

# Lecture 13: Modeling Physics (Part II: Linear systems)

## Stability Verification

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# Deadlines

Project proposal is due 3/3, 11:59 pm CT

**See Canvas announcement** for some example project ideas on ML + verification

Homework 2 due 3/10, 11:59 pm CT

Two writing problems + two programming problems

# Some project ideas

[https://canvas.illinois.edu/courses/44138/discussion\\_topics/604292](https://canvas.illinois.edu/courses/44138/discussion_topics/604292)

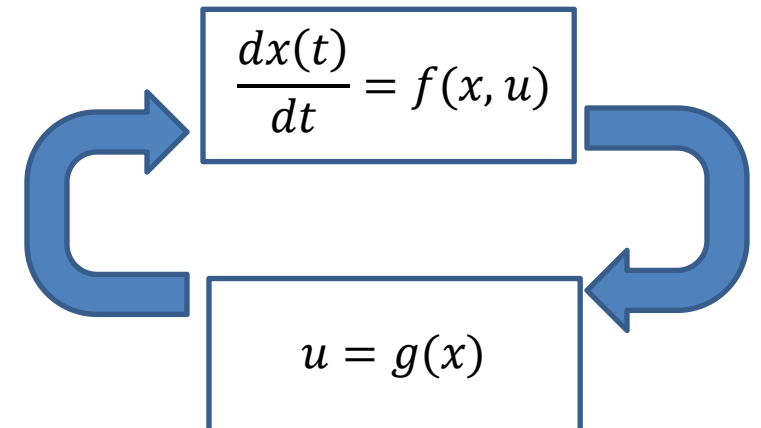
# Review: dynamical systems

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



# Review: dynamical systems

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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).

# Review: Existence and uniqueness of solutions

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

In general, for nonlinear dynamical systems we do not have closed form solutions, but there are numerical solvers

# Linear system and solutions

$$\dot{x}(t) = A(t)x(t) + B(t)u(t) \quad (\text{Linear time varying})$$

$$\dot{x}(t) = Ax(t) + Bu(t) \quad (\text{Linear time invariant})$$

For a given initial state  $x_0 \in \mathbb{R}^n$ ,  $u(t) \in PC(\mathbb{R}, \mathbb{R}^n)$  the *solution* is a function

$$\xi(t, x_0, u): \mathbb{R} \rightarrow \mathbb{R}^n$$

Note that  $t$  is the variable.

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

# Linear system and solutions

$$\dot{x}(t) = Ax(t) + Bu(t)$$

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

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

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



# Linear system and solutions

- $\xi(\cdot, x_0, u): \mathbb{R} \rightarrow \mathbb{R}^n$  is a linear function of the initial state  $x_0$  and input  $u$ , (**linearity property**). Let us first focus on the linear function  $\xi(\cdot, x_0, 0)$  about the initial state  $x_0$
- Define  $\Phi(\cdot)x_0 = \xi(\cdot, x_0, 0)$
- This  $\Phi(\cdot): \mathbb{R} \rightarrow \mathbb{R}^{n \times n}$  is called the state transition matrix

# Linear time invariant system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

A and B are not function of t.

Solution of the system  $\xi$  can be explicitly derived. How to do that?

Consider the decomposition property, we solve two problems:

$$\forall t \in \mathbb{R}, x_0 \in \mathbb{R}^n, u \in PC(\mathbb{R}, \mathbb{R}^m), \xi(t, x_0, u) = \xi(t, x_0, \mathbf{0}) + \xi(t, \mathbf{0}, u)$$

# Linear time invariant system

First set input  $u(t)$  to 0 (we do this due to the decomposition)

$$\frac{dx(t)}{dt} = Ax(t), x(t=0) = x_0$$

Due to linearity, the solution is in this form:

$$x(t) = \phi(t)x_0 = (E + \phi_1 t + \phi_2 t^2 + \dots + \phi_n t^n + \dots)x_0$$

Taylor expansion of  $\Phi(t)$

Substitute into the differential equation:

$$\begin{aligned} \frac{d}{dt}(\phi(t)x_0) &= A\phi(t)x_0 \\ (\phi_1 + 2\phi_2 t + \dots + n\phi_n t^{n-1} + \dots)x_0 &= (A + A\phi_1 t + A\phi_2 t^2 + \dots + A\phi_n t^n + \dots)x_0 \end{aligned}$$

# Linear time invariant system

$$\frac{d}{dt}(\phi(t)x_0) = A\phi(t)x_0$$
$$(\phi_1 + 2\phi_2 t + \dots + n\phi_n t^{n-1} + \dots)x_0 = (A + A\phi_1 t + A\phi_2 t^2 + \dots + A\phi_n t^n + \dots)x_0$$

We want to solve  $\Phi(t)$ , by comparing the terms:

$$\begin{aligned}\phi_1 &= A \\ \phi_2 &= \frac{1}{2}A\phi_1 = \frac{1}{2!}A^2 \\ \phi_3 &= \frac{1}{3}A\phi_2 = \frac{1}{3!}A^3 \\ &\dots \\ \phi_n &= \frac{1}{n!}A^n.\end{aligned}$$

# Linear time invariant system

$$\phi_1 = A$$

$$\phi_2 = \frac{1}{2} A \phi_1 = \frac{1}{2!} A^2$$

$$\phi_3 = \frac{1}{3} A \phi_2 = \frac{1}{3!} A^3$$

...

$$\phi_n = \frac{1}{n!} A^n.$$

$$\phi(t) = e^{At} = E + At + \frac{1}{2!} A^2 t^2 + \frac{1}{3!} A^3 t^3 + \dots + \frac{1}{n!} A^n t^n + \dots$$

$$\forall t \in \mathbb{R}, x_0 \in \mathbb{R}^n, u \in PC(\mathbb{R}, \mathbb{R}^m), \xi(t, x_0, u) = \boxed{\xi(t, x_0, \mathbf{0})} + \xi(t, 0, u)$$

This part done

# Linear time invariant system

Consider the decomposition property, we solve two problems:

$$\forall t \in \mathbb{R}, x_0 \in \mathbb{R}^n, u \in PC(\mathbb{R}, \mathbb{R}^m), \xi(t, x_0, u) = \xi(t, x_0, \mathbf{0}) + \xi(t, \mathbf{0}, u)$$

Now the second part

Now for  $\xi(t, \mathbf{0}, u)$ , assume  $x_0 = \mathbf{0}$ , solve  $\frac{dx(t)}{dt} = Ax(t) + Bu(t)$

Rearrange:  $\frac{dx(t)}{dt} - Ax(t) = Bu(t)$

Multiply a common factory:  $e^{-At} \frac{dx(t)}{dt} - e^{-At} Ax(t) = e^{-At} Bu(t)$

Note the perfect differential:  $\frac{d}{dt}(e^{-At}x(t)) = e^{-At} Bu(t)$

# Linear time invariant system

$$\frac{d}{dt}(e^{-At}x(t)) = e^{-At}Bu(t)$$

Integration on both sides:

$$\int_0^t \frac{d}{d\tau}(e^{-A\tau}x(\tau)) = \int_0^t e^{-A\tau}Bu(\tau) d\tau$$

$$e^{-At}x(t) - e^{A0}x(0) = \int_0^t e^{-A\tau}Bu(\tau) d\tau$$

Since  $x(0) = 0$ :

$$x(t) = e^{At} \int_0^t e^{-A\tau}Bu(\tau) d\tau$$

$$x(t) = \int_0^t e^{A(t-\tau)}Bu(\tau) d\tau$$

# Linear time invariant system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

Define Matrix exponential:

$$e^{At} = 1 + At + \frac{1}{2!}(At)^2 + \dots = \sum_0^{\infty} \frac{1}{k!}(At)^k$$

**Theorem.**  $\xi(t, x_0, u) = \underbrace{\Phi(t)x_0}_{\text{Zero input}} + \underbrace{\int_0^t e^{A(t-\tau)}Bu(\tau)d\tau}_{\text{Zero state}}$

Here  $\Phi(t) := e^{At}$  is the state-transition matrix



# Example

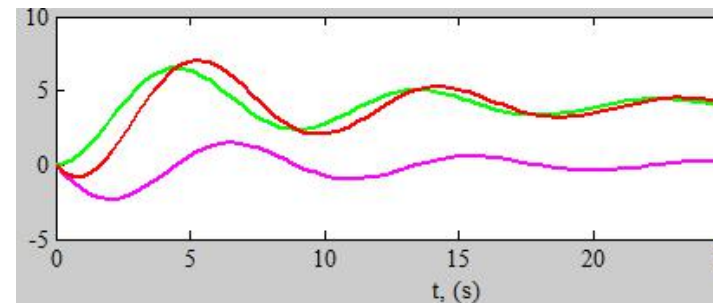
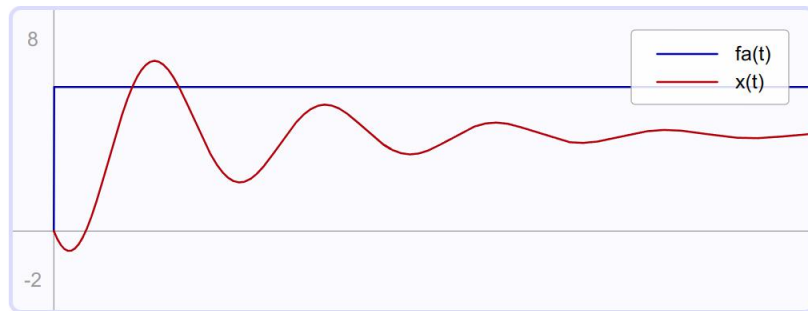
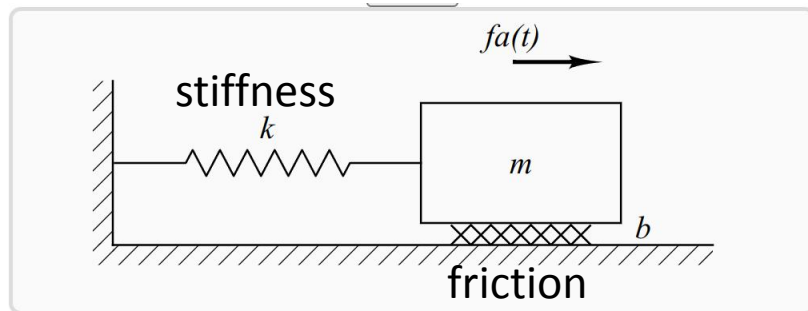
$$\frac{dx(t)}{dt} = Ax(t) + Bu(t)$$

States  $x$ : position (0m), velocity (-2m/s),

Input  $u(t)$ : Force  $f_a(t) = 6$  Newtons

$$m \frac{dx_2(t)}{dt} = u(t) - b \frac{dx_1(t)}{dt} - kx_1(t)$$

$$x_2(t) = \frac{dx_1(t)}{dt}$$



Zero state  
Complete  
Zero input

# Solution of linear time-varying systems in $\Phi$

More generally, for time varying systems we have

**Theorem.**

$$\xi(t, t_0, x_0, u) = \Phi(t, t_0)x_0 + \int_{t_0}^t \Phi(t, \tau)B(\tau)u(\tau)d\tau$$

Note that  $\Phi(t, t_0)$  here also includes  $t_0$  as an parameter

# Discrete time models / discrete transition systems

- $x(t + 1) = f(x(t), u(t))$
- $x(t + 1) = f(x(t))$  No input (autonomous)
- Execution defined as:  $x_0, f(x_0), f^2(x_0), \dots$
- Can be define as an automaton  $A = \langle Q, Q_0, T \rangle$ 
  - $Q = \mathbb{R}^n, Q_0 = \{x_0\}$
  - $T: \mathbb{R}^n \rightarrow \mathbb{R}^n; T(x) = f(x)$
- Deterministic

# Discretized or sampled-time model

- $\dot{x}(t) = f(x(t), u(t))$
- Assume:  $u \in PC(\mathbb{R}, U)$  where  $U \subseteq \mathbb{R}^m$  is a finite set
- Given the solution  $\xi(t, x_0, u)$
- Fix a sampling period  $\delta > 0$
- $A_\delta = \langle Q, Q_0, U, T \rangle$ 
  - $Q = \mathbb{R}^n$ ,  $Q_0 = \{x_0\}$ ,  $Act = U$ ,
  - $T \subseteq \mathbb{R}^n \times U \times \mathbb{R}^n$ ;  $(x, u, x') \in T$  iff  $x' = \xi(\delta, x, u)$

# Properties for dynamical systems

What type of properties are we interested in?

- Invariance (as in the case of automata)
- State remains bounded
- Converges to target
- Bounded input gives bounded output (BIBO)

# Requirements: Stability

- We will focus on time invariant autonomous systems (closed systems, systems without inputs)
- $\dot{x}(t) = f(x(t))$ ,  $x_0 \in \mathbb{R}^n$ ,  $t_0 = 0$
- $\xi(t)$  is the solution
- $|\xi(t)|$  norm
- $x^* \in \mathbb{R}^n$  is an **equilibrium point** if  $f(x^*) = 0$ .
- For analysis we will assume **0** to be an equilibrium point without loss of generality

# Example: Pendulum

Pendulum equation

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

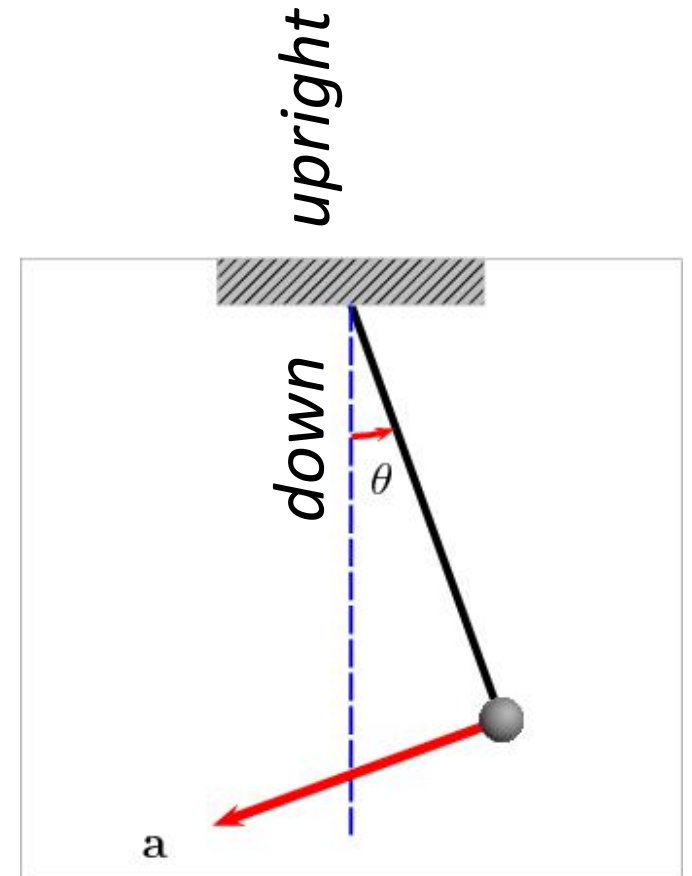
$$\dot{x}_2 = x_1$$

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

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

$k$ : friction coefficient

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



*CW*

$x$  (m)

*speed=0*

*stable*

*unstable*

*down*

*upright*

*CCW*

