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EE 587

01/09/23

# Nonlinear Control Systems

## Lecture 1

Today: What is this course about?

Topic: Nonlinear Dynamical Systems

Ex: Static equation:  $F(x) = 0$

↳ no time dependence

$x := \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n \quad (\mathbb{R}^{n \times 1})$

$x_i \in \mathbb{R}$ : scalar

$F$ : nonlinear function of vector  $x$

①

Special case:  $F(x) = Ax - b$

$A \in \mathbb{R}^{m \times n}$ : matrix ( $m$  rows;  $n$  columns)

$b \in \mathbb{R}^m$ : vector (with  $m$  components)

$Ax - b = 0 \Leftrightarrow Ax = b$ : Topic of EE510  
(Linear Algebra)

Ex: Optimization

Unconstrained optimization problem

minimize  $g(x)$

Optimality <sup>$x$</sup>  conditions: (necessary)

$\nabla g(x) = 0$ ; If  $F(x) := \nabla g(x) \Rightarrow$

$F(x) = 0$

②

Challenge: In general, it is difficult to find an explicit solution to  
 $\nabla g(x) = 0 \quad (F(x) = 0)$

One approach: use iterative techniques to find a solution (e.g. gradient descent)

$$x^{k+1} = x^k - \underbrace{\alpha^k}_{\text{step-size}} \nabla g(x^k) : \text{discrete time}$$

$$\frac{dx}{dt} = -\nabla g(x) : \text{continuous time (gradient flow)}$$

Dynamical systems described by nonlinear recursive (DT) or ODEs (CT) relations ③



Q: What are transient and asymptotic properties of nonlinear dynamical systems?

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Ex: 
$$\dot{x} = -x + \Phi(Ax + Bu + b)$$

$F(x) = 0$ : implicit Neural Network

$x$ : Neural state ( $x \in \mathbb{R}^n$ )

$\Phi$ : nonlinear activation function

$A, B$ : synaptic weights

$U$ : input

$b$ : bias term

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Recurrent NN:

$$\frac{dx}{dt} = -x + \Phi(Ax + Bu + b)$$

In both examples: ODEs of the form

$$\frac{dx}{dt} = f(x) \quad (*) \quad t: \text{time } (t \in \mathbb{R}_+)$$

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

$$f: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

(nonlinear  
function  
of  $x$ )

In discrete time

$$x^{t+1} = h(x^t); \quad t = 0, 1, 2, 3, \dots$$

$$\xrightarrow{\text{IF}} \frac{x^{t+1} - x^t}{\Delta t} \approx f(\underline{x^t}) \Rightarrow$$

$$h(x^t) = x^t - \Delta t \cdot f(x^t)$$

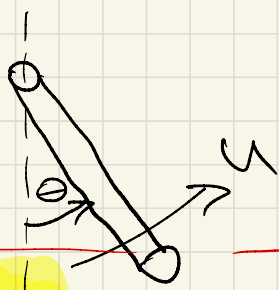
Forward  
Euler

discretization  
(explicit)

We'll mostly focus on  
ODEs in EE587 (CT problems)

(5)

Ex:



2nd Newton's law leads to a nonlinear differential Eq.

$$m \cdot l \cdot \frac{d^2 \theta}{dt^2} + l \cdot k \cdot \frac{d\theta}{dt} + m \cdot g \cdot \sin \theta = u$$

Not in form  $\textcircled{*} \left[ \frac{dx}{dt} = f(x, u) \right]$

Q: How to bring higher order <sup>external input</sup>  $\theta$  into the 1st order form  $\textcircled{*}$  (with input  $u$ ):

$$\begin{aligned} x_1 &= \theta \rightarrow \dot{x}_1 = \dot{\theta} = x_2 \\ x_2 &= \frac{d\theta}{dt} = \dot{\theta} \rightarrow \dot{x}_2 = \ddot{\theta} = \left[ -\frac{g}{l} \sin x_1, -\frac{k}{m} \dot{\theta} + \frac{1}{ml} u \right] \textcircled{6} \end{aligned}$$

Thus:  $\dot{x} = f(x, u)$  with  $x := \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

$$f(x, u) = \begin{bmatrix} f_1(x_1, x_2, u) \\ f_2(x_1, x_2, u) \end{bmatrix} = \begin{bmatrix} -\frac{g}{l} \sin x_1 - \frac{k}{m} x_2 + \frac{1}{ml} u \end{bmatrix}$$

Note: For systems [ODEs] without derivatives  
 (higher order)  
 of input  $u$  [wrt time], we can choose  
 "physical state variables" as components  
 of the vector  $x$  [phase coordinate form]

$$x_1 = \text{"output"} = \text{angle } \theta$$

$$x_2 = \frac{d \text{output}}{dt} = \text{velocity } \dot{\theta}$$

(7)

Note:  $\dot{x} = f(x, u)$ :  $f$  does NOT explicitly depend on time  
time-invariant system (only function of  $x$  and  $u$ )

Time-varying systems:

$$\dot{x} = f(x, u, t) \rightarrow \text{explicit time dependence}$$

Ex: For pendulum:

$m, l, k = \text{const} \Rightarrow$  time-invariant system

otherwise: time-varying

$\rightarrow$  if either of them depends on time  $t$

EE 505: (Linear Systems)

LT) Systems:  $\dot{x} = Ax + Bu$  ↑  $A, B$  constant matrices

Time-varying:  $\dot{x} = A(t)x + B(t)u$

→ solution given by:

$$x(t) = \underbrace{e^{At} x(0)}_{\text{unforced response}} + \underbrace{\int_0^t e^{A(t-\tau)} B u(\tau) d\tau}_{\text{forced response}}$$

unforced response  
(response to  $x_0$ )

forced response  
(response to inputs)

Note: For pendulum: principle of superposition does NOT apply!!! (9)

For nonlinear systems, principle of superposition does NOT hold and very rarely we can write an explicit solution.

☹️ We don't want to give up, in spite of this pessimistic statement: we want to study properties of nonlinear dynamical systems by exploiting structural features on nonlinear terms

Ex: Logistic Equation  $\rightarrow$  growth function  
 $\dot{x} = \alpha \cdot (1 - \frac{x}{K}) \cdot x$  (depends on  $x$ )

$\alpha, K > 0$ : positive scalars (constants)

$$x(t) \in \mathbb{R}_+$$

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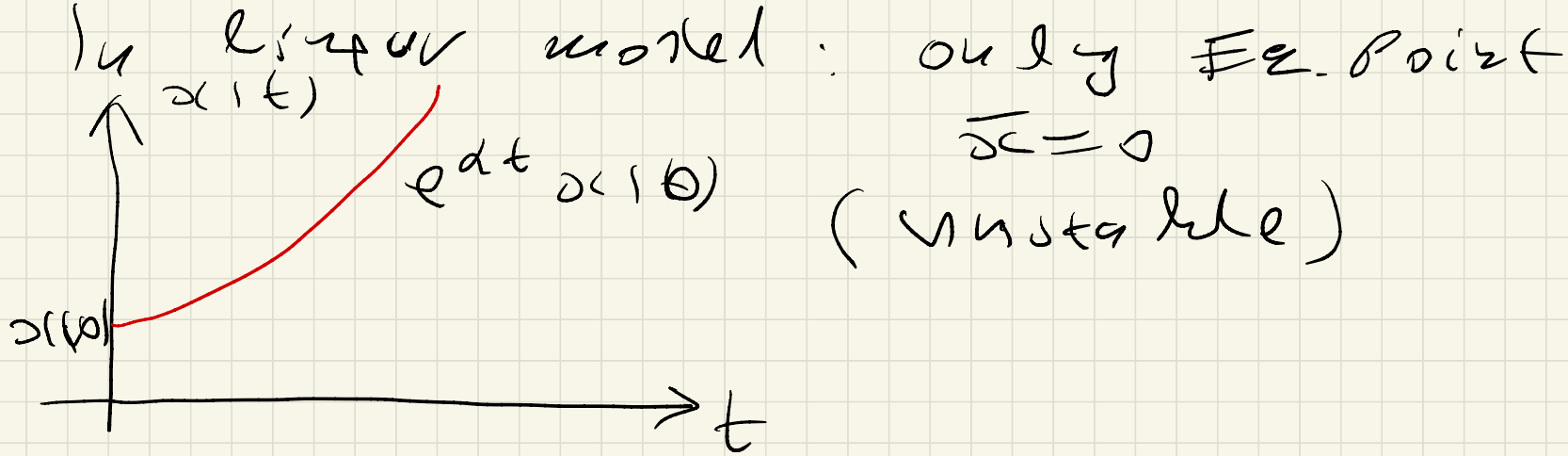
Compare with linear model  
(constant growth function)

$$\dot{x} = \alpha \cdot x \Rightarrow x(t) = e^{\alpha t} x(0)$$

For  $\alpha > 0 \Rightarrow x(t)$  monotonically increasing  
 $x(t) \xrightarrow{t \rightarrow \infty} \infty$

(11)

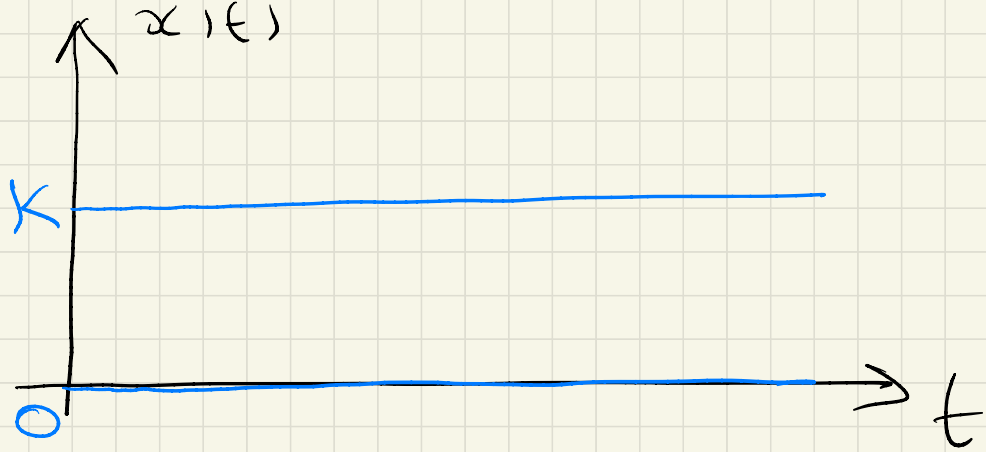




Nonlinear model: Eq. points:

$$0 = \alpha \cdot \left(1 - \frac{\bar{x}}{K}\right) \bar{x} \Rightarrow \boxed{\bar{x}_1 = 0; \bar{x}_2 = K}$$

Two Eq. points: zero and K  
carrying capacity



Next time: we'll see that nonlinear model saturates the growth of population (reasonable) and we'll "study" features of the response (e.g. stability of  $\pm$  points)