{f}(\mathbf{x}) = 0$$

- $\frac{\partial I(\mathbf{x})}{\partial \mathbf{x}}$  is the gradient vector of  $I(\mathbf{x})$
- $I(\mathbf{x})$  defines level sets which contain the flow

### Pendulum example

$$\dot{q} = p$$

$$\dot{p} = -\frac{g}{l}\sin q$$

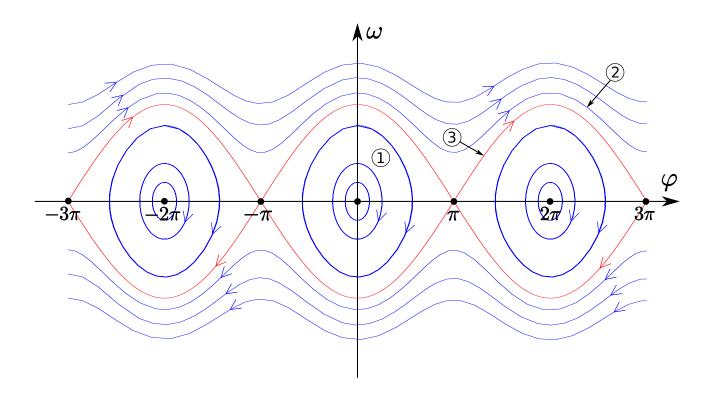
The total stored energy E is conserved

$$E = \frac{1}{2}p^2 - \frac{g}{l}\cos q$$

i.e.

$$\frac{dE}{dt} = p\dot{p} + \dot{q}\frac{g}{l}\sin q = 0$$

# Pendulum example



Phase plane of pendulum and level sets of constant energy

# Duffing Oscillator for $\delta$ =0

$$\dot{x} = y$$

$$\dot{y} = x - x^3$$

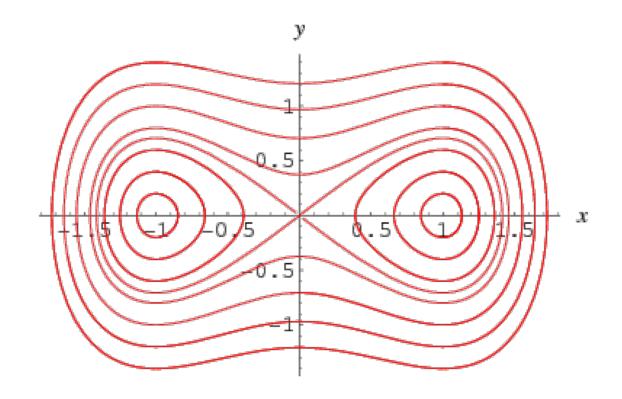
We require

$$\frac{dI}{dt} = \frac{\partial I}{\partial x}\dot{x} + \frac{\partial I}{\partial y}\dot{y} = 0$$
$$\frac{\partial I}{\partial x}y + \frac{\partial I}{\partial y}(x - x^3) = 0$$

So, for example

$$I = \frac{y^2}{2} - \frac{x^2}{2} + \frac{x^4}{4}$$

# Duffing Oscillator for $\delta$ =0



Level sets of  $I(\mathbf{x})$  in the phase plane of the duffing oscillator

# Hamiltonian systems

Hamiltonian systems have vector fields that possess an integral

**Definition**: Systems of the form

$$\dot{\mathbf{p}} = \mathbf{f}(\mathbf{p}, \mathbf{q})$$
  
 $\dot{\mathbf{q}} = \mathbf{g}(\mathbf{p}, \mathbf{q})$ 

such that

$$f(\mathbf{p}, \mathbf{q}) = \partial H(\mathbf{p}, \mathbf{q})/\partial \mathbf{q}, \qquad g(\mathbf{p}, \mathbf{q}) = -\partial H(\mathbf{p}, \mathbf{q})/\partial \mathbf{p}$$

are called Hamiltonian Systems.

- $\mathbf{p}$  and  $\mathbf{q}$  are real vectors with n elements
- H is a twice differentiable function called the Hamiltonian
- **q** is the vector of generalised positions, **p** the vector of generalised momenta
- All Hamiltonian systems are conservative by construction

# More on Hamiltonian systems

- If  $(\mathbf{p}^*, \mathbf{q}^*)$  is an equilibrium and  $H(\mathbf{p}, \mathbf{q}) > 0$  in a region surrounding the equilibrium, then the equilibrium is stable
- A Newtonian system of the form  $\ddot{x} = f(x)$  can be written as a Hamiltonian system by summing the potential energy and kinetic energy

$$\dot{x} = v$$

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

$$H(x, y) = \frac{v^2}{2} - \int_{x_0}^{x} f(s) ds$$

# **Gradient Systems**

**Definition**: Let  $V(\mathbf{x})$  be a twice differentiable function in a region  $D \subseteq \mathbb{R}^n$ . The system

$$\dot{x}_i = -\frac{\partial V}{\partial x_i}$$

is called a **gradient** system.

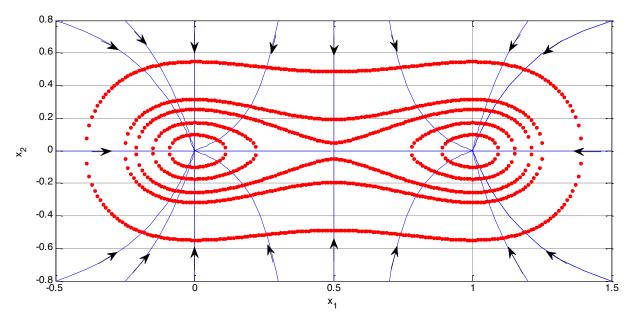
- Equilibrium points are the critical points of V. Away from critical points the trajectories are orthogonal to the level sets of V.
- If  $\mathbf{x}^*$  is a strict local minimum of V then  $V(\mathbf{x}) V(\mathbf{x}^*)$  is a Lyapunov function for  $\mathbf{x}^*$ , showing that  $\mathbf{x}^*$  is asymptotically stable. If  $\mathbf{x}^*$  is a strict local maximum, then the equilibrium is unstable.

# **Example Gradient System**

$$\dot{x} = -4x(x-1)(x-0.5)$$
$$\dot{y} = -2y$$

Has

$$V(x,y) = x^2(x-1)^2 + y^2$$



# Relationship between Gradient and Hamiltonian Systems

The system

$$\dot{x} = f(x, y) = \frac{\partial H}{\partial y}$$

$$\dot{y} = g(x, y) = -\frac{\partial H}{\partial x}$$

is orthogonal to

$$\dot{x} = g(x, y)$$

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

• They have the same equilibria, centres map to nodes, saddles to saddles and foci to foci.