

Lecture 14: Modeling Physics

Huan Zhang

huan@huan-zhang.com

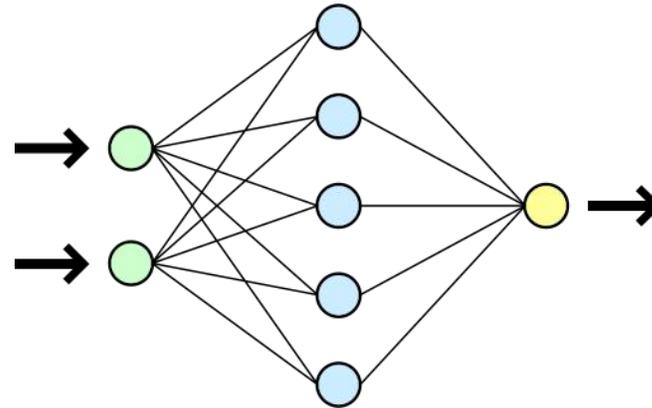
Deadlines

- HW2 due 3/17 11:59 pm
- Two programming questions
- A lot of bonus questions to help you get better grades
- Start early!

Modeling computation -> Modeling Physics



Modeling computation
(code, algorithms, ...)



Modeling neural networks
(a special form of computation)

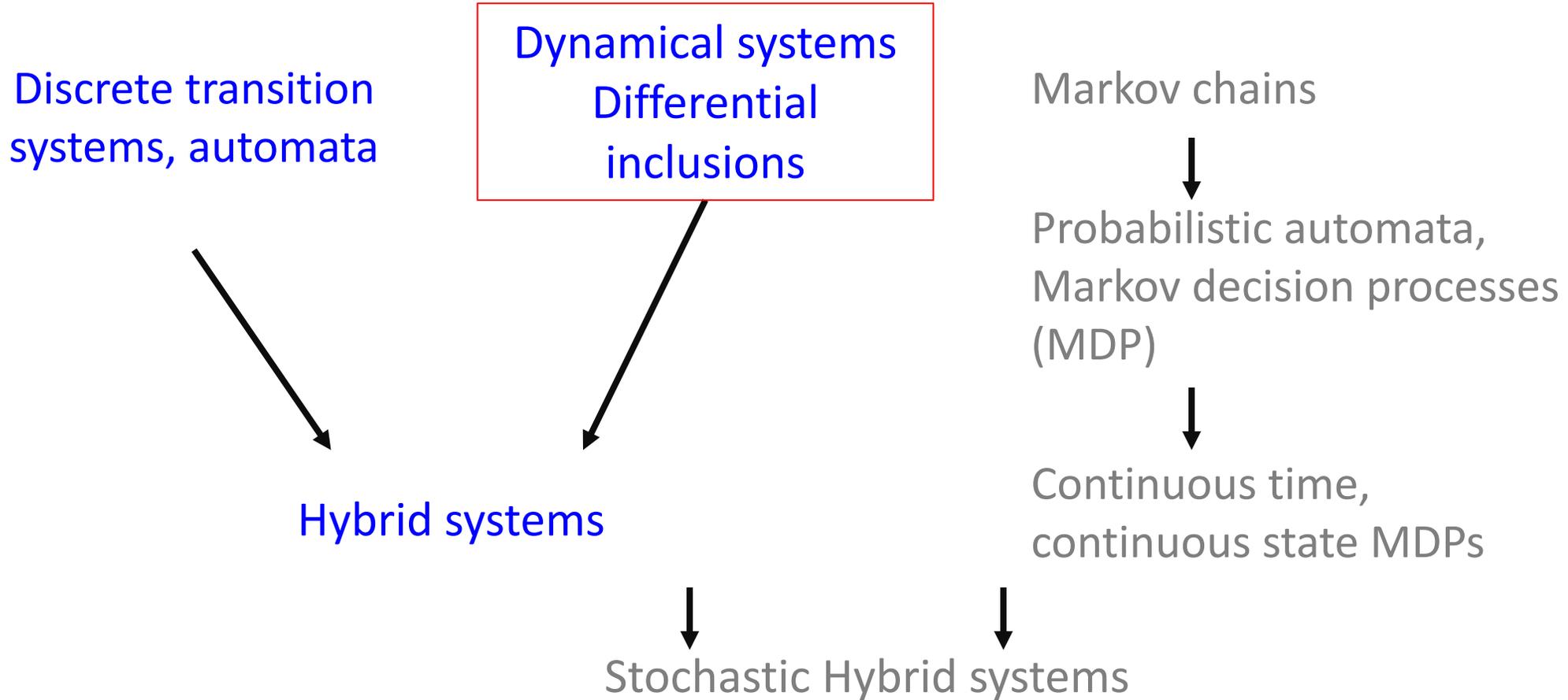


Modeling cyber-physical
systems
(dynamical systems)

Plan

- Dynamical system models
 - What is a dynamical system?
 - 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).

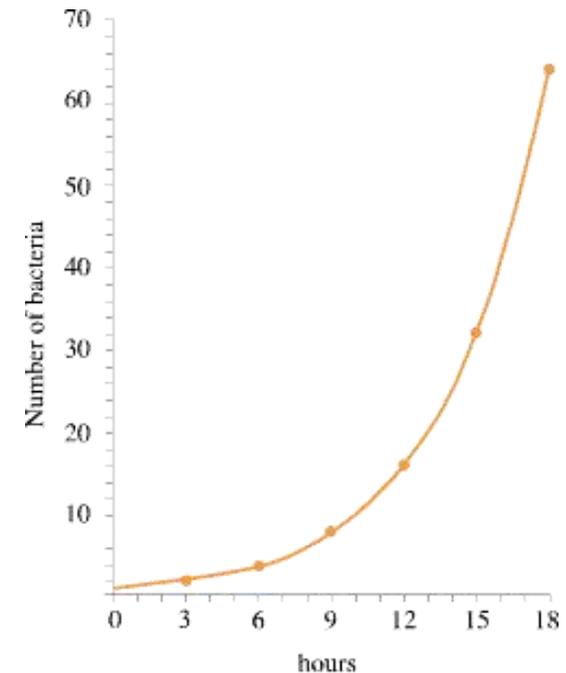
Introduction to dynamical systems

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, ...



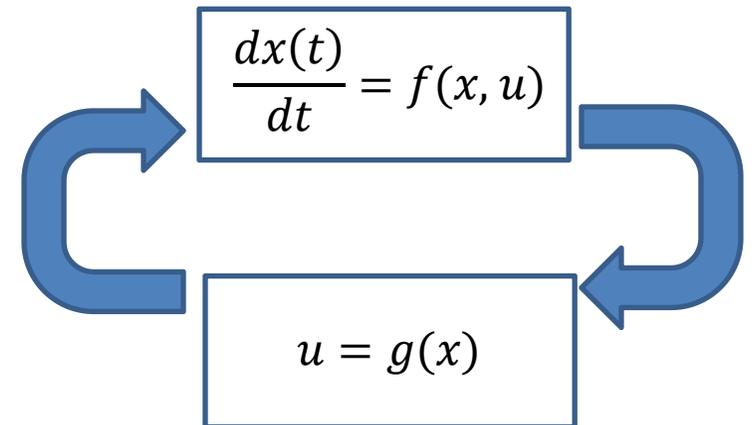
Introduction to dynamical systems

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

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

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

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

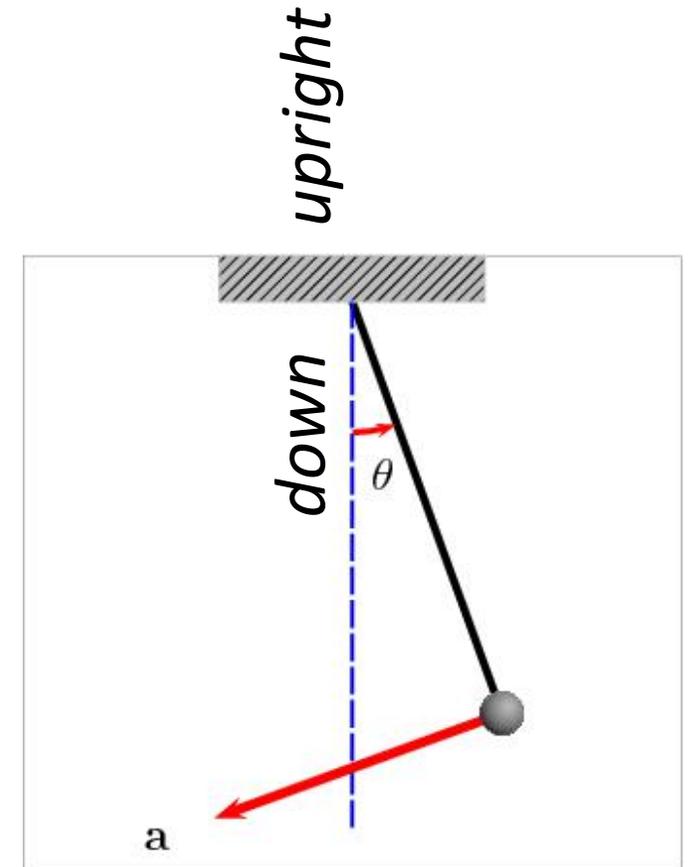
$$\begin{bmatrix} \dot{x}_2 \\ \dot{x}_1 \end{bmatrix} = \begin{bmatrix} -\frac{g}{l} \sin(x_1) - \frac{k}{m} x_2 \\ x_2 \end{bmatrix}$$

g : gravitational acceleration l : pendulum length

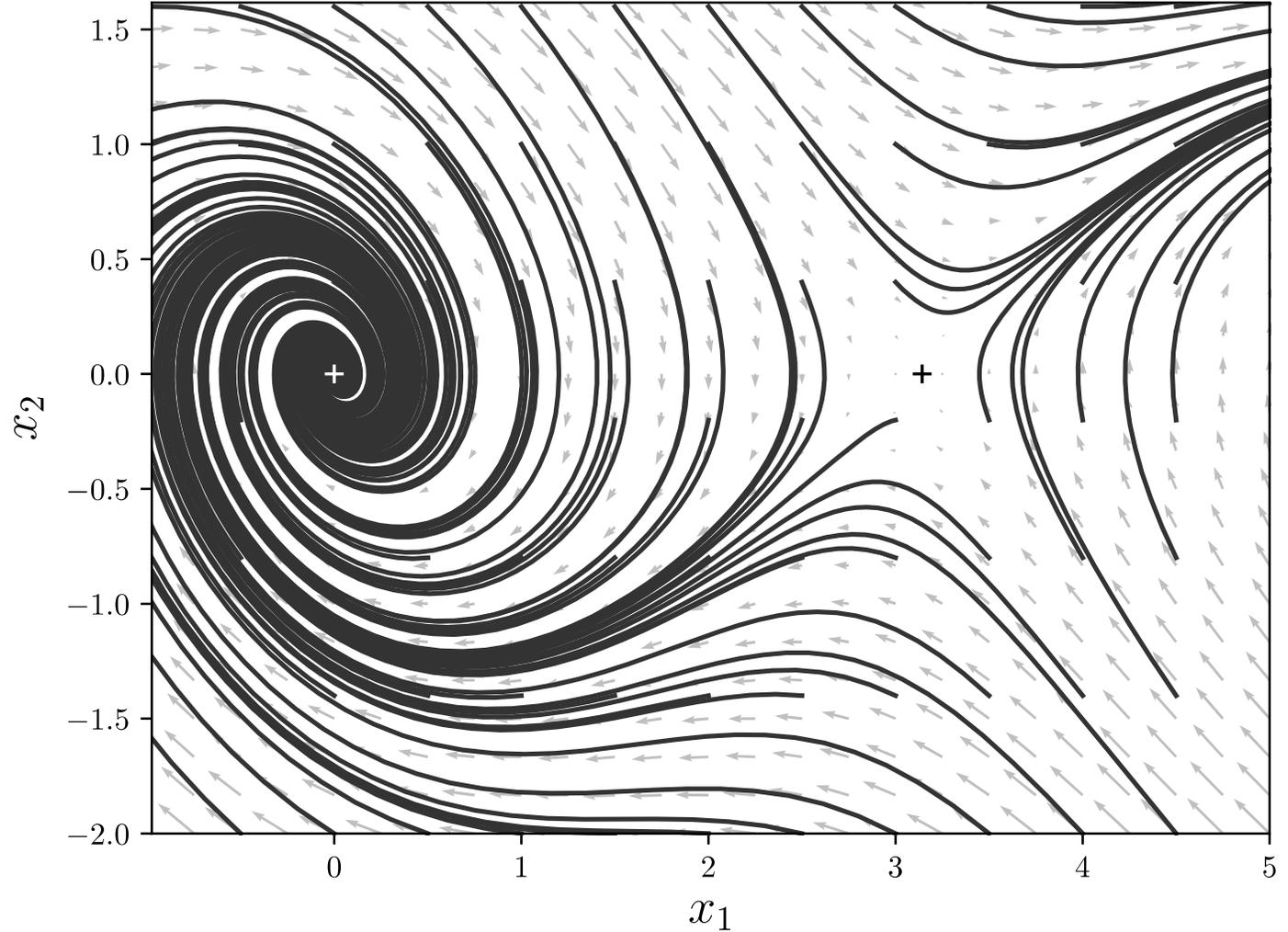
k : friction coefficient m : mass

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

torque = moment of inertia \times angular acceleration



Phase portrait of pendulum with friction

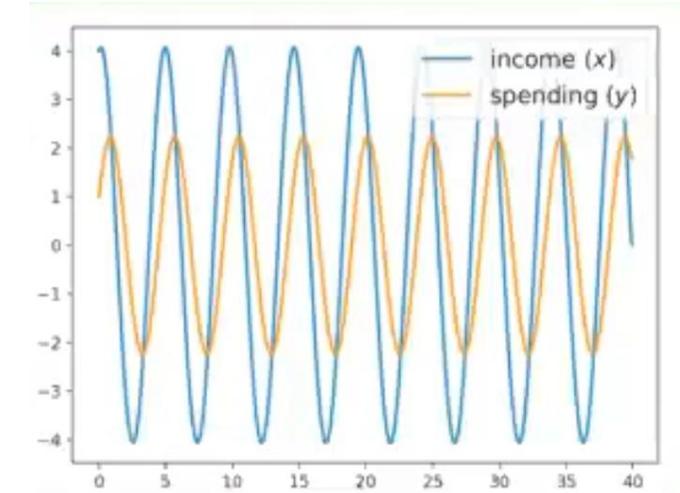
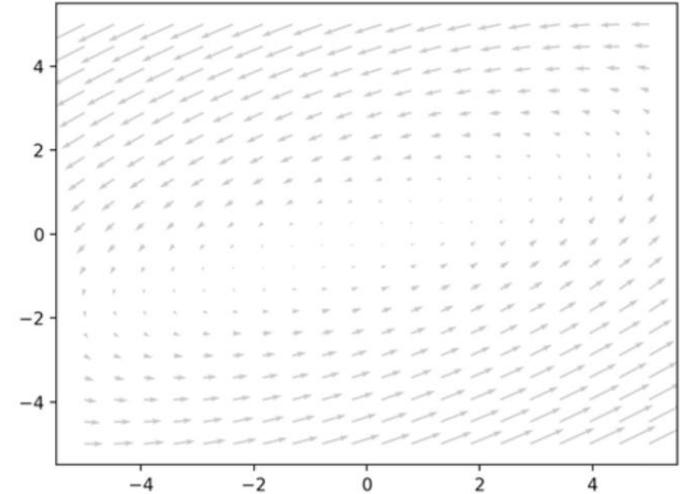


Example: Simple model of an economy

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

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

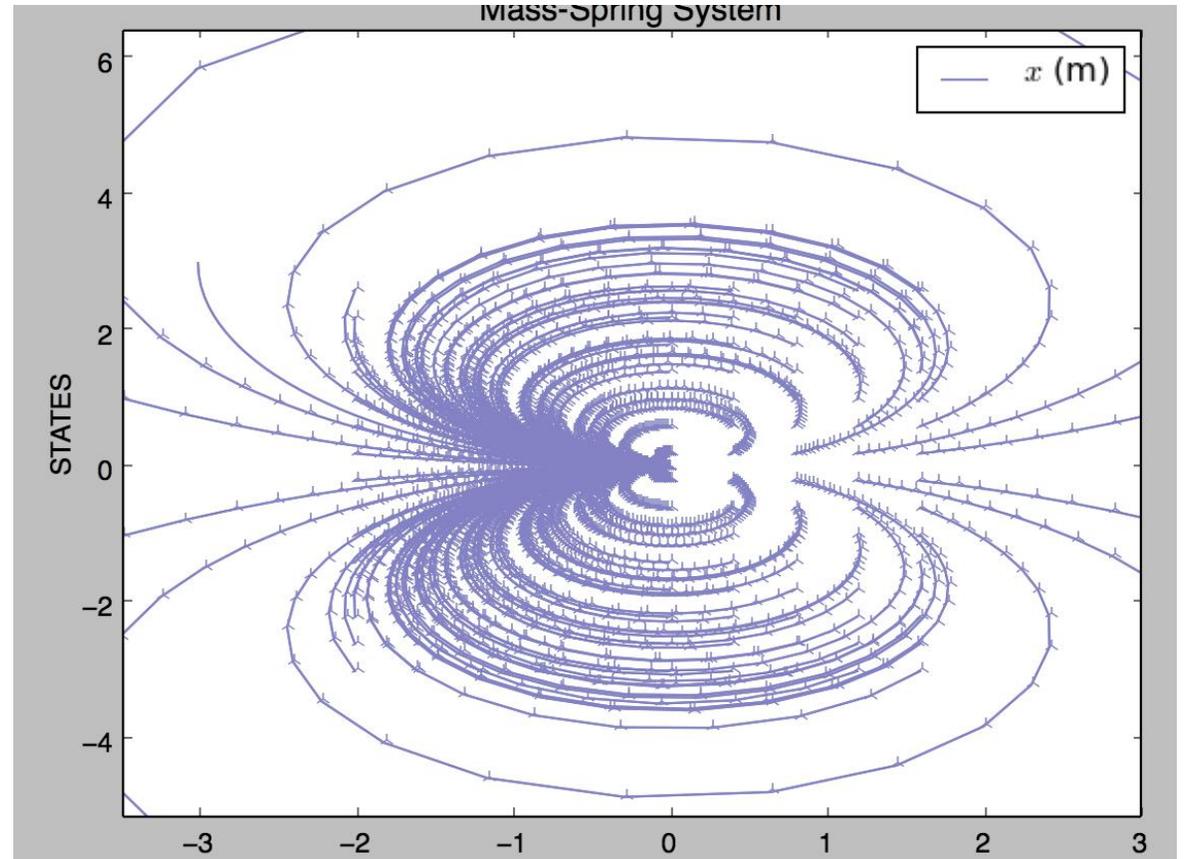
- $\dot{y} = \beta(x - y - g)$



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

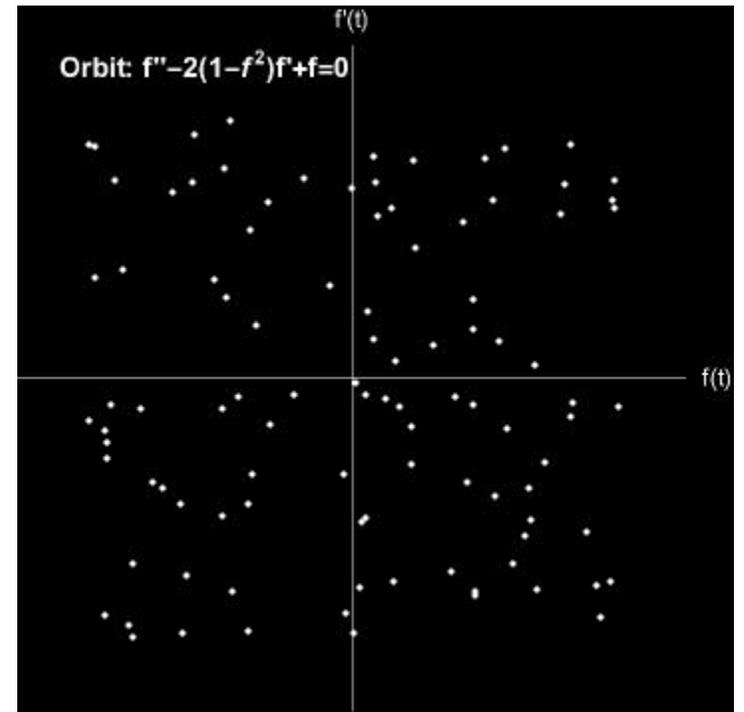
Van der pol oscillator

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

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

coupling coefficient μ

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

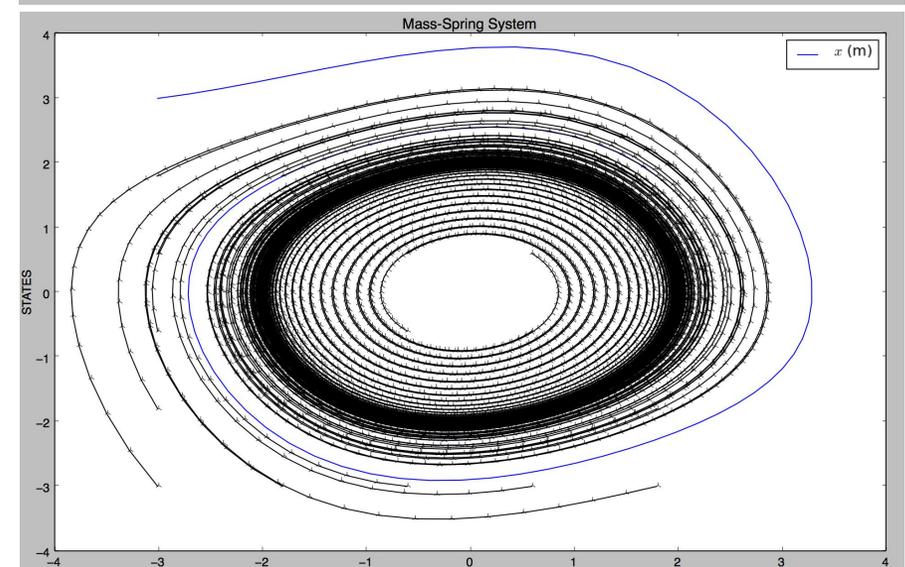
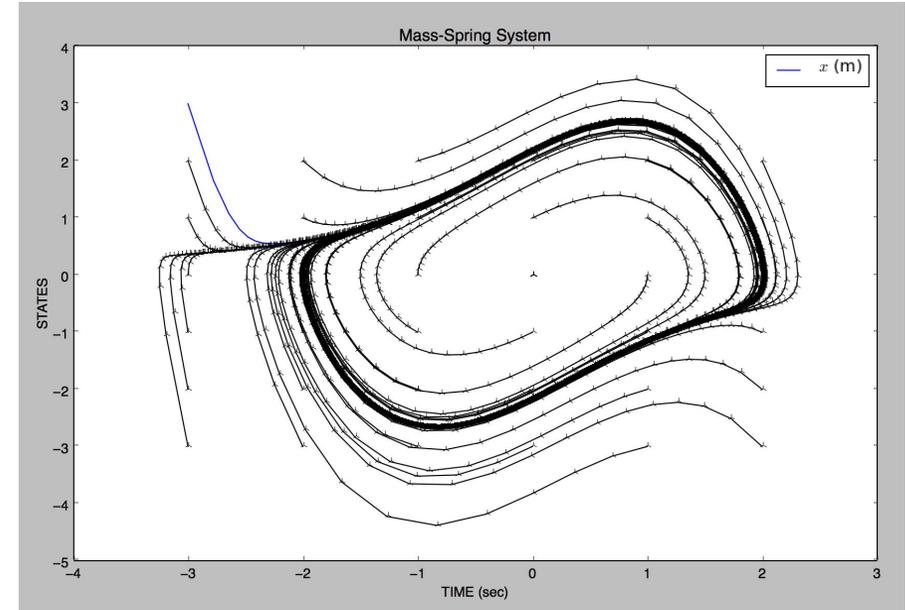


Van der pol oscillator

Van der pol oscillator

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

$\mu = 2$ (top) 0.1 (bottom)



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

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.

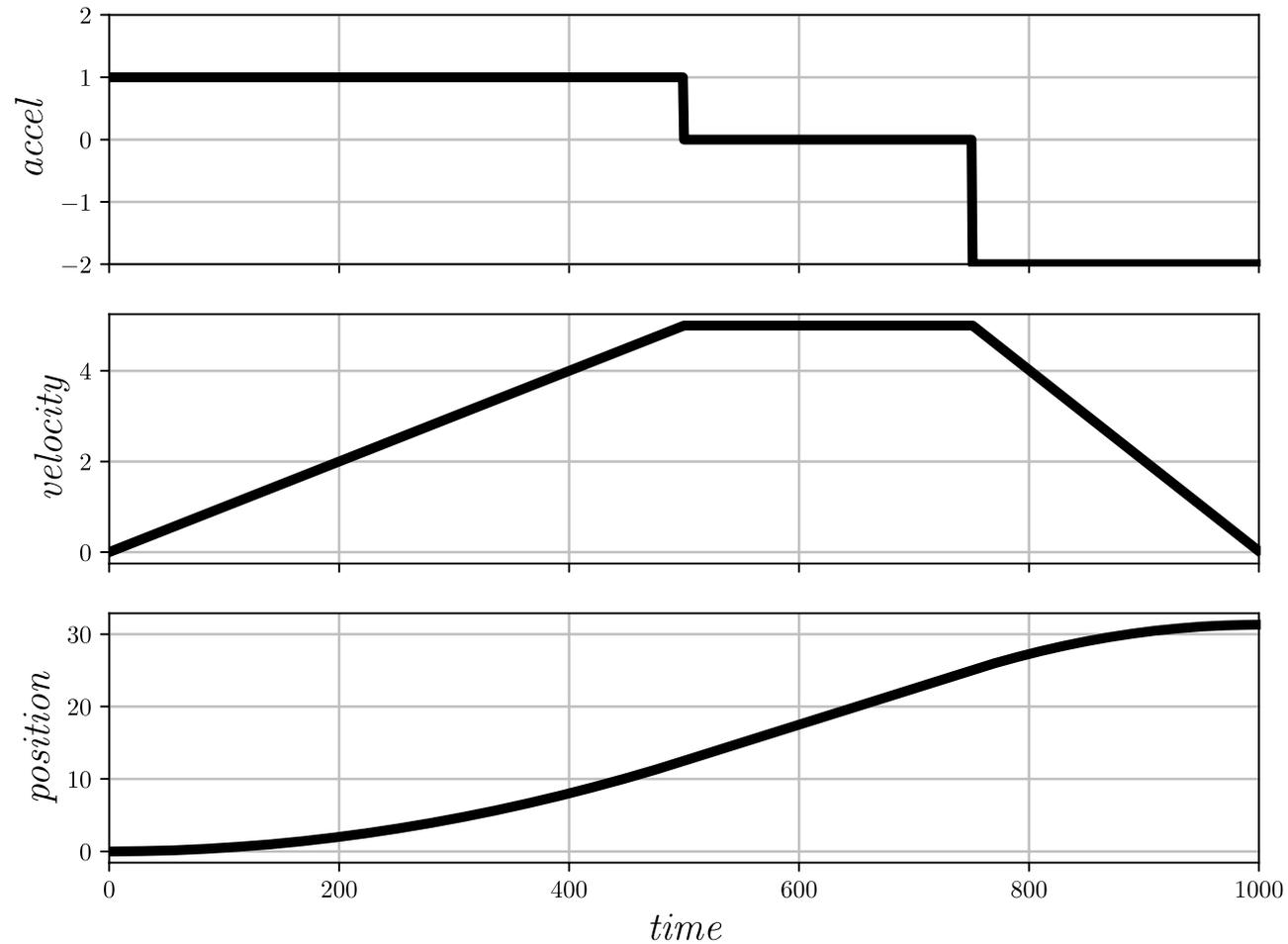
Definition 1. (First attempt) Given x_0 and u , $\xi: \mathbb{R} \rightarrow \mathbb{R}^n$ is a solution or trajectory iff

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

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

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

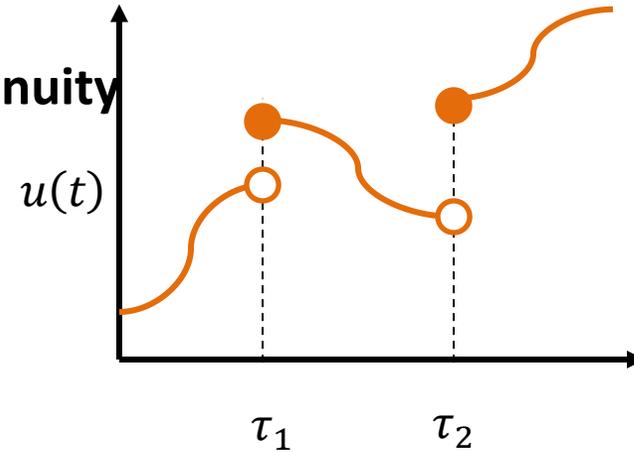
Getting from point a to point b



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

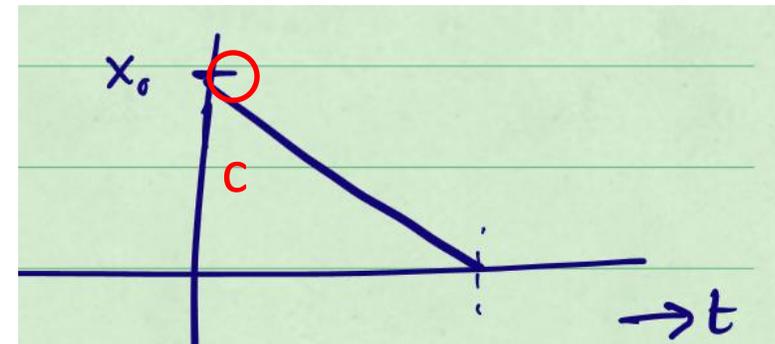
When can we guarantee the existence of solutions?

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; at $t=c$, cannot find ξ such that $\dot{\xi}$ exists and suddenly changes from -1 to 1

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



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} (has issue when x is 0)
 x^2 (has issue when x is infinity)

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 ξ : continuity with respect to first and third argument, linearity, decomposition

Linear time-varying systems

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

$u(t)$ 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. **linearity:** $\forall t, t_0 \in \mathbb{R}, x_{01}, x_{02} \in \mathbb{R}^n, u_1, u_2 \in PC(\mathbb{R}, \mathbb{R}^m), a_1, a_2 \in \mathbb{R}, \xi(t, t_0, a_1x_{01} + a_2x_{02}, a_1u_1 + a_2u_2) = a_1\xi(t, t_0, x_{01}, u_1) + a_2\xi(t, t_0, x_{02}, u_2)$
4. **decomposition:** $\forall t, t_0 \in \mathbb{R}, x_0 \in \mathbb{R}^n, u \in PC(\mathbb{R}, \mathbb{R}^m), \xi(t, t_0, x_0, u) = \xi(t, t_0, x_0, \mathbf{0}) + \xi(t, t_0, \mathbf{0}, u)$