

Autonomous Linear Dynamical Systems

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MIDTERM

~ OUT AFTER CLAS ON THURSDAY

COME BACK ON FRIDAY @ 2PM.

RULES OF MIDTERM — OPEN BOOK, NOTES, MATLAB
WEBSITES, WIKIPEDIA...

EXCEPT NO HUMAN/AI HELP

NO PUBLISHED SOLUTIONS



Autonomous Linear Dynamical Systems

$$\dot{x} = Ax + Bu$$

- Autonomous Linear Dynamics Systems
- Examples
- Higher order systems
- Linearization near equilibrium point
- Linearization along trajectory



Autonomous Linear Dynamical Systems (1.4)

continuous time autonomous linear dynamical system

$$\dot{x} = Ax + B \overset{\text{0. column}}{\mu}$$

C.T.S. $A \neq f(t)$

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

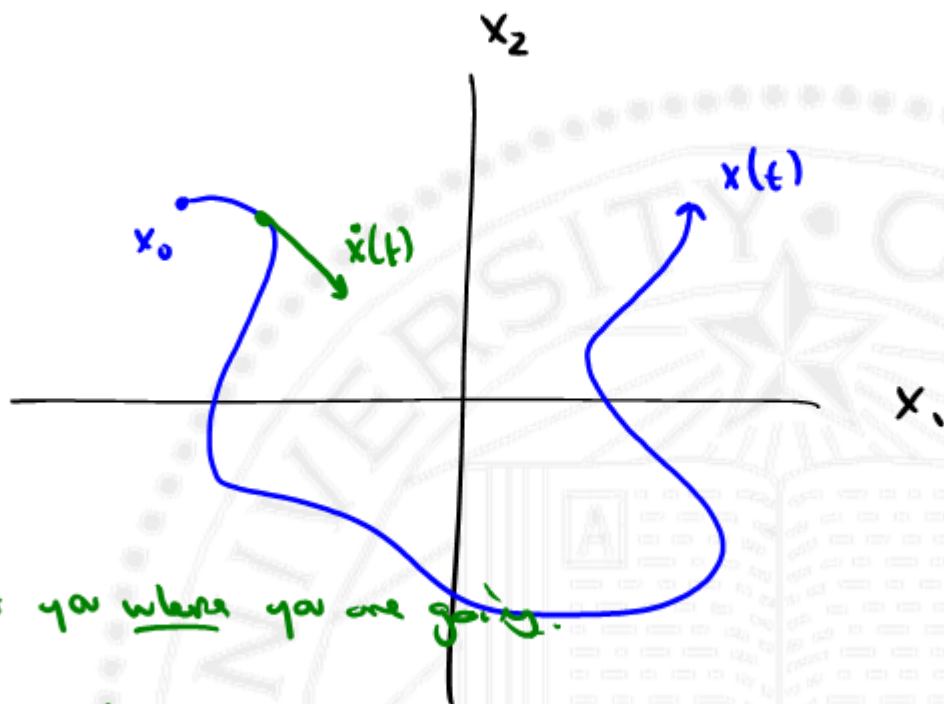
$x(t) \in \mathbb{R}^n$ called "state" x_i called i^{th} state

n - state dimension, order, number of states.

$A(t) \triangleq$ dynamics matrix $\in \mathbb{R}^{n \times n}$



Autonomous Linear Dynamical Systems (2.4)



A tells you where you are going.

$$\dot{x} = \frac{dx}{dt}$$

$x(t)$ trajectory

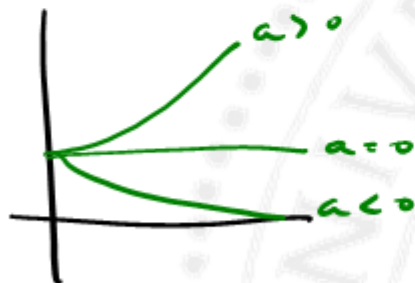
$\dot{x}(t)$ tangent to trajectory



Autonomous Linear Dynamical Systems (3.4)

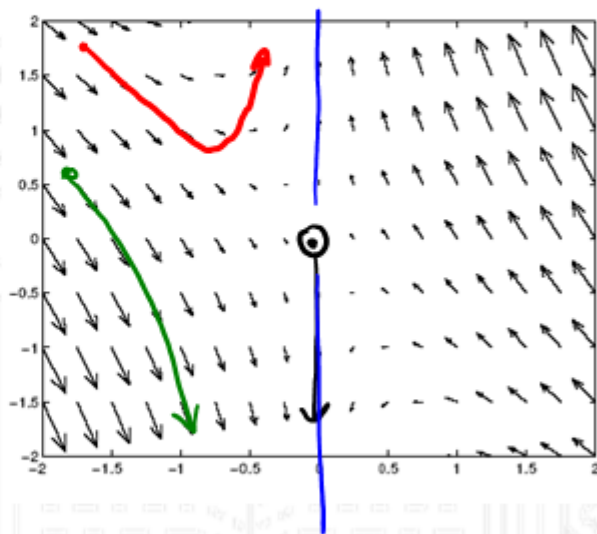
$$\dot{x} = \begin{bmatrix} -1 & 0 \\ 2 & 1 \end{bmatrix} x$$

NON-OSCILLATORY BEHAVIOR



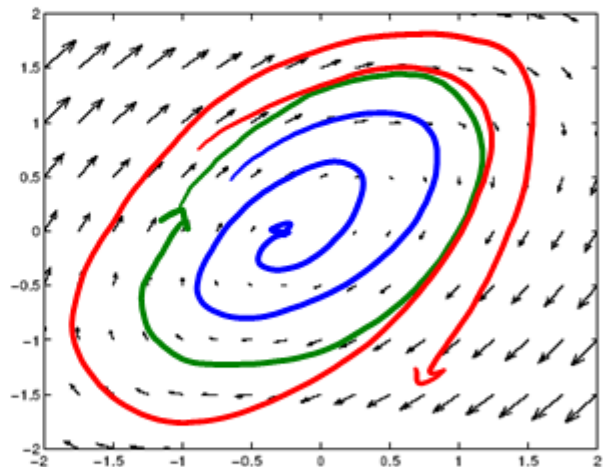
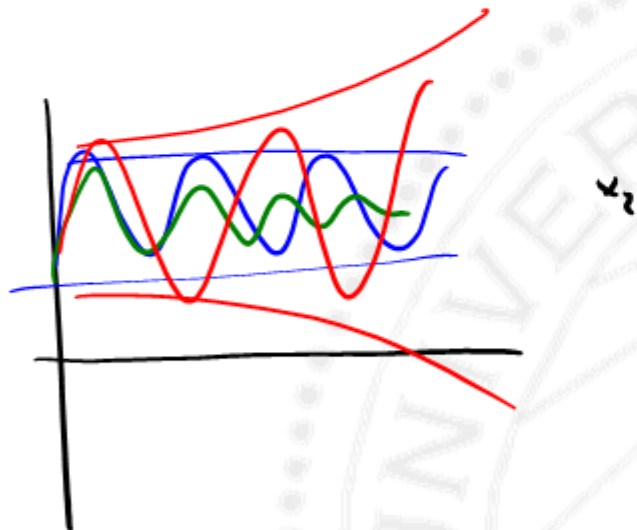
$$\dot{x} = ax \rightarrow x(t) = x_0 e^{at}$$

VECTOR FIELD
PHASE PLANE \mathbb{R}^2



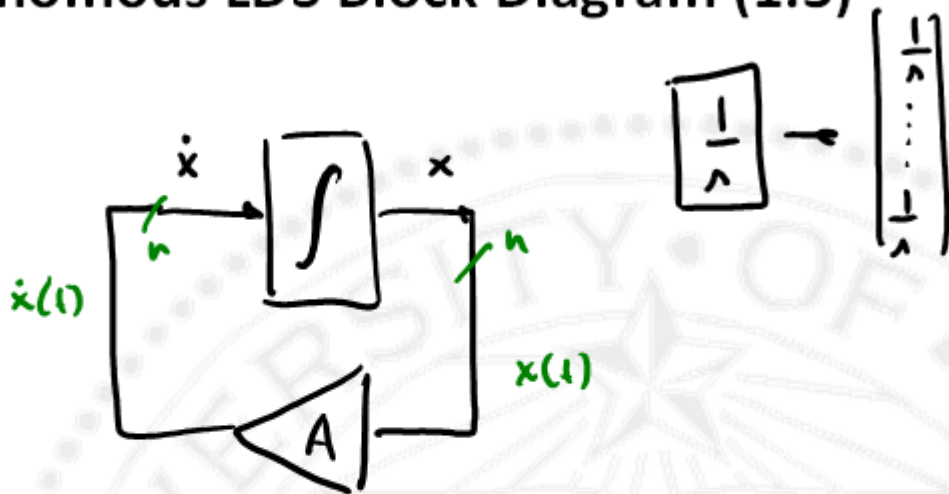
Autonomous Linear Dynamical Systems (4.4)

$$\dot{x} = \begin{bmatrix} -0.5 & 1 \\ -1 & 0.5 \end{bmatrix} x$$



Autonomous LDS Block Diagram (1.3)

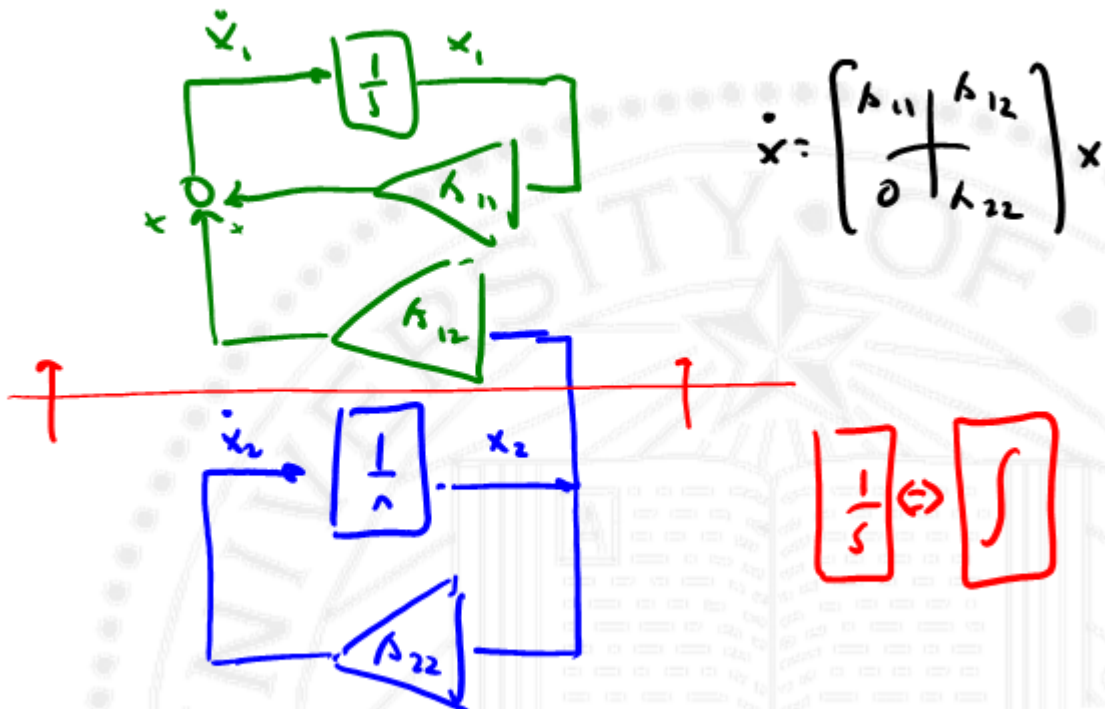
$$\dot{x} = Ax$$



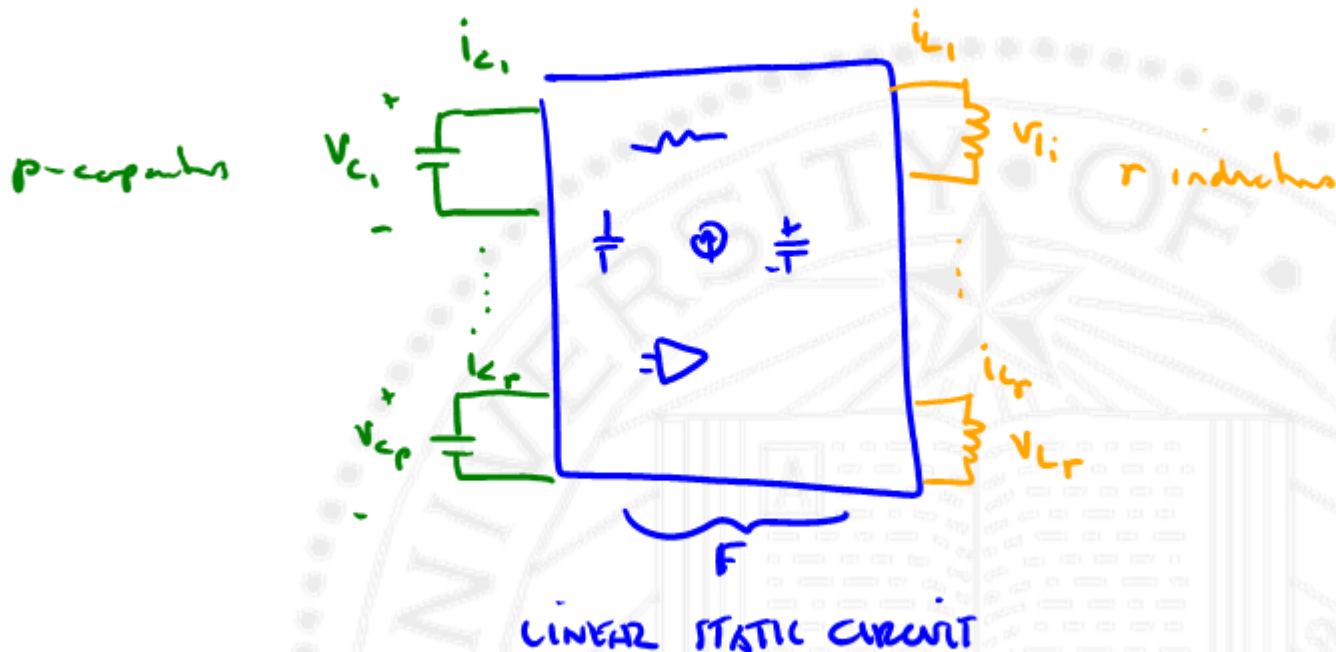
coupling comes from Dynamics Matrix, "A".



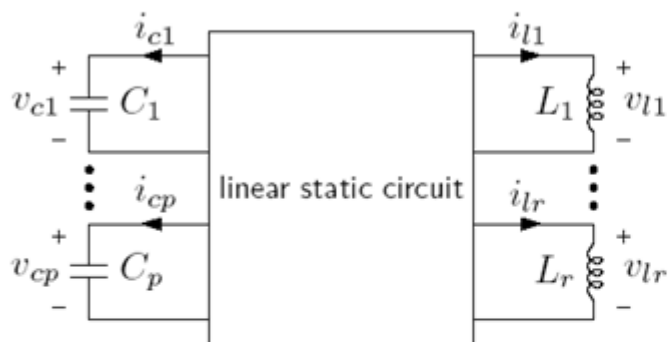
Autonomous LDS Block Diagram (2.3)



Autonomous LDS Block Diagram (3.3)



Linear Circuit Example (1.2)



$$\begin{bmatrix} i_c \\ v_l \end{bmatrix} = F \begin{bmatrix} v_c \\ i_l \end{bmatrix}$$

$$C \frac{dv_c}{dt} = i_c$$

$$L \frac{di_l}{dt} = v_l$$

$$C = \begin{bmatrix} C_1 & & 0 \\ & \ddots & \\ 0 & & C_p \end{bmatrix}$$

$$L = \begin{bmatrix} L_1 & & 0 \\ & \ddots & \\ 0 & & L_r \end{bmatrix}$$



Linear Circuit Example (2.2)

$$x = \begin{bmatrix} \psi_c \\ i_c \end{bmatrix} \quad \dot{x} = \begin{bmatrix} \dot{\psi}_c \\ \dot{i}_c \end{bmatrix}$$

$$\dot{x} = Ax$$

$$\dot{x} = \underbrace{\begin{bmatrix} \dot{c}' & 0 \\ 0 & \dot{c}' \end{bmatrix}}_A F x$$

spica

A



Chemical Reactions (1.3)

chemical reaction involving n chemicals

x_i : concentration of chemical i

$$\frac{dx_i}{dt} = a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n$$

$A \Rightarrow$ usually very sparse.



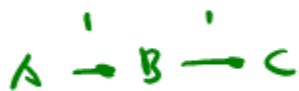
$$\dot{x} = \begin{bmatrix} -k_1 & 0 & 0 \\ k_1 & -k_2 & 0 \\ 0 & k_2 & 0 \end{bmatrix} x$$



Chemical Reactions (2.3)

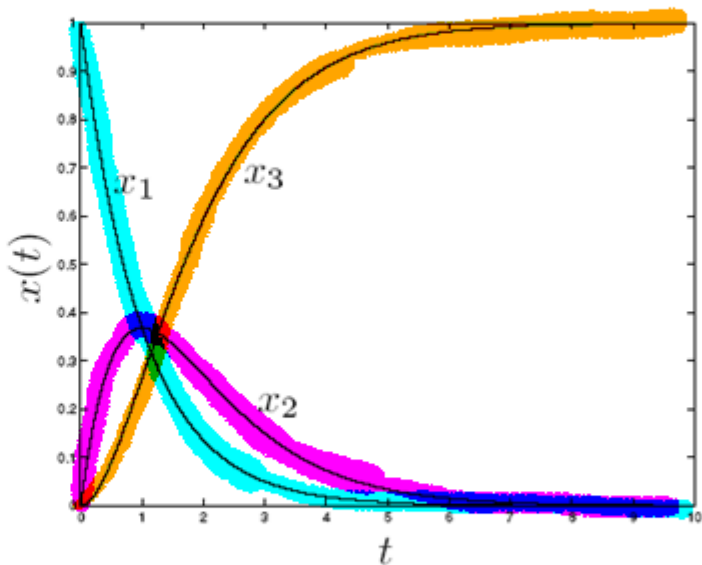
$$k_1 = k_2 = 1$$

$$x(\omega) = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$



$$\dot{x} = \begin{bmatrix} -1 & 0 & 0 \\ 1 & -1 & 0 \\ 0 & 1 & 0 \end{bmatrix} x + \begin{matrix} - \\ + \\ 0 \end{matrix}$$

$$\dot{x}_2 = x_1 - x_2$$



Finite-State Discrete-Time Markov Chain (1.3)

$z(k) \in \{1 \dots n\}$ random sequence with

$$\text{Prob}(z(k+1) = i \mid z(k) = j) \triangleq P_{ij}$$

$$P \in \mathbb{R}^{n \times n}$$

matrix of transition probabilities
jump markov state

$$\text{Prob}(z(k) = 1, 2, \dots) = [1 \ 1 \ 1 \ 0 \ \dots] p(k)$$

$$p(k+1) = P p(k) \rightarrow p(k+1) \neq p(k) P^T$$



Finite-State Discrete-Time Markov Chain (2.3)

P is sparse

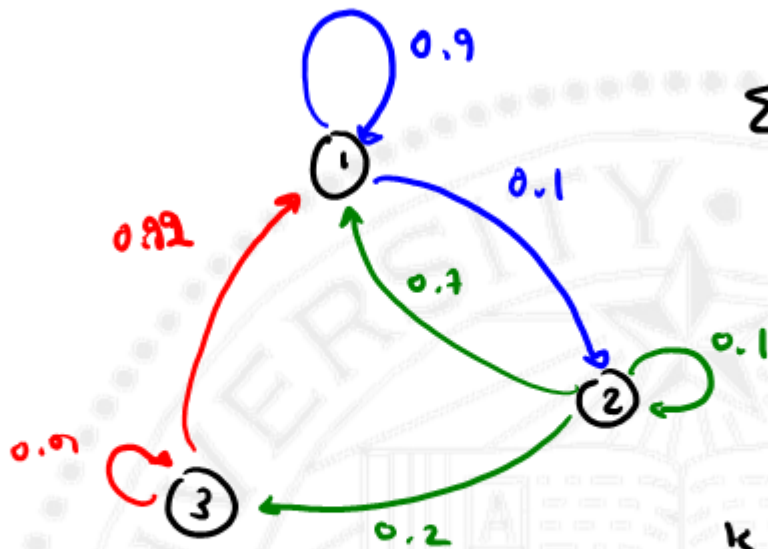
nodes are all states

edges show directed probabilities



Finite-State Discrete-Time Markov Chain (3.3)

- ① "ok"
- ② "down"
- ③ "under repair"



$$\sum_{col} P = 1.$$

$$P_{k+1} = \begin{bmatrix} 0.9 & 0.7 & 0.99 \\ 0.1 & 0.1 & 0 \\ 0 & 0.2 & 0.01 \end{bmatrix} P_k$$

$$\frac{P^k p(0)}{P^k} = \text{prob.}$$

Markov chain
Mixing



Numerical Integration of Continuous System (1.3)

compute an approximate solution of $\dot{x} = Ax$ $x(0) = x_0$

small time step: h

x does not change very much over time interval h .

$$x(t+h) \approx x(t) + h \dot{x}(t) = x(t) + hAx(t) \\ = [I + Ah]x(t)$$

$$x(kh) \approx [I + Ah]^k x(0) \quad \leftarrow \text{Forward euler}$$

$$x(t+h) = [I - Ah]^{-1} x(t) \quad \leftarrow \text{Backward euler} \\ \text{"works"}$$

Runga-Kutta



Higher Order LDS (1.3)

$$x^{(k)} = A_{k-1} x^{(k-1)} + \dots + A_1 x^{(1)} + A_0 x \quad x^{(1)} \in \mathbb{R}^n$$

$\dot{x} = Ax$ embedded $x^{(n)}$ n th derivative of x

$$x^{(0)} = x$$

$$x^{(1)} = \dot{x}$$

$$x^{(2)} = \ddot{x}$$

$$x^{(3)} = \overset{\cdot}{\ddot{x}}$$

\vdots

$$z = \begin{bmatrix} x \\ \dot{x} \\ \ddot{x} \\ \vdots \\ x^{(k-1)} \end{bmatrix} \in \mathbb{R}^{nk}$$

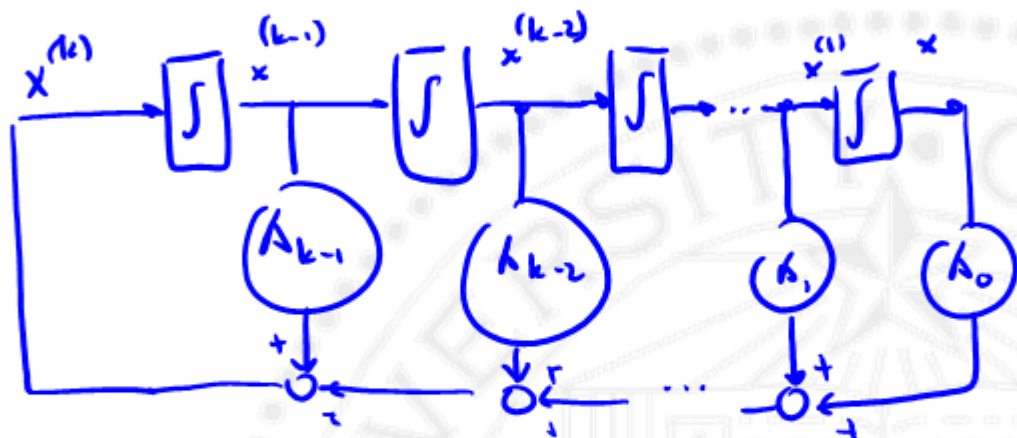
← shift matrix

$$\dot{z} = \begin{bmatrix} 0 & I & 0 & \dots & 0 \\ 0 & 0 & I & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & I \\ A_0 & A_1 & \dots & \dots & A_{k-1} \end{bmatrix} z$$

← Block companion form.



Higher Order LDS (2.3)



chain of integrators



Example: Mechanical Systems

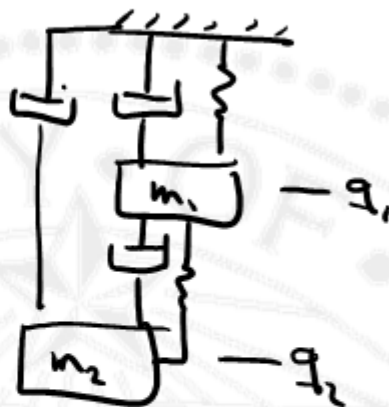
$$M\ddot{q} + D\dot{q} + Kq = 0$$

$q(t) \in \mathbb{R}^k$ general displacements

$M \triangleq$ mass matrix $\begin{pmatrix} m_1 \\ \vdots \\ m_n \end{pmatrix}$

$K \triangleq$ stiffness matrix

$D \triangleq$ damping matrix



$$x \triangleq \begin{bmatrix} q \\ \dot{q} \end{bmatrix} \quad \dot{x} = \begin{bmatrix} \dot{q} \\ \ddot{q} \end{bmatrix} = \begin{bmatrix} 0 & I \\ -\tilde{r}^k & -\tilde{m}^d \end{bmatrix} \begin{bmatrix} q \\ \dot{q} \end{bmatrix}$$

λ x = Ax



Linearization near Equilibrium (1.3)

Non-linear time invariant differential equation.

$$\dot{x} = f(x) \quad f: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

x_e equilibrium point $f(x_e) = 0$ $x(t) = x_e \neq t$.

$x(t)$ near x_e $\dot{x}(t) = f(x(t)) = f(x_e) + \underbrace{Df(x_e)}_{\text{Jacobian of } f} (x(t) - x_e)$

$$\delta x \triangleq x(t) - x_e$$

$$\underline{\dot{\delta x}} \approx \underbrace{Df(x_e)}_F \underline{\delta x}$$

Small signal model
linearized dynamics
perturbed equilibria

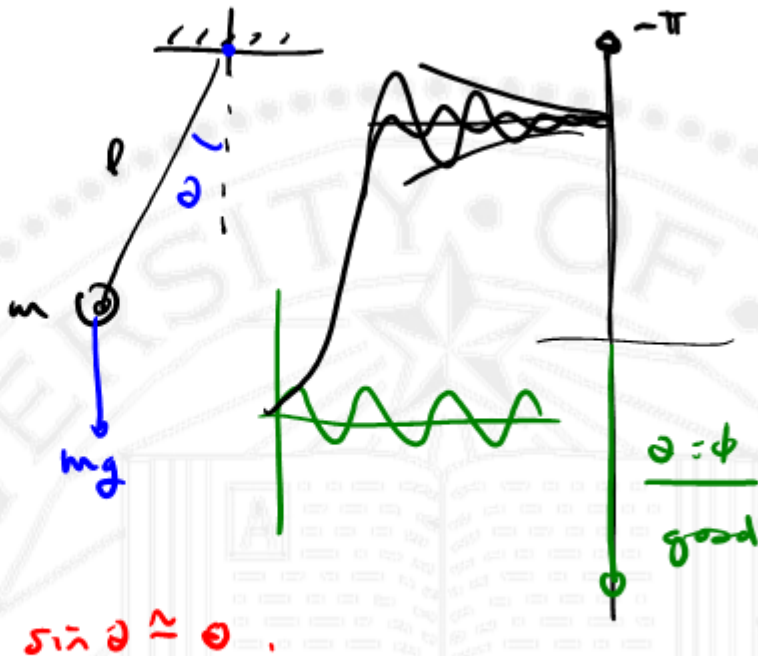


Linearization near Equilibrium (2.3)

$$m l^2 \ddot{\theta} = -m g l \sin \theta$$

$$\begin{bmatrix} \dot{\theta} \\ \ddot{\theta} \end{bmatrix} = \begin{bmatrix} \dot{\theta} \\ -\frac{g}{l} \sin \theta \end{bmatrix}$$

$$\delta \dot{x} = \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & 0 \end{bmatrix} \delta x$$



Does Linearization "work"? (1.2)

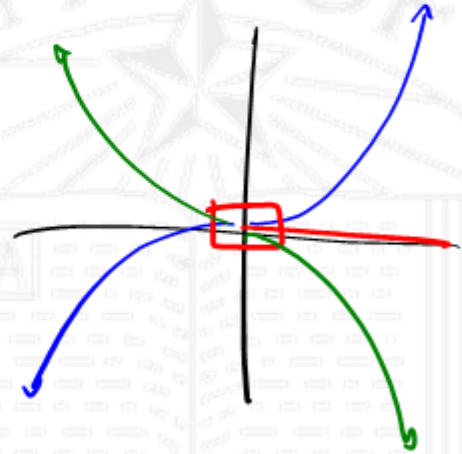
$$\dot{x} = -x^3 \quad x_e = 0 \quad x(0) > 0$$

linearized system: $\delta \dot{x} = 0$ $x(1) = \sqrt{\frac{1}{x_0^2} + 2t}$

$$\dot{z} = z^3 \quad z(0) > 0$$

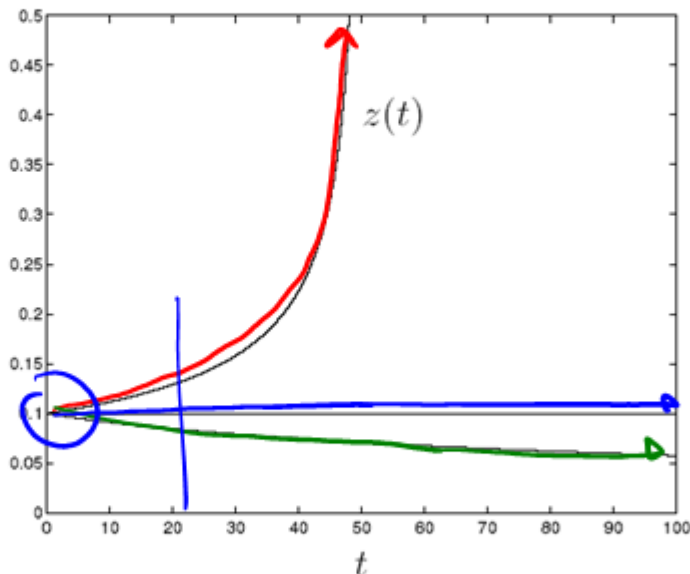
$$z(1) = \sqrt{\frac{1}{z_0^2} - 2t}$$

finite escape time $t = \frac{1}{2z_0^2}$



Does Linearization “work”? (2.2)

$$x(0) = y(0) = 0.1$$



$$\delta x(t) = \delta z(t)$$
$$x(t)$$



Linearization along Trajectory (1.3)

$$x_{\text{drag}}: \mathbb{R}_+ \rightarrow \mathbb{R}^n$$

$$\dot{x}_{\text{drag}}(t) = f(x_{\text{drag}}(t), t)$$

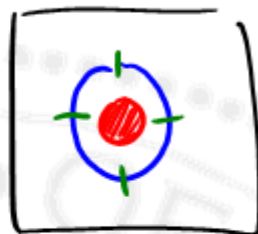
$x(t)$ some other trajectory $\dot{x}(t) = f(x(t), t)$

$x(t)$ is near $x_{\text{drag}}(t)$

$$\frac{d}{dt}(x - x_{\text{drag}}) = f(x, t) - f(x_{\text{drag}}, t) \approx D_x f(x_{\text{drag}}, t)(x - x_{\text{drag}})$$

$$\dot{\delta x} \approx D_x f(x_{\text{drag}}, t) \delta x$$

DRAW FREE



GP-B



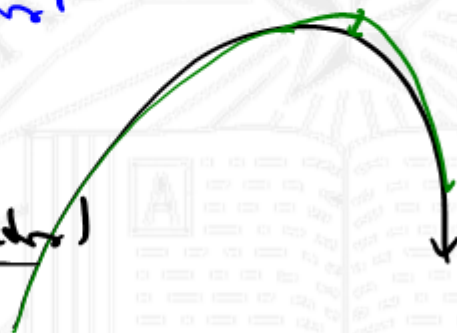
Linearization along Trajectory (2.3)

Linearized or variational trajectory

orbital transfer

optimal control

neighbors optimal control



Linearization along Trajectory (3.3)

linearized oscillator (OE)

x_{drag} T -periodic solution NLDE

$$\dot{x}_{drag}(t) = f(x_{drag}, t)$$

$$x_{drag}(t+T) = x_{drag}(t)$$

$$\dot{\delta x} = A(t) \delta x$$

T -periodic system

$$X(t) = Df(x_{drag}, t)$$

"Floquet System"



Questions?

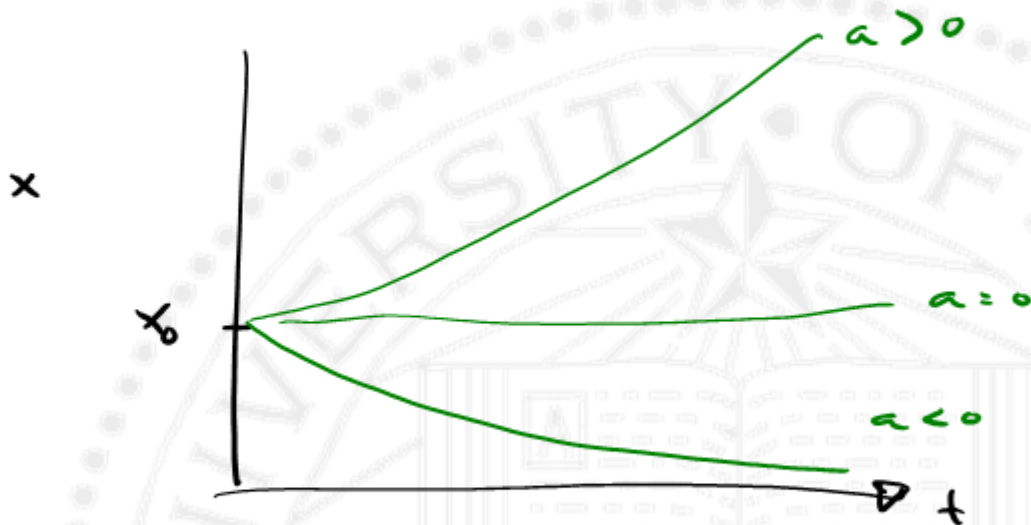


Solution via Laplace Transform and Matrix Exponential

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Winter 2016



$$\dot{x} = ax \quad \leftarrow \text{scalar} \quad x(t) = x_0 e^{at}$$



$$\underline{\dot{x}} = A \underline{x} \rightarrow \underline{x}(t) = e^{At} x_0$$



Solution via Laplace Transform and Matrix Exponential

- Laplace Transform
- Solving $x' = Ax$ via Laplace Transform
- State transition matrix
- Matrix Exponential
- Qualitative behavior and Stability



Laplace Transform of a Matrix (1.3)

$$\mathcal{L}\{x(t)\} \triangleq \int_0^{\infty} x(t) e^{-st} dt \triangleq X(s) \quad \begin{array}{l} x: \mathbb{R}_+ \rightarrow \mathbb{R} \\ X: \mathbb{C} \rightarrow \mathbb{C} \end{array}$$

$$\mathcal{L}\{\dot{x}(t)\} = sX(s) - x_0$$

$$\mathcal{L}\{\dot{x} = ax\} = sX(s) - x_0 = aX(s) \rightarrow (s-a)X(s) = x_0$$

$$X(s) = \frac{x_0}{s-a}$$

$$x(t) = \mathcal{L}^{-1}\left\{\frac{x_0}{s-a}\right\} = \underline{x_0 e^{at}}$$



Laplace Transform of a Matrix (2.3)

$z: \mathbb{R}_+ \rightarrow \mathbb{R}^{p \times q}$ \leftarrow trajectory, matrix valued signal

$Z = \mathcal{L}(z)$ where $Z: D \subseteq \mathbb{C} \rightarrow \mathbb{C}^{p \times q}$

$$Z(s) = \int_0^{\infty} e^{-st} z(t) dt$$

\uparrow complex

$$Z(s) = \int_0^{\infty} z(t) e^{-st} dt$$

$D \hat{=}$ domain of convergence

D includes at least $\{s \mid \operatorname{Re}(s) > a\}$

where $a: |z_{ij}(t)| \leq \alpha e^{at} \quad t > 0 \quad i=1 \dots p \quad j=1 \dots q$

