


Lecture 2

011123

Last time:

- Class logistics
- State-space Models ; $\frac{dx}{dt} = f(x, u)$
- Examples: pendulum; logistic equation

Today:

- Equilibrium points
- Linearization
- Flows of first order (scalar) systems
 $x(t) \in \mathbb{R}$

Unforced nonlinear system (no input)

$$\frac{dx}{dt} = f(x) \quad ; \quad (*)$$

nonlinear
function

t : time

$$x(t) \in \mathbb{R}^n$$

$$x_i(t) \in \mathbb{R}$$

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

$$x(t) = \begin{bmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{bmatrix}$$

IF Note: extends to matrices

$$\dot{X} = F(X) \quad ; \quad X \in \mathbb{R}^{m \times n}$$

Eq. Points: constant trajectories

i.e. solutions to (*) that do NOT
change with time

(2)

$x(t) = \bar{x} = \text{const.}$ for all t

Eq. points

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

$x(t_0) = \bar{x} \Rightarrow x(t) = \bar{x}$ for all t

In particular $\frac{d\bar{x}}{dt} \equiv 0 \quad \textcircled{+} \Rightarrow$

$$\frac{d\bar{x}}{dt} = 0 = f(\bar{x}) \Rightarrow$$

We can determine Eq. points by solving nonlinear equation

$$f(\bar{x}) = 0$$

Ex: Logistic Equation

$$\dot{x} = \alpha \left(1 - \frac{x}{K} \right) x ; x(t) \in \mathbb{R}_+$$

$f(x)$

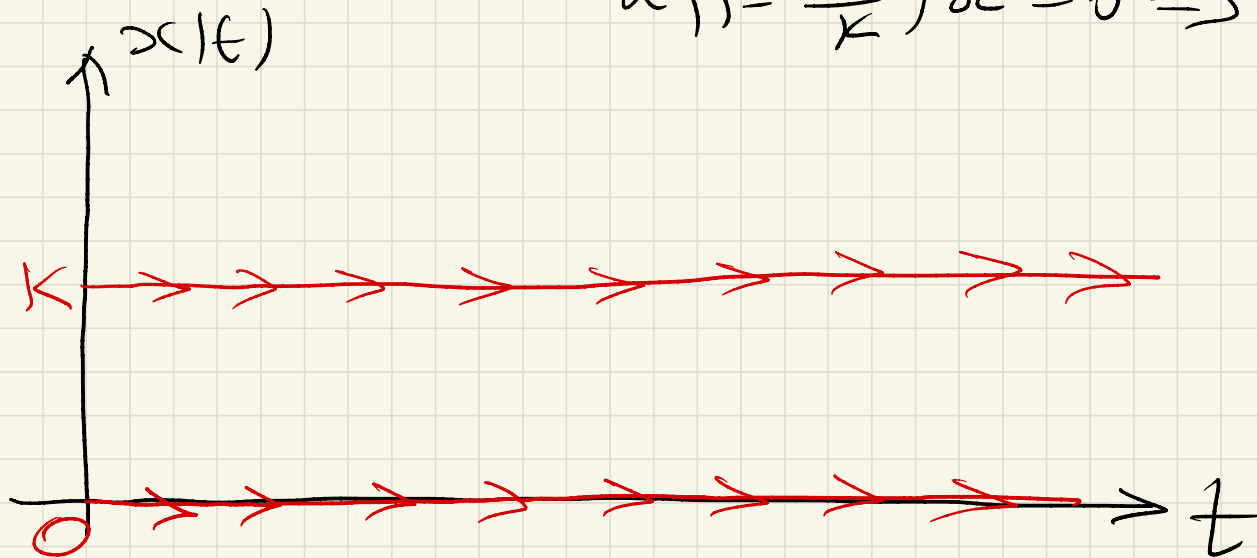
$\alpha, K > 0$ (constant parameters)

Eq. points: $f(\bar{x}) = 0$

$$\alpha \left(1 - \frac{\bar{x}}{K} \right) \bar{x} = 0 \Rightarrow$$

$$\bar{x}_1 = 0$$

$$\bar{x}_2 = K$$



Q: What happens if $x(t_0) \neq \bar{x}$, or $\frac{\partial x}{\partial x_2}$

If we had a linear model

$$\dot{x} = \alpha x \Rightarrow \bar{x} = 0$$

$$x(t) = e^{\alpha t} x(t_0) \xrightarrow{\alpha > 0} x(t) \uparrow$$

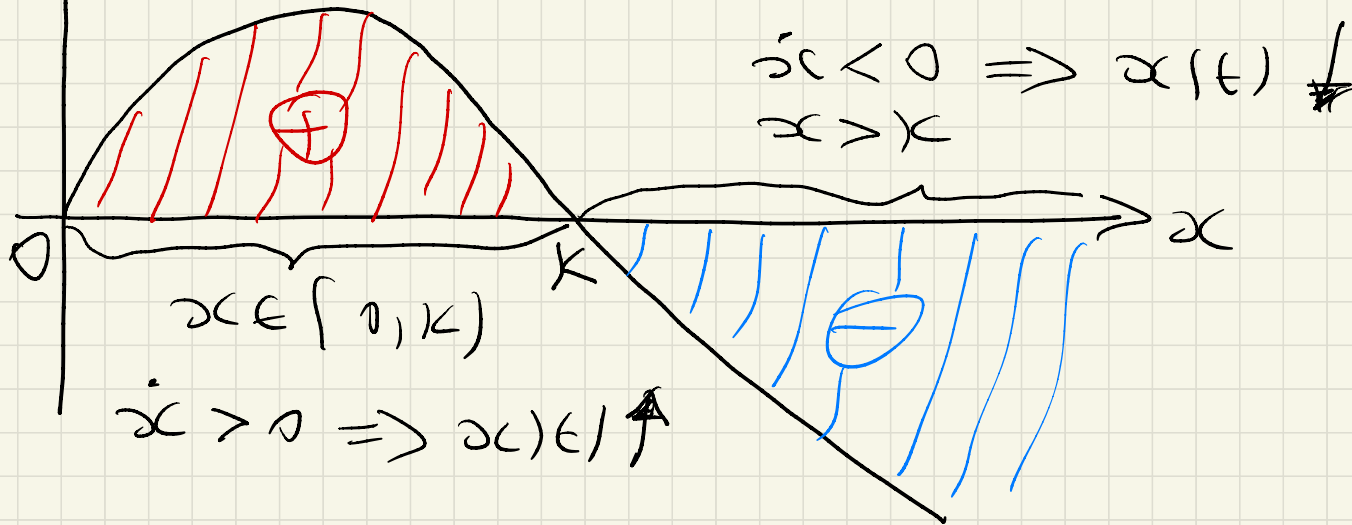
$$\lim_{t \rightarrow \infty} x(t) = \infty$$

In nonlinear model, there is another eq. point that influences dynamical properties of our system (dependence of solutions on time)

Useful tool (for scalar systems)

Plot of $f(x)$ v.s. x

$$\dot{x} = f(x) = \alpha \left(1 - \frac{x}{K} \right) x = \alpha \left(x - \frac{x^2}{K} \right)$$



Based on this:

Qualitatively we should have:



Note: We have derived these conclusions without solving differential eq.

We'll formalize this approach in the course

Quantitative analysis necessary to figure out rate of convergence; again, will do this without explicitly solving ODEs.
(Lyapunov-based analysis: later in the course)

Linearization: tool for local analysis around certain trajectory (does NOT have to be an eq. point) ("like in small")

Given $\dot{x} = f(x)$ and solution \bar{x} [can be time dependent]

decompose $x(t)$ as: \longrightarrow ⑧

$$x(t) = \bar{x}(t) + \tilde{x}(t) \quad \dots (1)$$

↓
gives trajectory
(e.g. eq point)

↪ perturbation;
fluctuation;
deviation
(around \bar{x})

Substitute (1) to $(*)$

$$\dot{x}(t) = \dot{\bar{x}}(t) + \dot{\tilde{x}}(t) \stackrel{(*)}{=} f(\bar{x} + \tilde{x}) \Rightarrow$$

$$\dot{\tilde{x}}(t) = f(\bar{x} + \tilde{x}) - \dot{\bar{x}}(t) \stackrel{(*)}{=} f(\bar{x}) \Rightarrow$$

Equation for $\tilde{x}(t)$ gives us (9)

Fluctuation Dynamics:

$$\frac{d\tilde{x}}{dt} = f(\bar{x} + \tilde{x}) - f(\bar{x})$$



Eg. for \tilde{x} in which \bar{x} is a coefficient

Note: and is \Leftrightarrow to $\textcircled{*}$

- $\textcircled{+}$ holds globally; i.e. for $\tilde{x}(t) \in \mathbb{R}^n$
(its approximation made yet)
- If $\bar{x} = \bar{x}(t)$ then $\textcircled{*}$ is time-varying
even if original system was
time-invariant

• If \bar{x} is an eq. point then \boxed{f} is time invariant if $\dot{x} = f(x)$ is time invariant

[Recall : $\dot{x} = f(x) : T1. \quad \ddot{x} = \alpha \left(1 - \frac{x}{x_c} \right) x$
 $\dot{x} = f(x, t) : T2. \quad \dot{x} = \alpha(t) \left(1 - \frac{x}{x_c} \right) x$]

If we use Taylor series expansion of f around \bar{x} , we have $O(\|\tilde{x}\|^2)$

$$f(\bar{x} + \tilde{x}) = f(\bar{x}) + \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}} \cdot \tilde{x} + \text{H.O.T.}$$

$$f(\bar{x} + \tilde{x}) - f(\bar{x}) \approx \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}} \cdot \tilde{x}$$

①①

Thus, linearization around \bar{x} yields:

$$\dot{\tilde{x}} = \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}} \tilde{x}$$

obtained by neglecting H.O.T. in \tilde{x}

Holds for arbitrary solution $\bar{x}(t)$ of $\dot{x} = f(x)$: does NOT have to be an eq. point

Ex: Logistic Eq.

$$f(x) = \alpha \left(1 - \frac{x}{K} \right) x = \alpha \left(x - \frac{x^2}{K} \right)$$

$$\frac{\partial f}{\partial x} = \alpha \left(1 - \frac{2x}{K} \right) \Rightarrow \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}} = \alpha \left(1 - \frac{2\bar{x}}{K} \right)$$

(2)

Clearly, Jacobian depends on \bar{x}
 In particular for $\bar{x}_1 = 0$ & $\bar{x}_2 = K$
 we have:

$$A_1 = \left. \frac{\partial f}{\partial x} \right|_{x=\bar{x}_1=0} = \alpha \cdot \left(1 - \frac{2 \cdot 0}{K} \right) = \alpha > 0$$

$$A_2 = \left. \frac{\partial f}{\partial x} \right|_{x=K} = \alpha \cdot \left(1 - \frac{2 \cdot K}{K} \right) = -\alpha < 0$$

$$\dot{\tilde{x}} = \begin{cases} \alpha \tilde{x} & \text{around } \bar{x}_1 = 0 \\ -\alpha \tilde{x} & \text{around } \bar{x}_2 = K \end{cases}$$

$$x(t) = \begin{cases} e^{\alpha t} \tilde{x}(0) & \text{around } \bar{x}_1 = 0 \\ e^{-\alpha t} \tilde{x}(0) & \text{around } \bar{x}_2 = K \end{cases} \quad (13)$$

Thus, small perturbations around $\bar{x}_1 = 0$ are going to grow [$\bar{x}_1 = 0$ is unstable]
whereas small perturbations around $\bar{x}_2 = k$ are going to decay [$\bar{x}_2 = k$ is locally asymptotically stable]

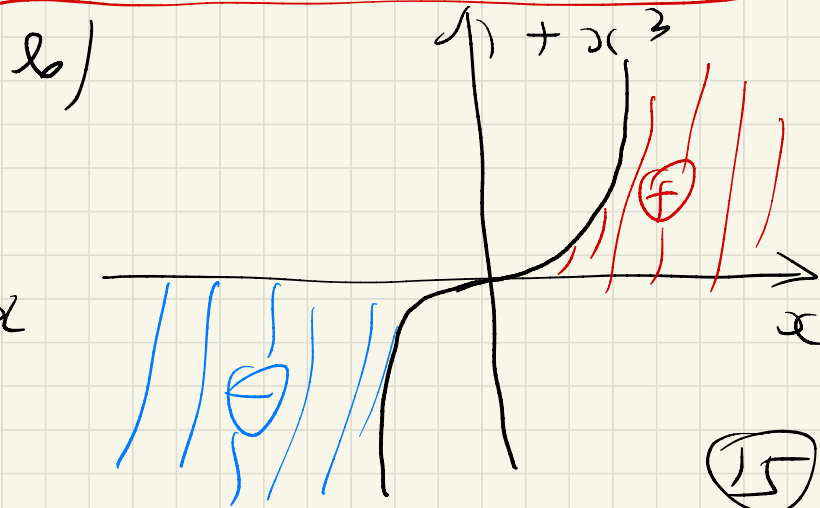
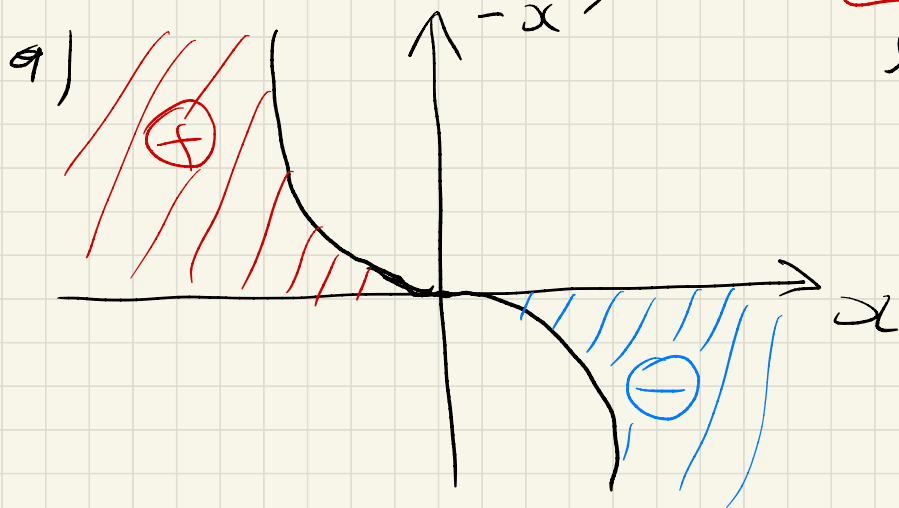
Later, we'll prove that stability of linearization around eq. point \bar{x} guarantees local asymptotic stability of \bar{x} for nonlinear system.

Ex: $\left. \begin{array}{l} a) \dot{x} = -x^3 \\ b) \dot{x} = +x^3 \end{array} \right\} \Rightarrow \bar{x} = 0 \text{ is the unique eq. point}$

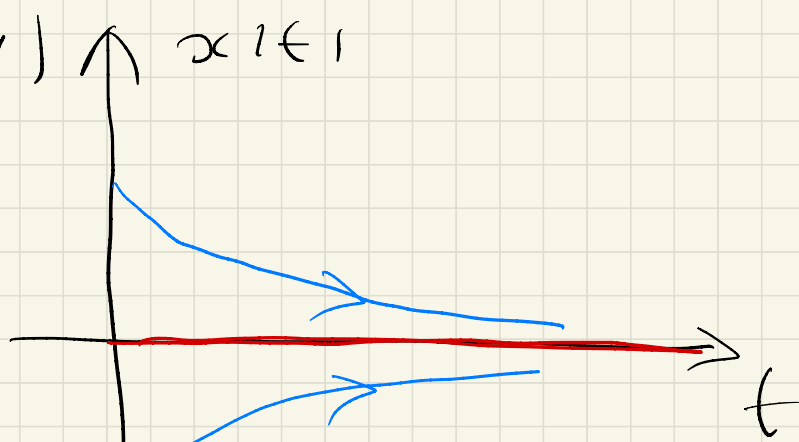
$$\left. \frac{\partial f}{\partial x} \right|_{\bar{x}=0} = \mp 3\bar{x}^2 = 0$$

$$\Rightarrow \dot{\tilde{x}} = 0 \cdot \tilde{x} \Rightarrow$$

$$\tilde{x}(t) = x(t_0) = \text{const}$$



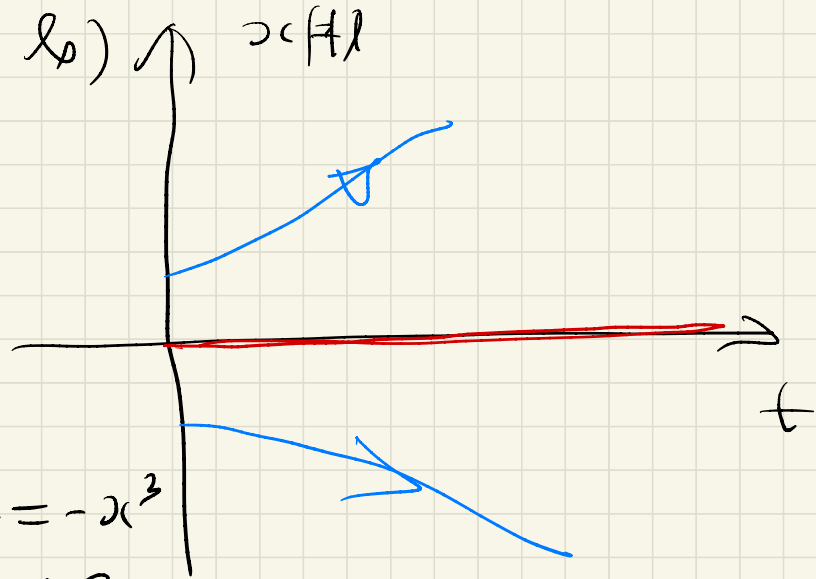
a) $x \in \mathbb{R}$



$$\dot{x} = 0 \text{ at } \dot{x} = -x^3$$

globally asymptotically
stable

b) $x \in \mathbb{R}$



✓ UNSTABLE

Note: Linearization CANNOT be
used to derive these conclusions
[wplm formalize this]

Note: For DT systems

$$x^+ = f(x) \quad ; \quad x^+ = x(t+1) = f(x(t))$$

$t = 0, 1, 2, \dots$

Eq. points: [E.g. implicit NN:]

$$\bar{x} = f(\bar{x}) \quad \left[\quad \bar{x} = \underline{\Phi}(Ax + b) \right]$$

Fluctuation Dynamics

$$\bar{x}^+ + \tilde{x}^+ = f(\bar{x} + \tilde{x})$$

$$\tilde{x}^+ = f(\bar{x} + \tilde{x}) - f(\bar{x})$$

Ex: $\bar{x} = \sin(\omega t)$; $x(t) \in \mathbb{R}$

$$\sin(\bar{x}) = 0 \Rightarrow \bar{x} = k\pi ; k = 0, \pm 1, \pm 2, \dots$$

Exercise: Plot $x(t)$ v.s. t
 (get insight from visualizing
 $\sin(x)$ v.s. x)

Linearize around $\bar{x} = k\pi$

