

# Lecture 12: Modeling Physics

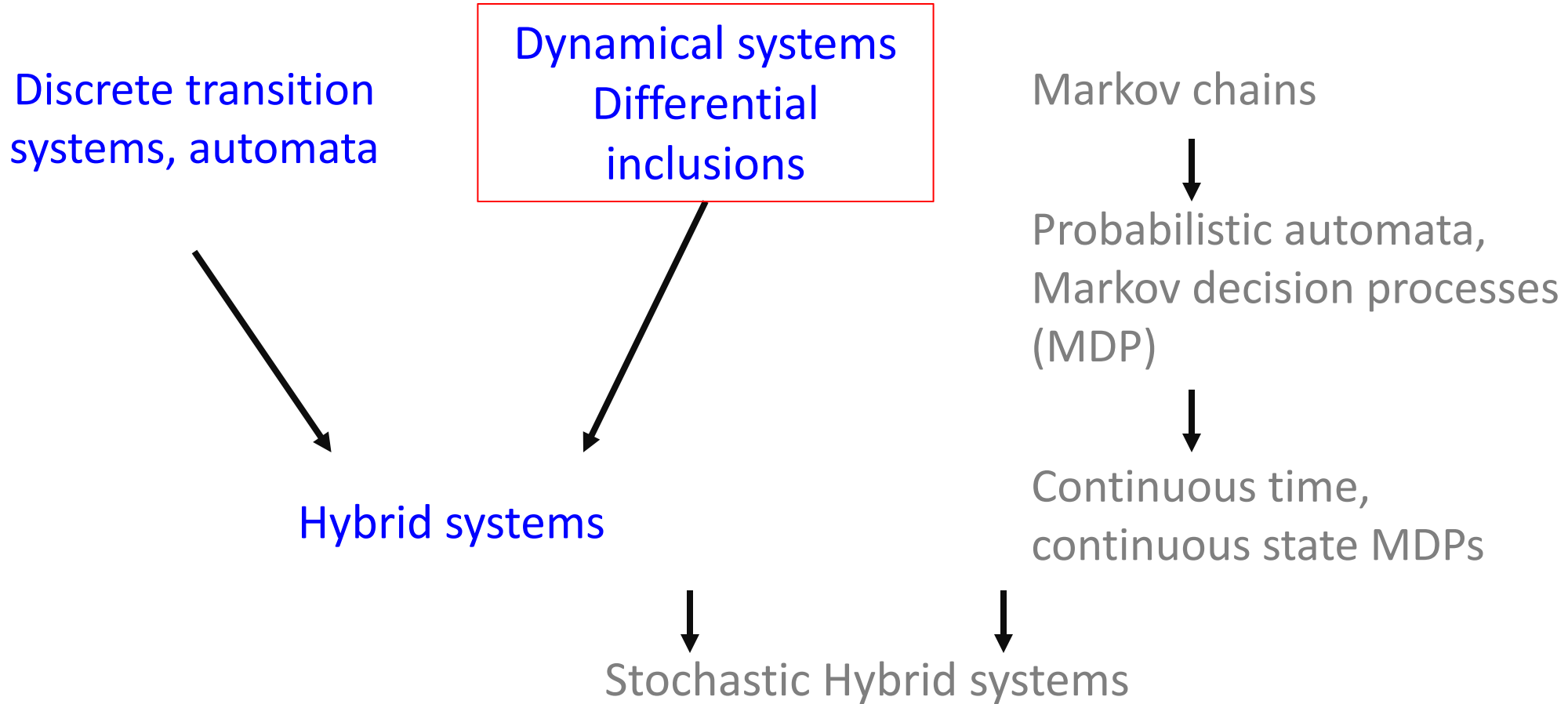
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# Plan

- Dynamical system models
  - notions of solutions
  - Linear dynamical systems
  - Connection to automata
  - Verification requirement: Stability
  - Lyapunov method to verify stability

# Map of CPS models



*All this was in the two plague years 1665 and 1666, for in those days I was in my prime of age for invention, and minded mathematics and philosophy more than at any time since.*

---Isaac Newton

From: Wilczek, Frank. A Beautiful Question: Finding Nature's Deep Design (p. 87).

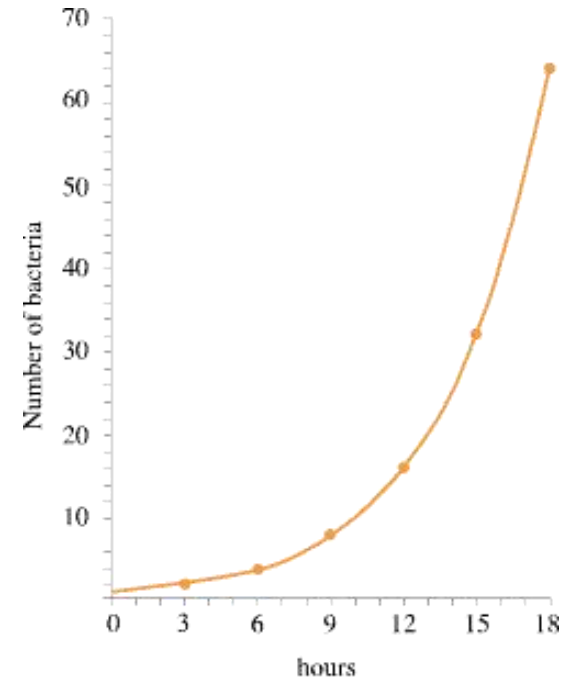
# Introduction to dynamical systems

Behaviors of physical processes are described in terms of instantaneous laws

Example: growth of bacteria

$$\frac{dx(t)}{dt} = x$$

Vehicles, weather, circuits, biomedical processes, ...



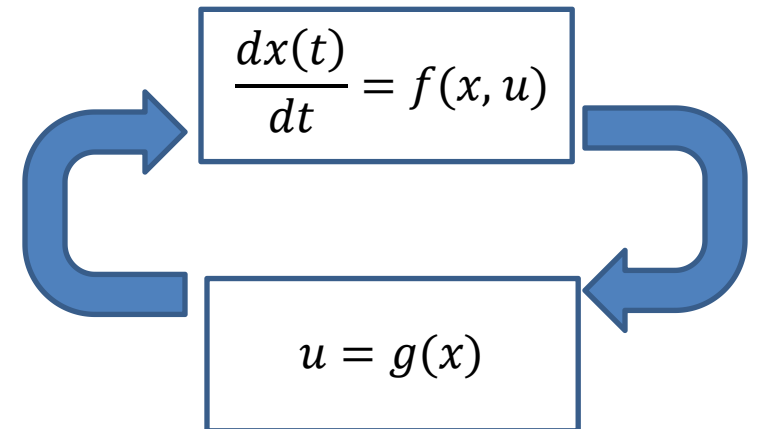
# Introduction to dynamical systems

Behaviors of physical processes are described in terms of instantaneous laws

Common notation:  $\frac{dx(t)}{dt} = f(x(t), u(t), t)$  – Eq. (1)

where time  $t \in \mathbb{R}$ ; **state**  $x(t) \in \mathbb{R}^n$ ; **input**  $u(t) \in \mathbb{R}^m$ ;  $f: \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \rightarrow \mathbb{R}^n$

Example.  $\frac{dx(t)}{dt} = v(t)$ ;  $\frac{dv(t)}{dt} = -g$



# Example: Pendulum

Pendulum equation

$$x_1 = \theta \quad x_2 = \dot{\theta}$$

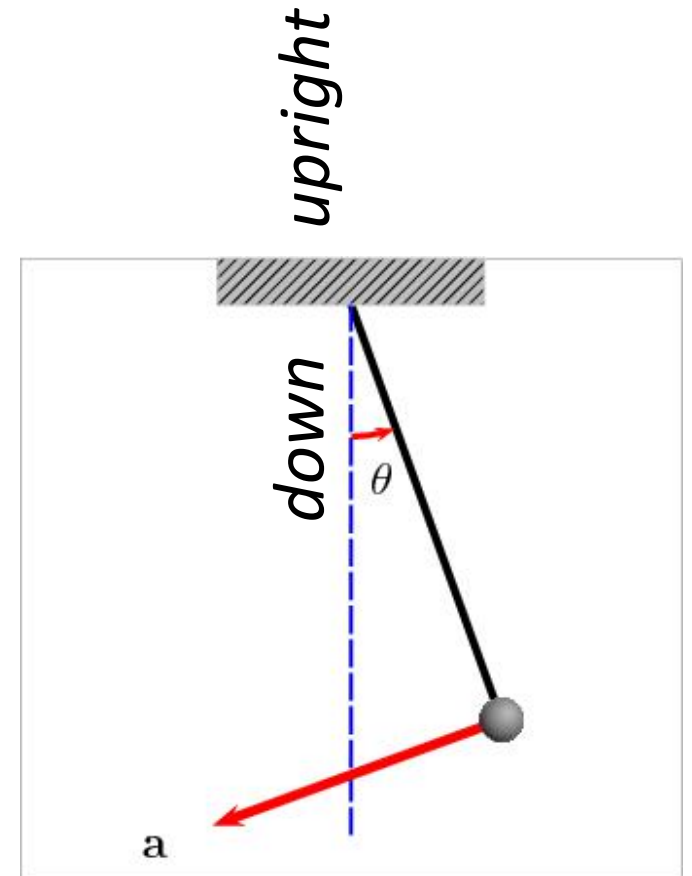
$$x_2 = \dot{x}_1$$

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

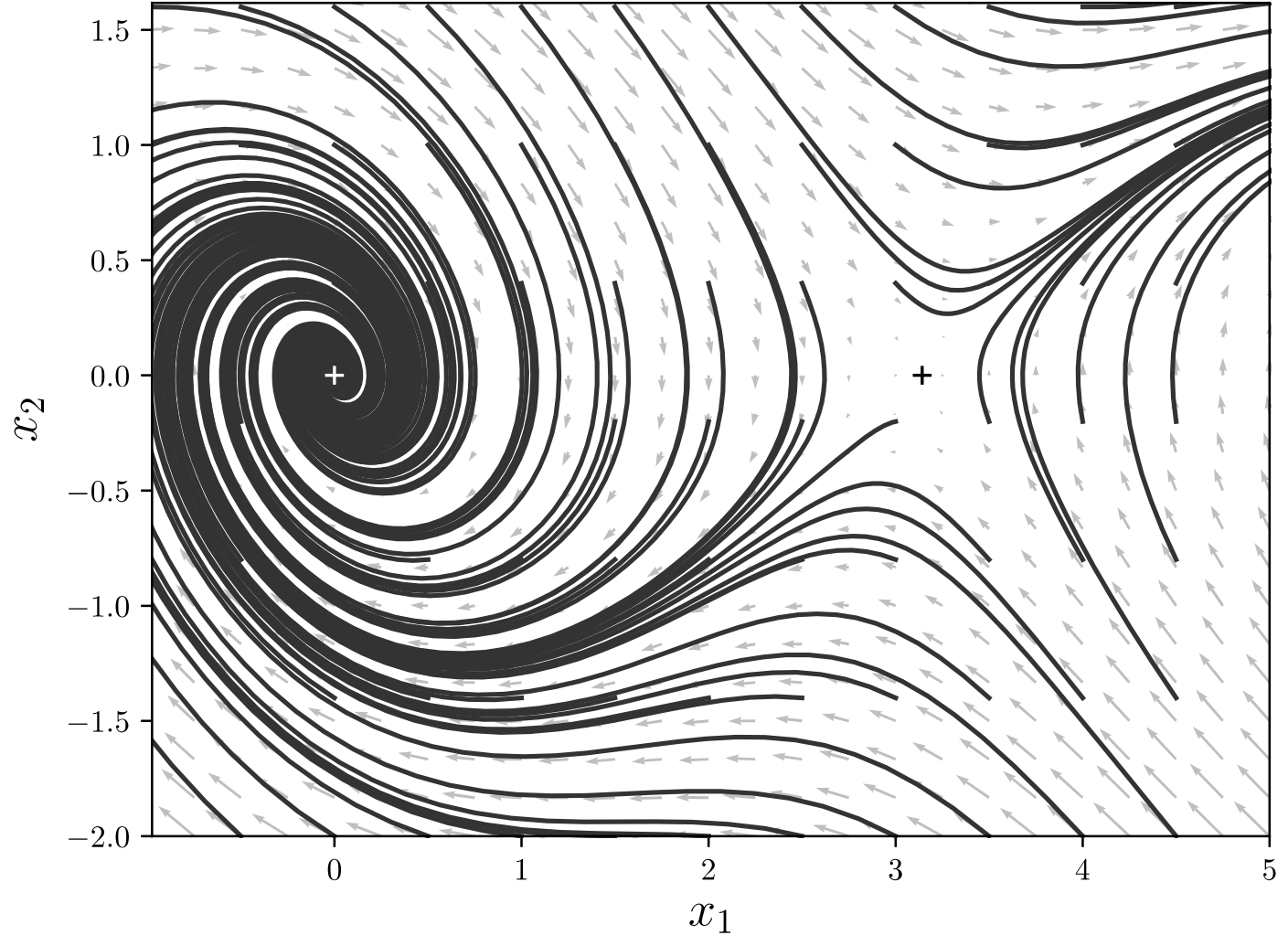
$$\begin{bmatrix} \dot{x}_2 \\ \dot{x}_1 \end{bmatrix} = \begin{bmatrix} -\frac{g}{l} \sin(x_1) - \frac{k}{m} x_2 \\ x_2 \end{bmatrix}$$

$k$ : friction coefficient

Two equilibrium points:  $(0,0)$ ,  $(\pi, 0)$



# Phase portrait of pendulum with friction



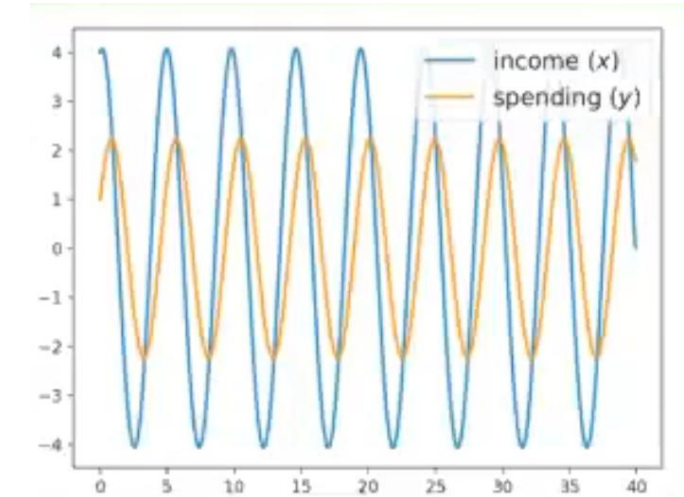
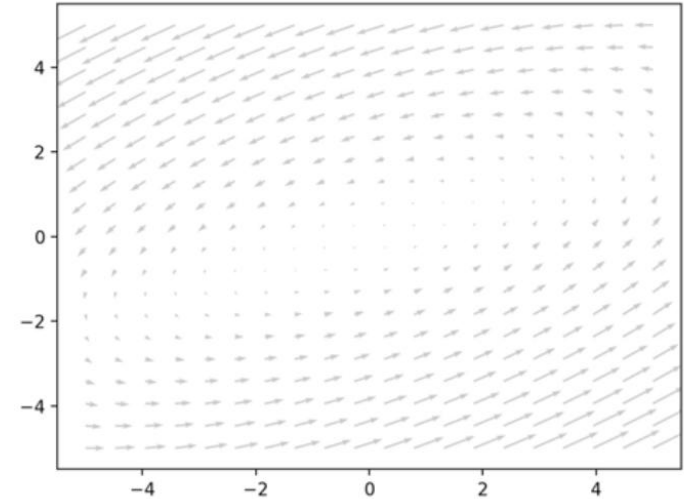


## Example: Simple model of an economy

- $x$ : national income
- $y$ : rate of consumer spending
- $g$ : rate government expenditure (control)

- $\dot{x} = x - \alpha y$

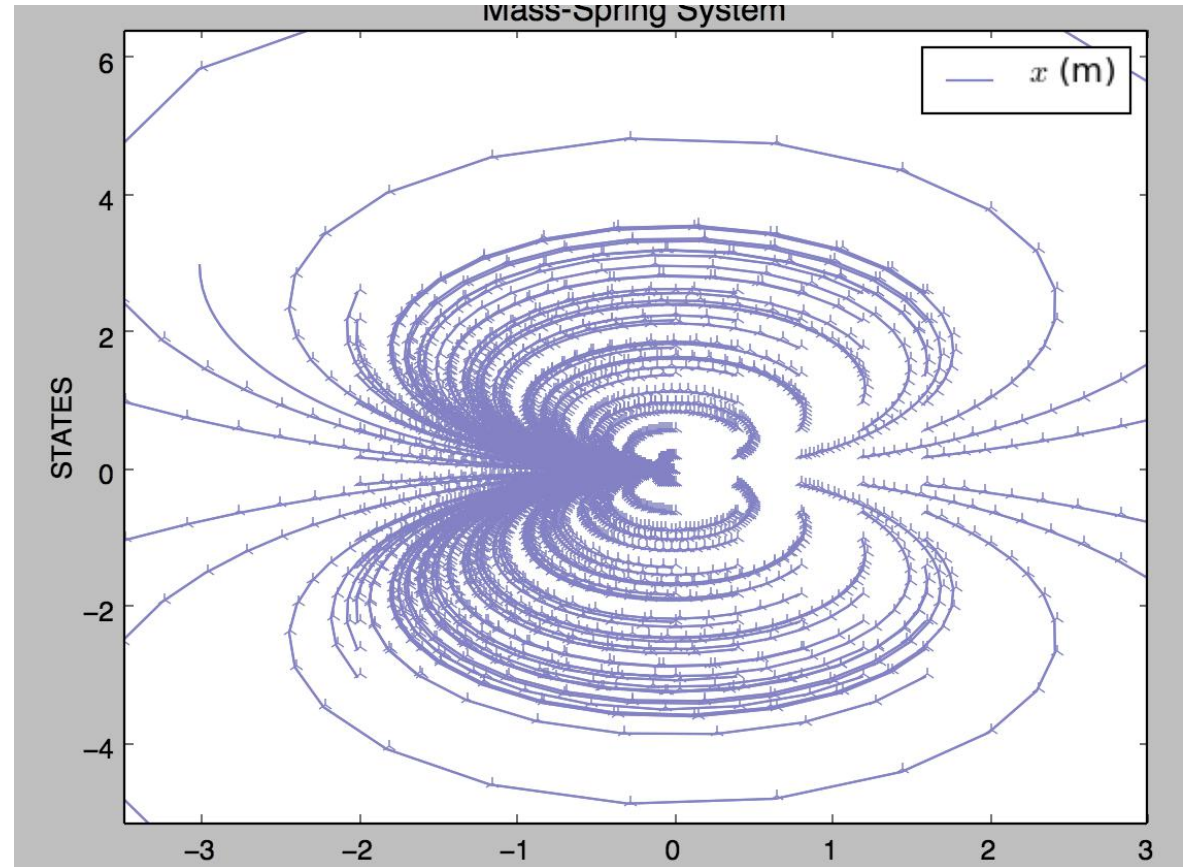
- $\dot{y} = \beta(x - y - g)$



# Butterfly

$$\begin{bmatrix} \dot{x}_2 \\ \dot{x}_1 \end{bmatrix} = \begin{bmatrix} 2x_1x_2 \\ x_1^2 - x_2^2 \end{bmatrix}$$

To plot ODE like this you can use `odeint` from `scipy`



# Van der pol oscillator

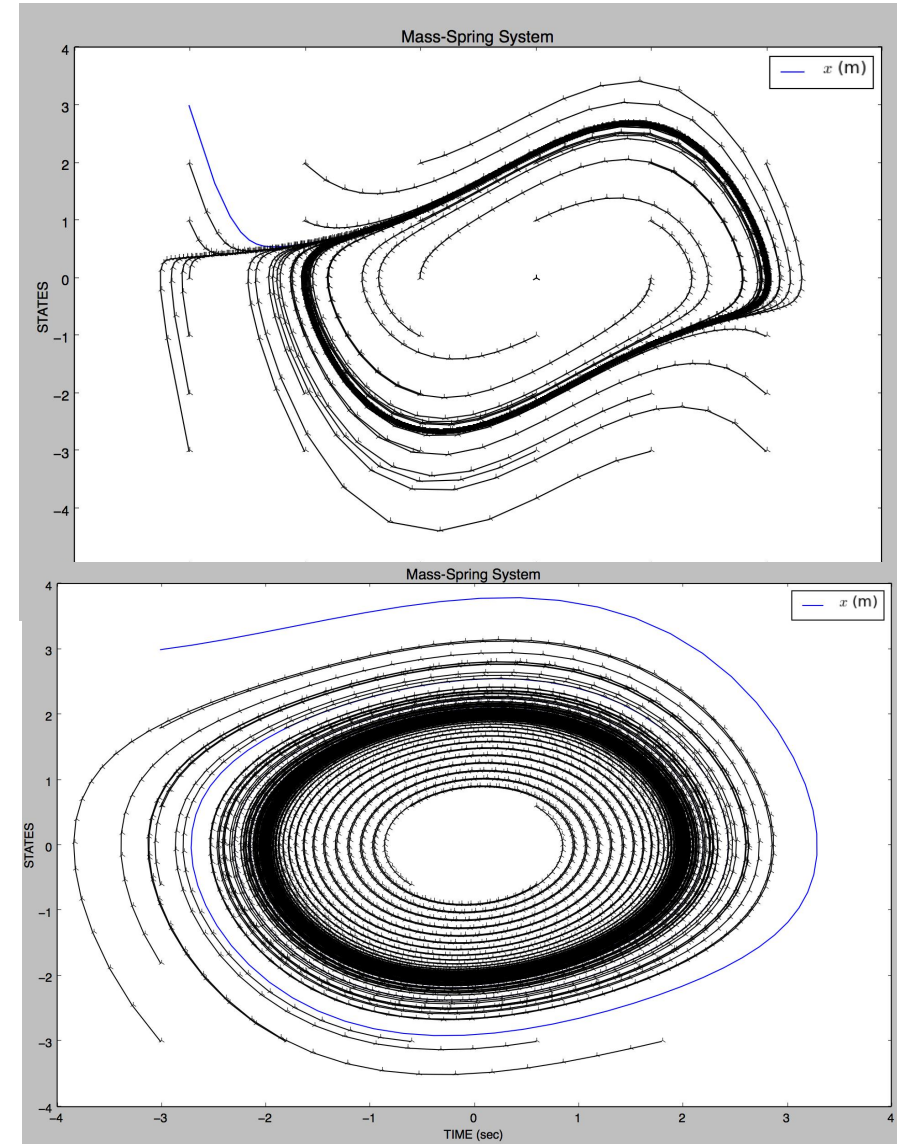
Van der pol oscillator

$$\frac{dx^2}{dt^2} - \mu(1 - x^2) \frac{dx}{dt} + x = 0$$

$$x_1 = x; x_2 = \dot{x}_1;$$

coupling coefficient  $\mu = 20.1$

$$\begin{bmatrix} \dot{x}_2 \\ \dot{x}_1 \end{bmatrix} = \begin{bmatrix} \mu(1 - x_1^2)x_2 - x_1 \\ x_2 \end{bmatrix}$$



# Introduction to dynamical systems

Behaviors of physical processes are described in terms of instantaneous laws

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where time  $t \in \mathbb{R}$ ; **state**  $x(t) \in \mathbb{R}^n$ ; **input**  $u(t) \in \mathbb{R}^m$ ;  $f: \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \rightarrow \mathbb{R}^n$

Example.  $\frac{dx(t)}{dt} = v(t)$ ;  $\frac{dv(t)}{dt} = -g$

Initial value problem: Given system (1) and initial state  $x_0 \in \mathbb{R}^n$ ,  $t_0 \in \mathbb{R}$ , and input  $u: \mathbb{R} \rightarrow \mathbb{R}^m$ , find a state trajectory or *solution* of (1).

# Notions of solution

What is a solution? Many different notions.

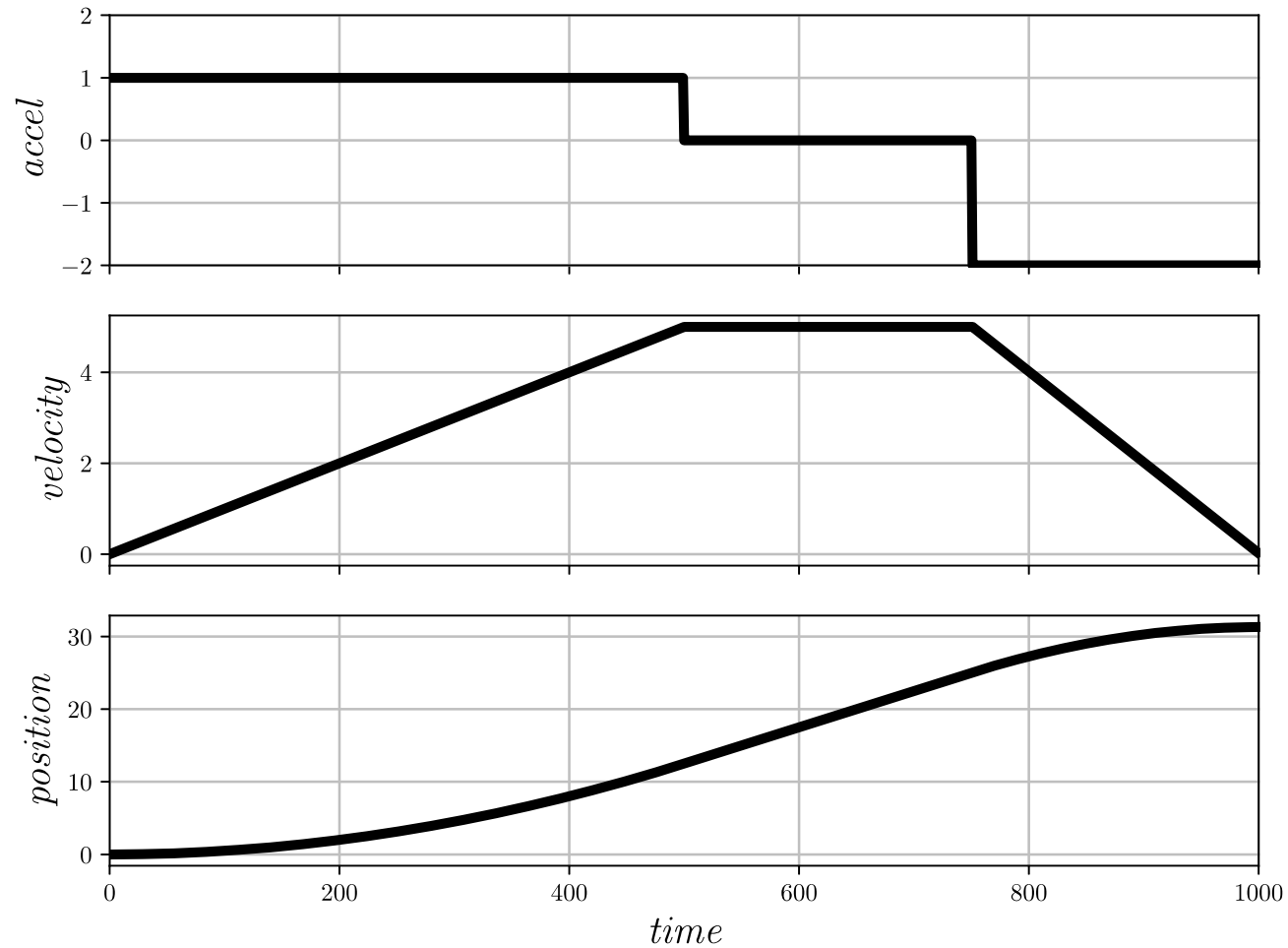
**Definition 1.** (First attempt) Given  $x_0$  and  $u$ ,  $\xi: \mathbb{R} \rightarrow \mathbb{R}^n$  is a solution or trajectory iff

$$(1) \xi(t_0) = x_0 \text{ and}$$

$$(2) \frac{d}{dt} \xi(t) = f(\xi(t), u(t), t), \forall t \in \mathbb{R}.$$

Mathematically makes sense, but too restrictive. Assumes that  $\xi$  is not only continuous, but also differentiable. This disallows  $u(t)$  to be discontinuous, which is often required for optimal control.

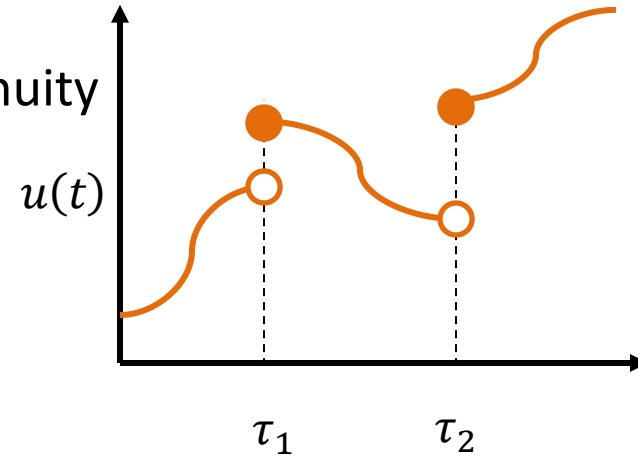
# Getting from point a to point b



# Modified notion

Definition.  $u(\cdot)$  is a piece-wise continuous with set of discontinuity points  $D \subseteq \mathbb{R}^m$  if

- (1)  $\forall \tau \in D, \lim_{t \rightarrow \tau^+} u(t) < \infty; \lim_{t \rightarrow \tau^-} u(t) < \infty$
- (2) Continuous from right  $\lim_{t \rightarrow \tau^+} u(t) = u(t)$
- (3)  $\forall t_0 < t_1, [t_0, t_1] \cap D$  is finite



$PC([t_0, t_1], \mathbb{R}^m)$  is the set of all piece-wise continuous functions over the domain  $[t_0, t_1]$

Definition 2. **Given  $x_0$  and  $u$ ,**  $\xi: \mathbb{R} \rightarrow \mathbb{R}^n$  is a **solution** or trajectory iff (1)  $\xi(t_0) = x_0$  and (2)  $\frac{d}{dt} \xi(t) = f(\xi(t), u(t), t), \forall t \in \mathbb{R} \setminus D$ .

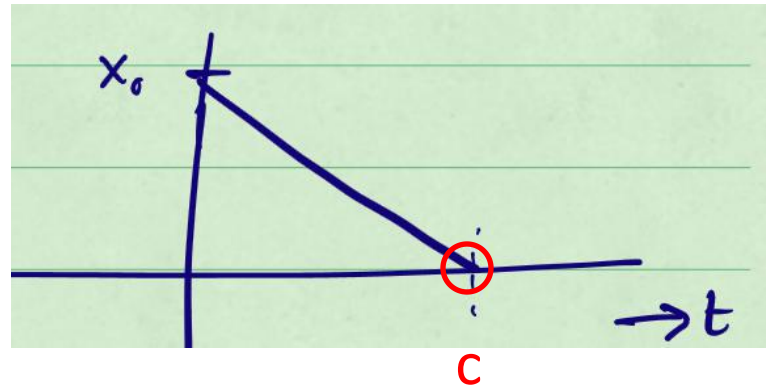
# When can we guarantee the existence of solutions?

Example.  $\dot{x}(t) = -\operatorname{sgn}(x(t))$ ;  $x_0 = c$ ;  $t_0 = 0$ ;  $c > 0$

Solution:  $\xi(t) = c - t$  for  $t \leq c$ ; check  $\dot{\xi} = -1 = -\operatorname{sgn}(\xi(t))$

Problem:  $-\operatorname{sgn}(x(t))$  is discontinuous, cannot find  $\xi$  such that  $\dot{\xi}$  exists and suddenly changes

$$\operatorname{sgn}(x) = \begin{cases} 1, & x \geq 0 \\ -1, & x < 0 \end{cases}$$





# When can we guarantee the existence of solutions?

Example.  $\dot{x}(t) = x^2; x_0 = c; t_0 = 0; c > 0$

Solution:  $\xi(t) = \frac{c}{1-tc}$  works for  $t \neq 1/c$ ;  $\dot{\xi} = \frac{-c(-c)}{(1-tc)^2} = (\xi(t))^2$

Problem: As  $t \rightarrow \frac{1}{c}$  then  $\xi(t) \rightarrow \infty$ ;  $f(x)$  grows too fast

# Lipschitz continuity

A function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is Lipschitz continuous if there exist  $L > 0$  such that for any pair  $x, x' \in \mathbb{R}^n$ ,  $\|f(x) - f(x')\| \leq L\|x - x'\|$

Examples:  $6x + 4$ ;  $|x|$ ; all differentiable functions with bounded derivatives

Non-examples:  $\sqrt{x}$ ;  $x^2$  (locally Lipschitz)

# Existence and uniqueness of solutions

**Theorem.** If  $f(x(t), u(t), t)$  is Lipschitz continuous in the first argument, and  $u(t)$  is PC then (1) has unique solutions.

In general, for nonlinear dynamical systems we do not have closed form solutions for  $\xi(t)$ , but there are numerical solvers

# Linear system and solutions

$$\dot{x}(t) = A(t)x(t) + B(t)u(t)$$

For a given initial state  $x_0 \in \mathbb{R}^n$ ,  $t_0 \in \mathbb{R}$  and  $u(\cdot) \in PC(\mathbb{R}, \mathbb{R}^n)$  the *solution* is a function  $\xi(\cdot, t_0, x_0, u): \mathbb{R} \rightarrow \mathbb{R}^n$

We studied several properties of  $\xi$ : continuity with respect to first and third argument, linearity, decomposition

# Linear time-varying systems

$$\dot{x}(t) = A(t)x(t) + B(t)u(t) \text{ --- (2)}$$

$u(t)$  continuous everywhere except  $D_x$

**Theorem.** Let  $\xi(t, t_0, x_0, u)$  be the solution for (2) with points of discontinuity,  $D_x$

1.  $\forall t_0 \in \mathbb{R}, x_0 \in \mathbb{R}^n, u \in PC(\mathbb{R}, \mathbb{R}^m), \xi(\cdot, t_0, x_0, u): \mathbb{R} \rightarrow \mathbb{R}^n$  is continuous and differentiable  
 $\forall t \in \mathbb{R} \setminus D_x$
2.  $\forall t, t_0 \in \mathbb{R}, u \in PC(\mathbb{R}, \mathbb{R}^m), \xi(t, t_0, \cdot, u): \mathbb{R}^n \rightarrow \mathbb{R}^n$  is continuous
3.  $\forall t, t_0 \in \mathbb{R}, x_{01}, x_{02} \in \mathbb{R}^n, u_1, u_2 \in PC(\mathbb{R}, \mathbb{R}^m), a_1, a_2 \in \mathbb{R}, \xi(t, t_0, a_1x_{01} + a_2x_{02}, a_1u_1 + a_2u_2) = a_1\xi(t, t_0, x_{01}, u_1) + a_2\xi(t, t_0, x_{02}, u_2)$
4.  $\forall t, t_0 \in \mathbb{R}, x_0 \in \mathbb{R}^n, u \in PC(\mathbb{R}, \mathbb{R}^m), \xi(t, t_0, x_0, u) = \xi(t, t_0, x_0, \mathbf{0}) + \xi(t, t_0, \mathbf{0}, u)$