

EE16B

Designing Information Devices and Systems II

Lecture 7B
Cont. stability of Linear State Models
Controllability

Today

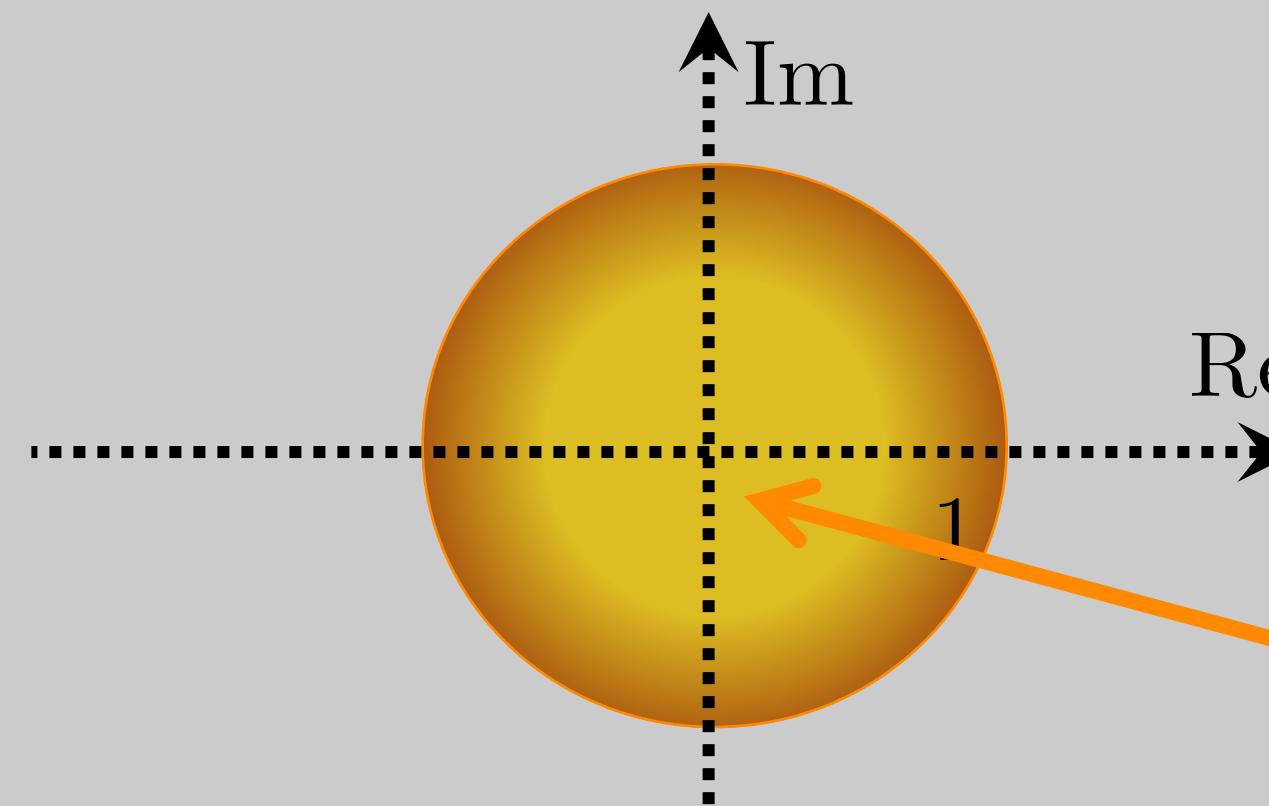
- Last time:
 - Derived stability conditions for disc. and cont. systems
 - Easy to analyze using eigenvalues
- Today:
 - Eigenvalues can predict system behaviour
 - Envelope (decay) and Oscillation (frequency)
 - Controllability of systems

HOW **NOT** TO LAND AN ORBITAL ROCKET BOOSTER

Stability -- Summary

Discrete-Time

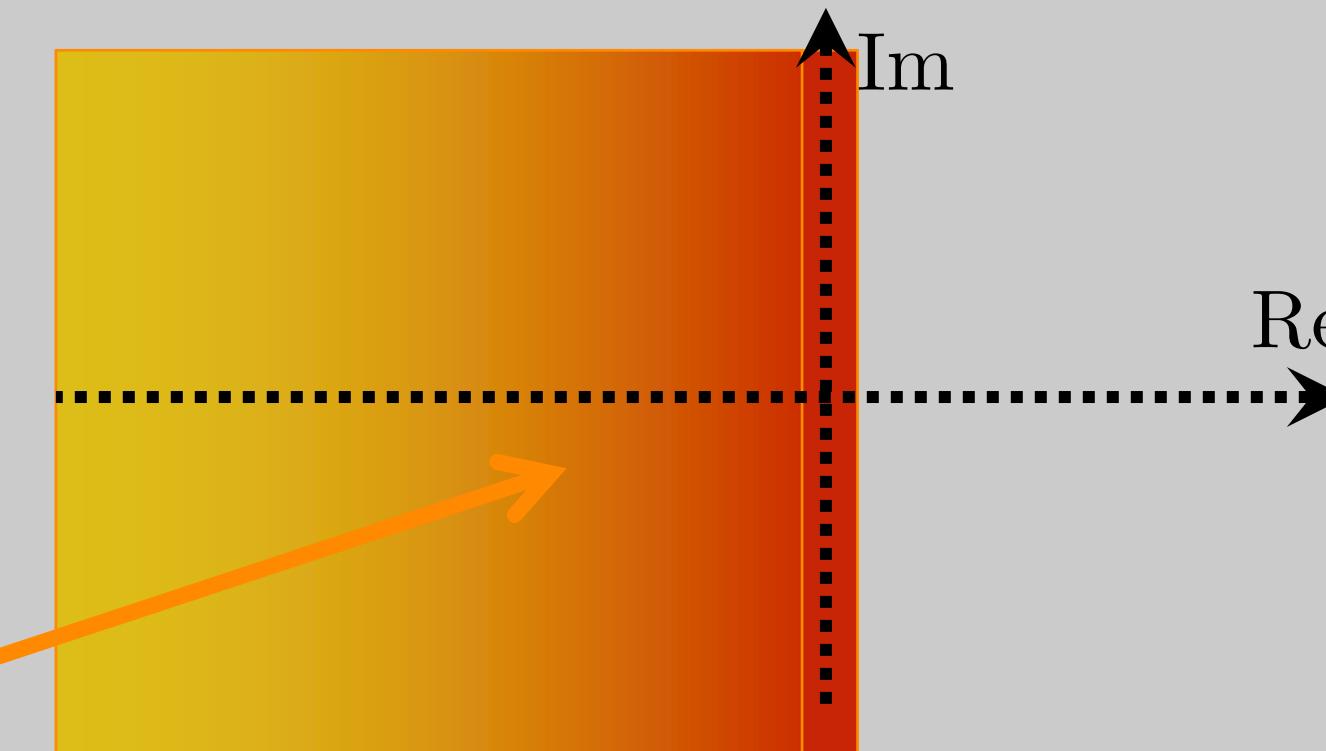
$$|\lambda_i(A)| < 1$$



Stable regions

Continuous-Time

$$\text{Real}\{\lambda_i(A)\} < 0$$



Stay away from boundaries! System uncertainty can
Move you over to unstable region

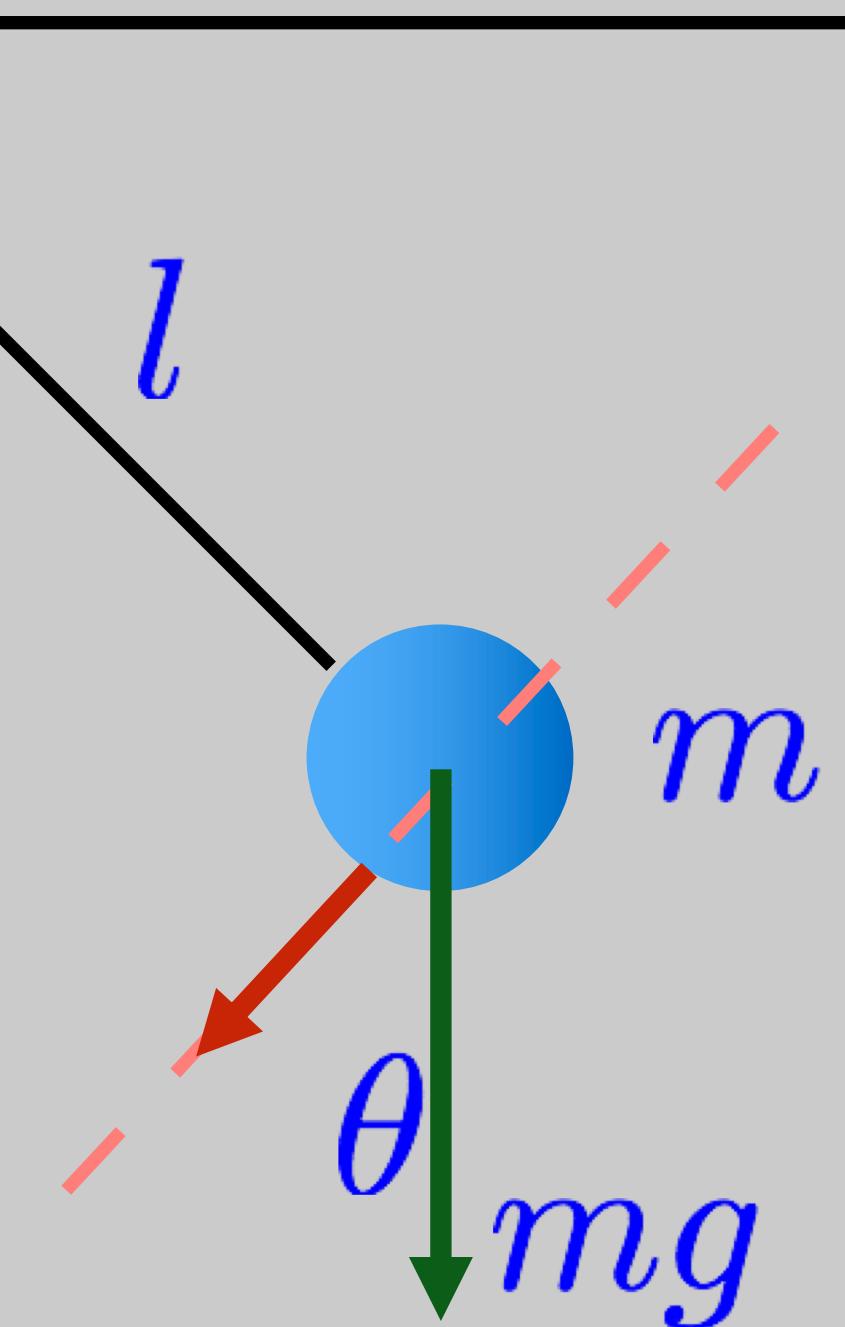
Back to the Pendulum

$$A_{\text{down}} = \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & -\frac{k}{m} \end{bmatrix}$$

$$A_{\text{up}} = \begin{bmatrix} 0 & 1 \\ \frac{g}{l} & -\frac{k}{m} \end{bmatrix}$$

$$|\lambda I - A_{\text{down}}| = \begin{bmatrix} \lambda & -1 \\ \frac{g}{l} & \lambda + \frac{k}{m} \end{bmatrix} = \lambda^2 + \frac{k}{m}\lambda + \frac{g}{l} = 0$$

$$\lambda_{1,2} = -\frac{k}{2m} \pm \frac{1}{2} \sqrt{\frac{k^2}{m^2} - 4\frac{g}{l}}$$



Back to the Pendulum

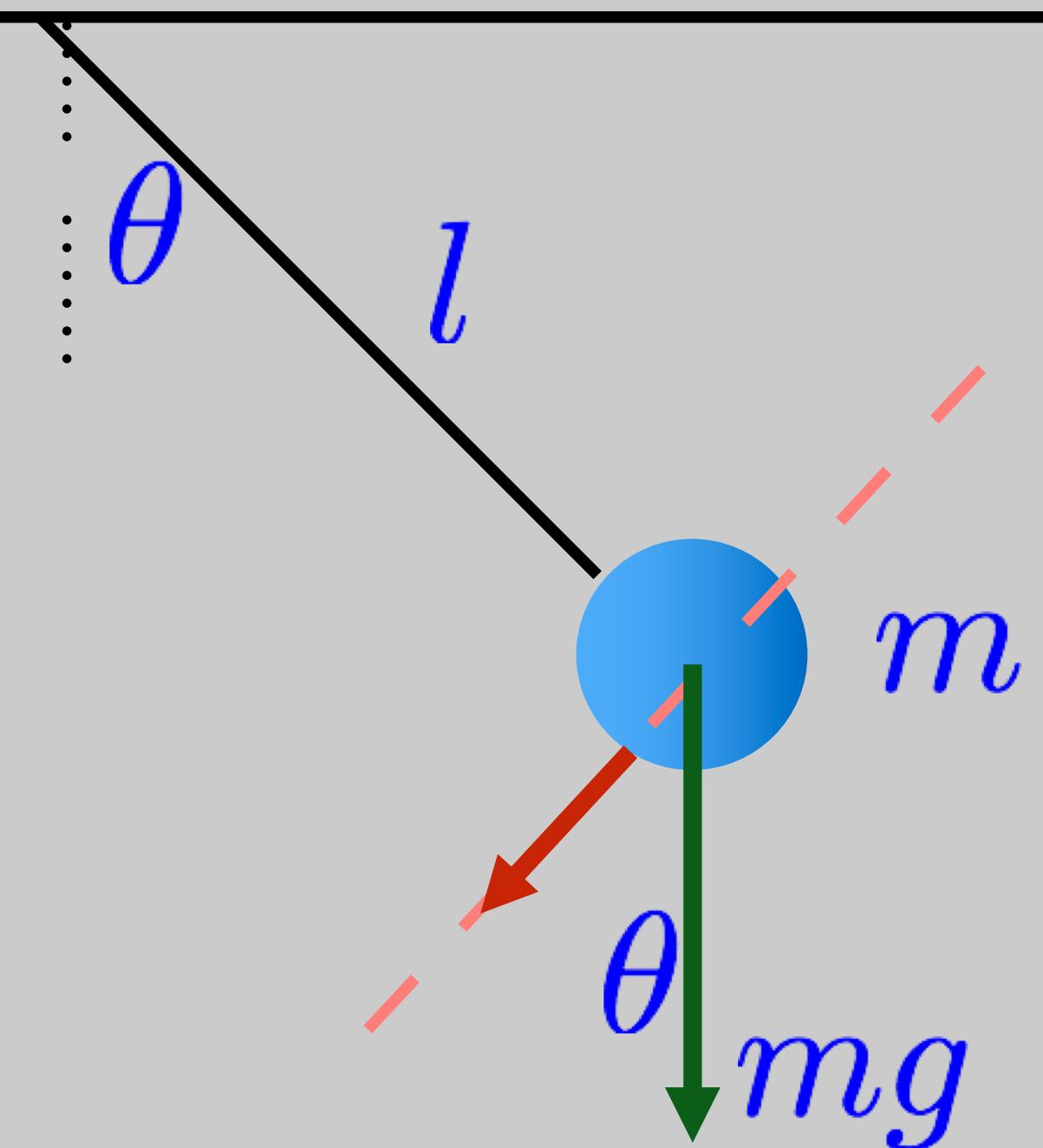
$$\lambda_{1,2} = -\frac{k}{2m} \pm \frac{1}{2} \sqrt{\frac{k^2}{m^2} - 4\frac{g}{l}}$$

If $\frac{k^2}{m^2} \geq 4\frac{g}{l}$, i.e, sqrt is real, then $\frac{k}{2m} \geq \frac{1}{2} \sqrt{\frac{k^2}{m^2} - 4\frac{g}{l}}$

So, $\lambda_{1,2}$ always negative -- stable!

If $\frac{k^2}{m^2} < 4\frac{g}{l}$, i.e, sqrt is imaginary, then $\text{Re}\{\lambda_{1,2}\} = -\frac{k}{2m}$

So, $\text{Re}\{\lambda_{1,2}\}$ always negative -- stable!

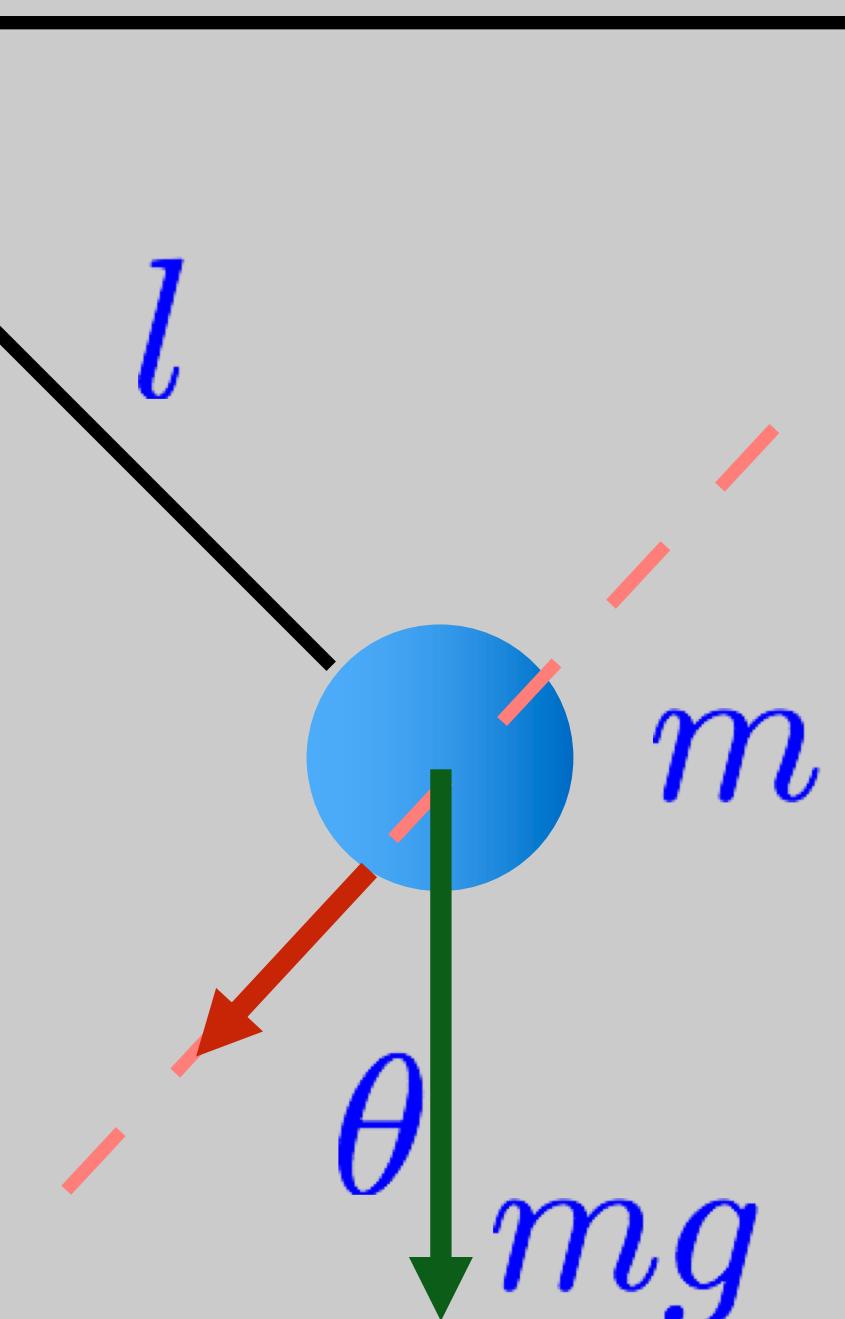


Back to the Pendulum

$$A_{\text{down}} = \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & -\frac{k}{m} \end{bmatrix}$$

$$A_{\text{up}} = \begin{bmatrix} 0 & 1 \\ \frac{g}{l} & -\frac{k}{m} \end{bmatrix}$$

$$|\lambda I - A_{\text{up}}| = \begin{bmatrix} \lambda & -1 \\ -\frac{g}{l} & \lambda + \frac{k}{m} \end{bmatrix} = \lambda^2 + \frac{k}{m}\lambda - \frac{g}{l} = 0$$



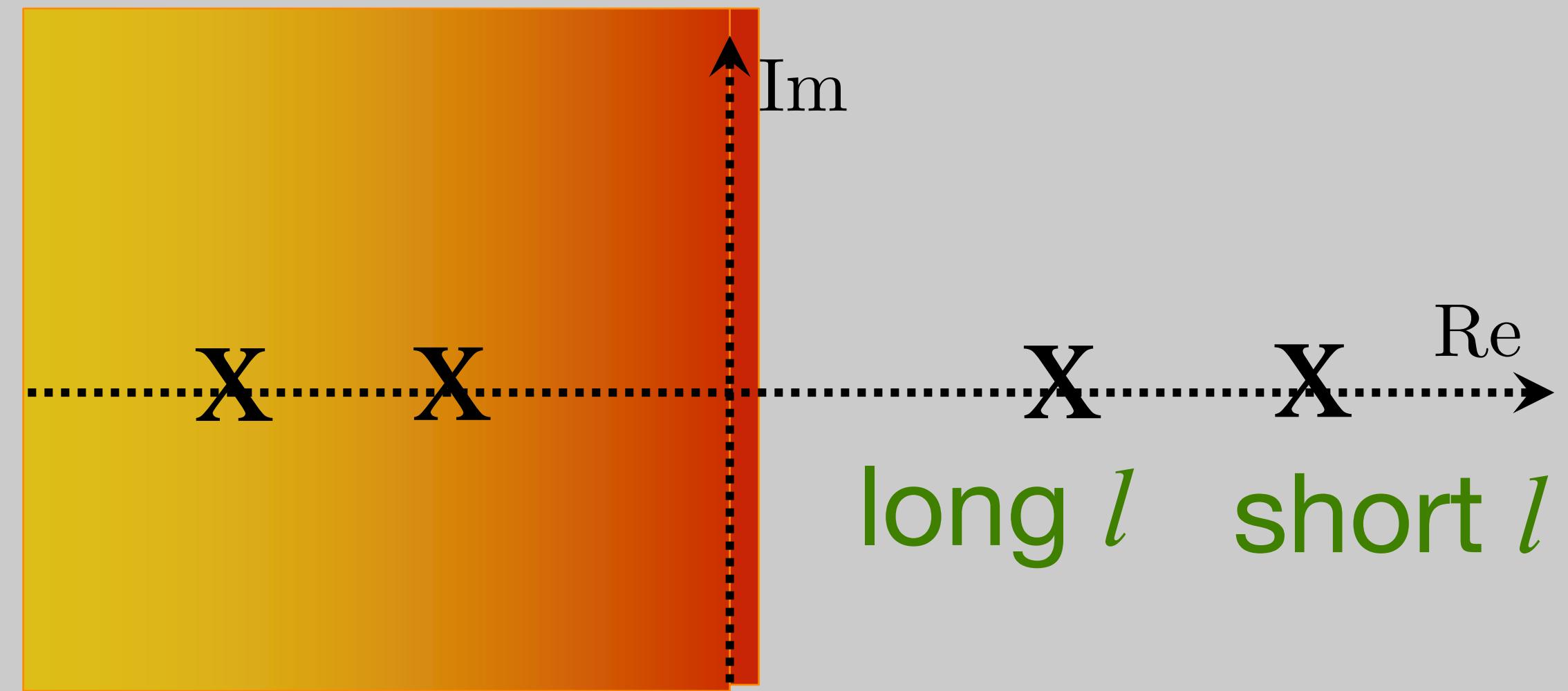
$$\lambda_{1,2} = -\frac{k}{2m} \pm \frac{1}{2} \sqrt{\frac{k^2}{m^2} + 4 \frac{g}{l}}$$

$$\lambda_1 > 0$$

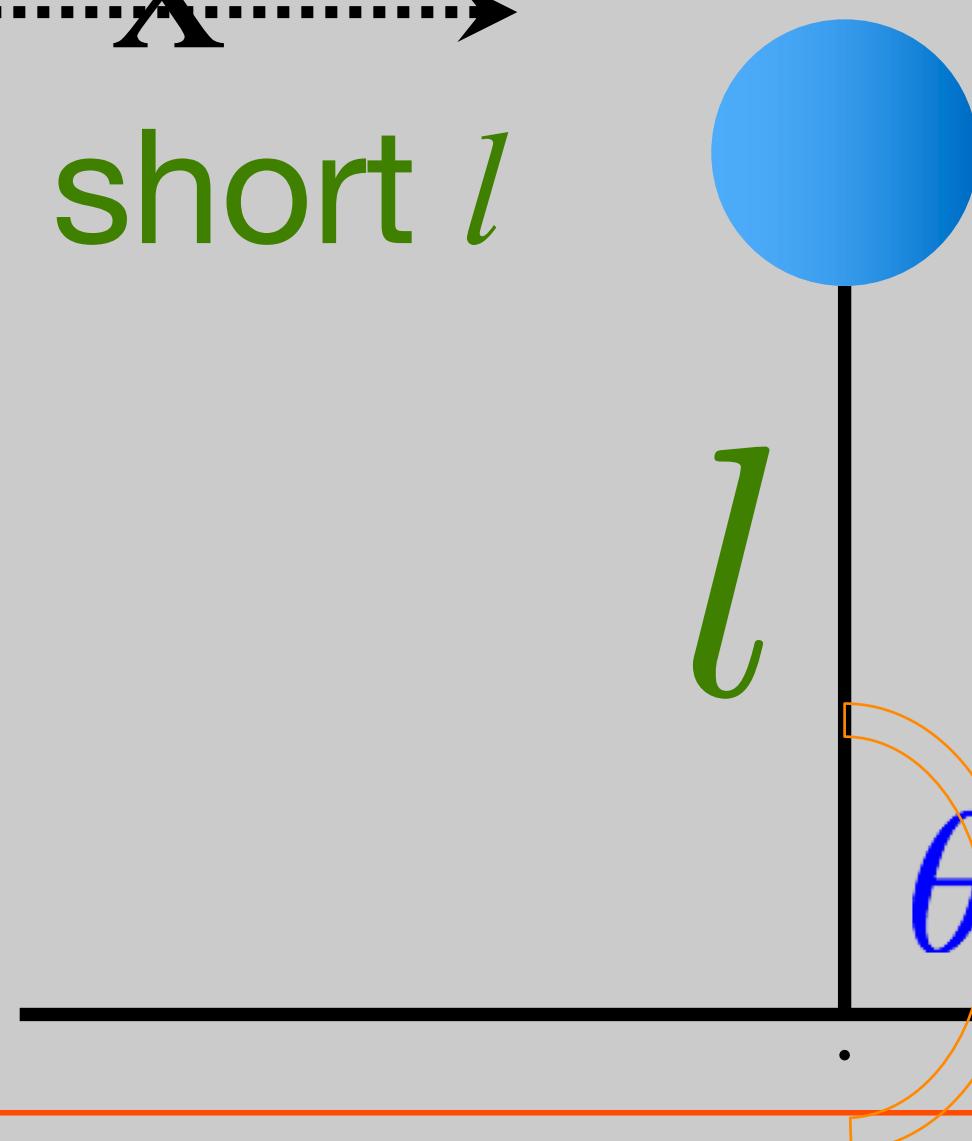
$$\lambda_2 < 0$$

Back to the Pendulum

$$\lambda_{1,2} = -\frac{k}{2m} \pm \frac{1}{2} \sqrt{\frac{k^2}{m^2} + 4 \frac{g}{l}}$$



$$\begin{aligned}\lambda_1 &> 0 \\ \lambda_2 &< 0\end{aligned}$$





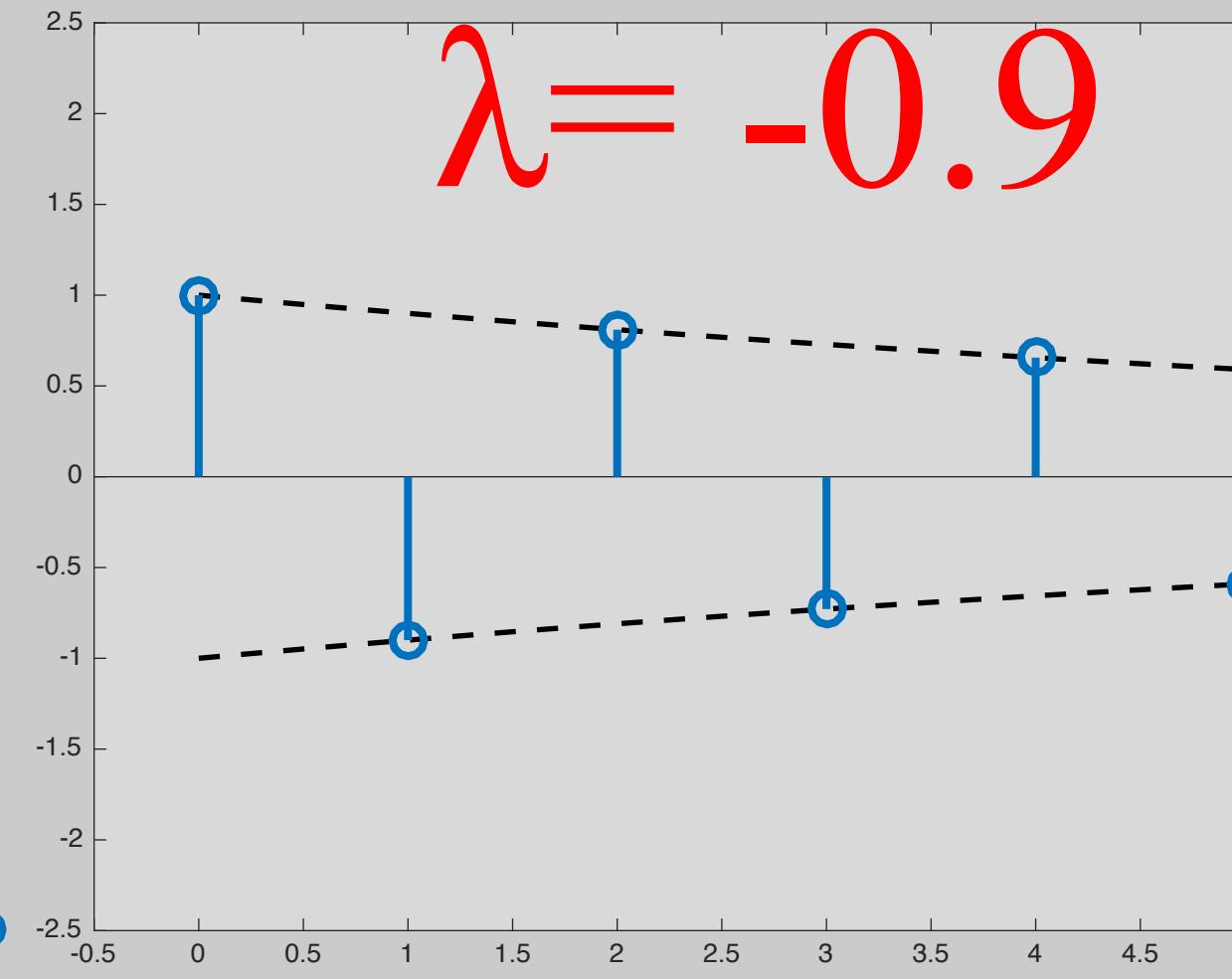
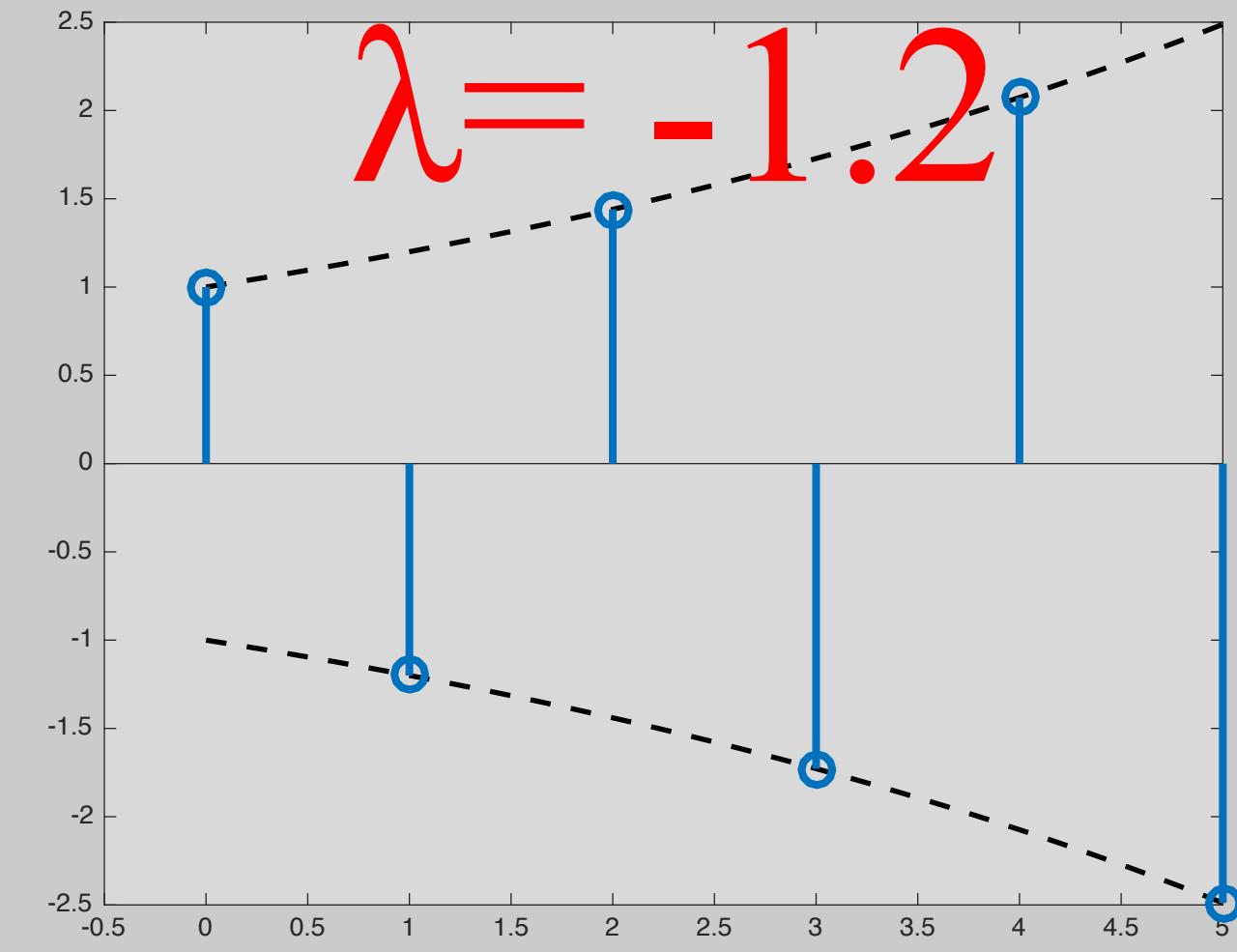
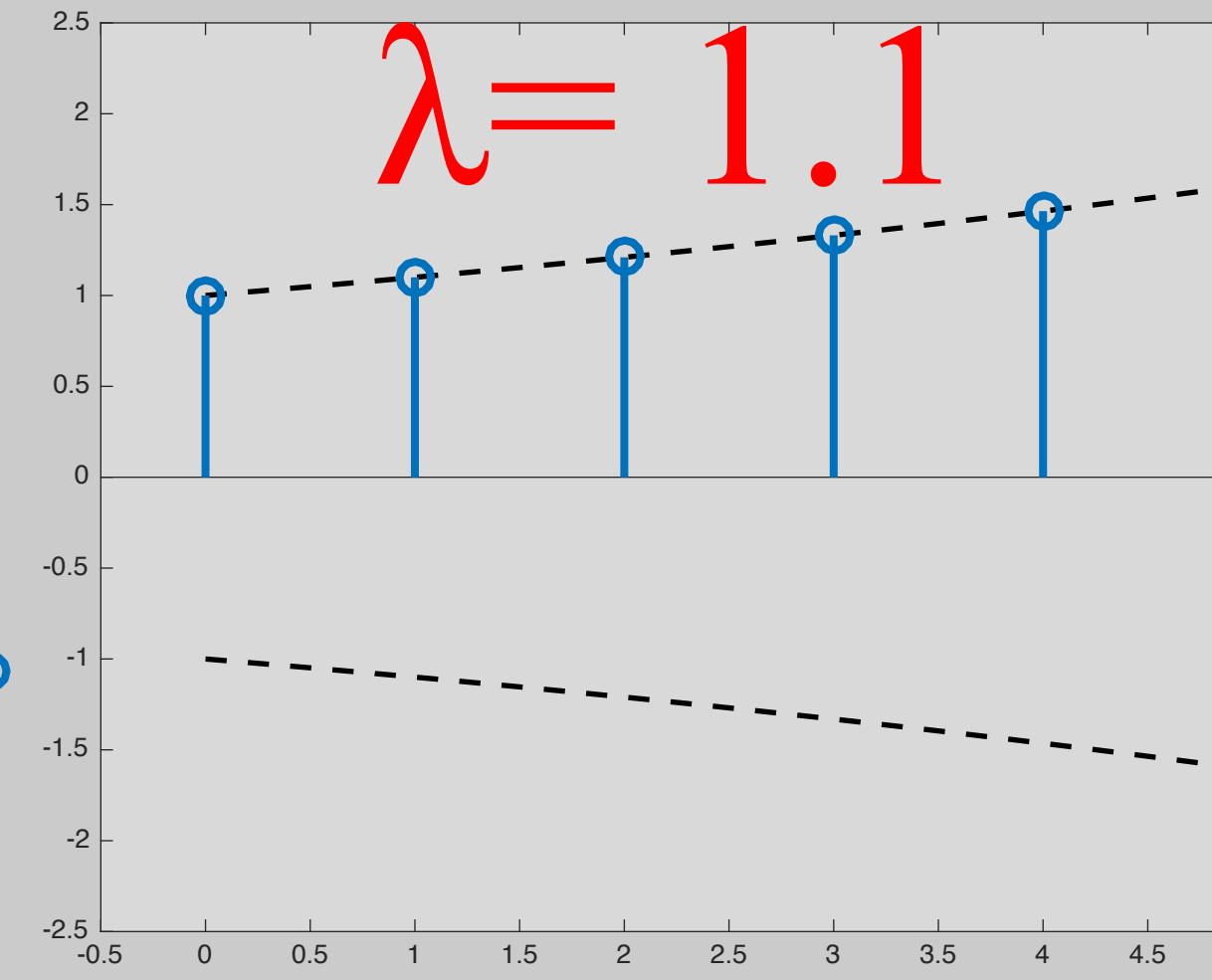
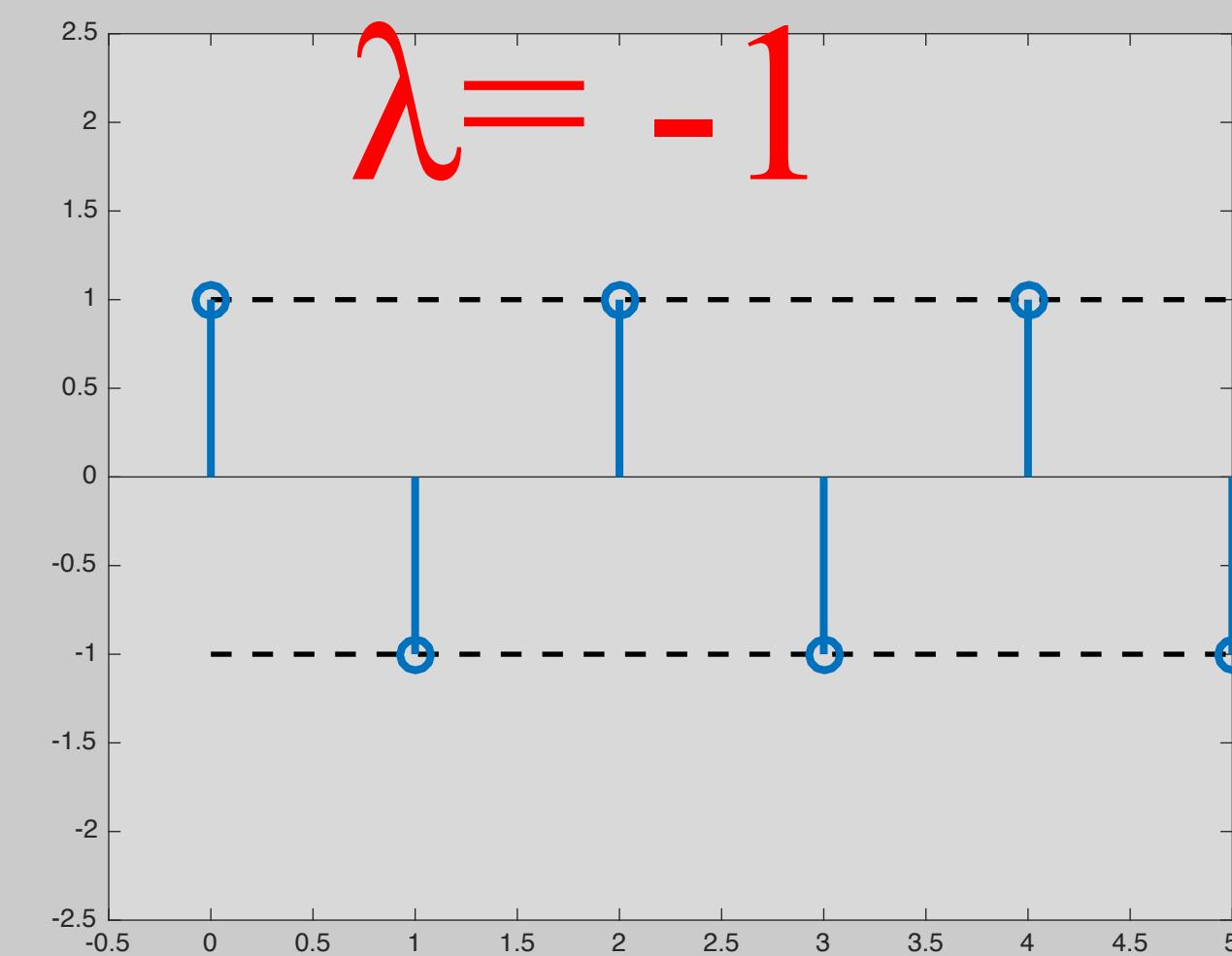
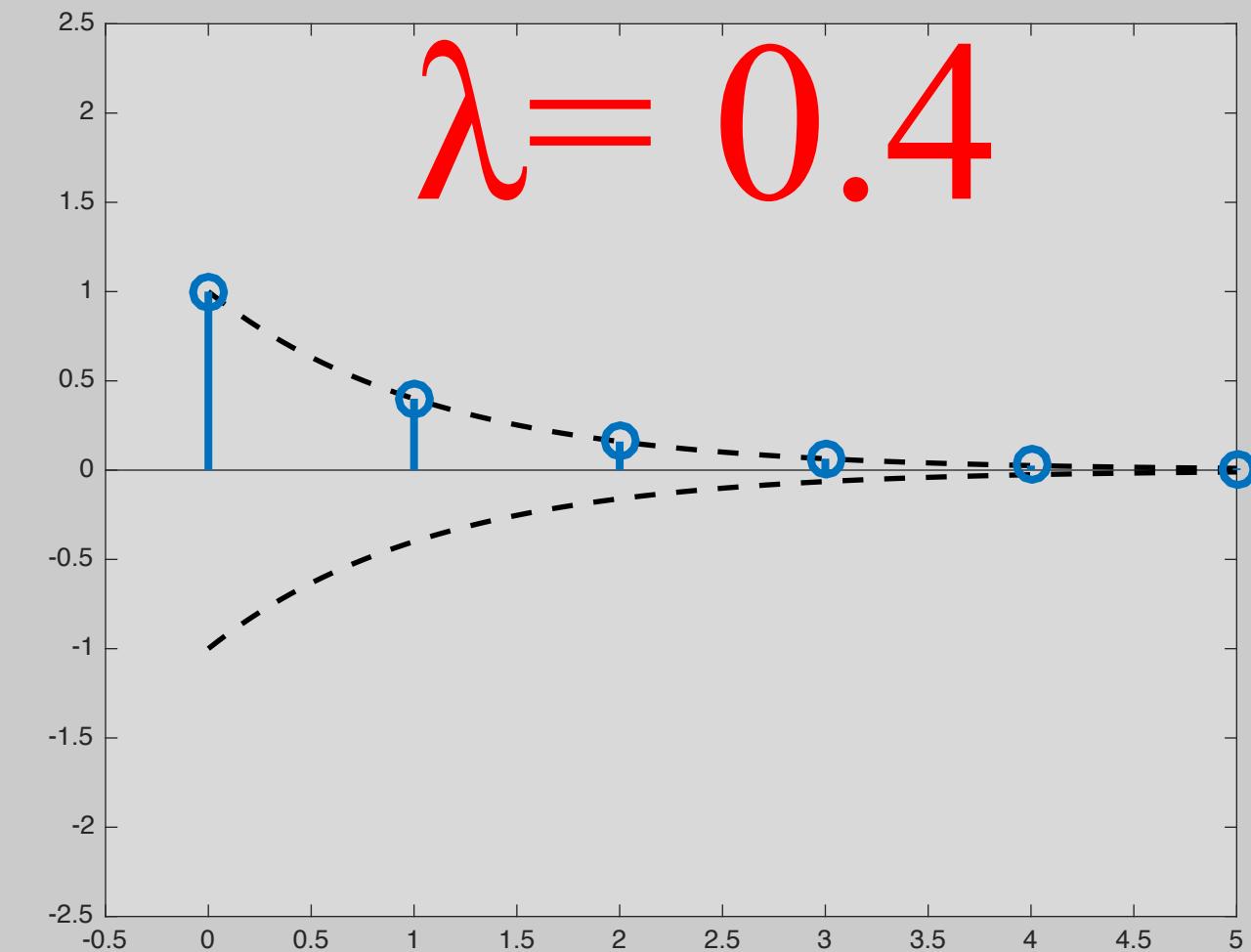
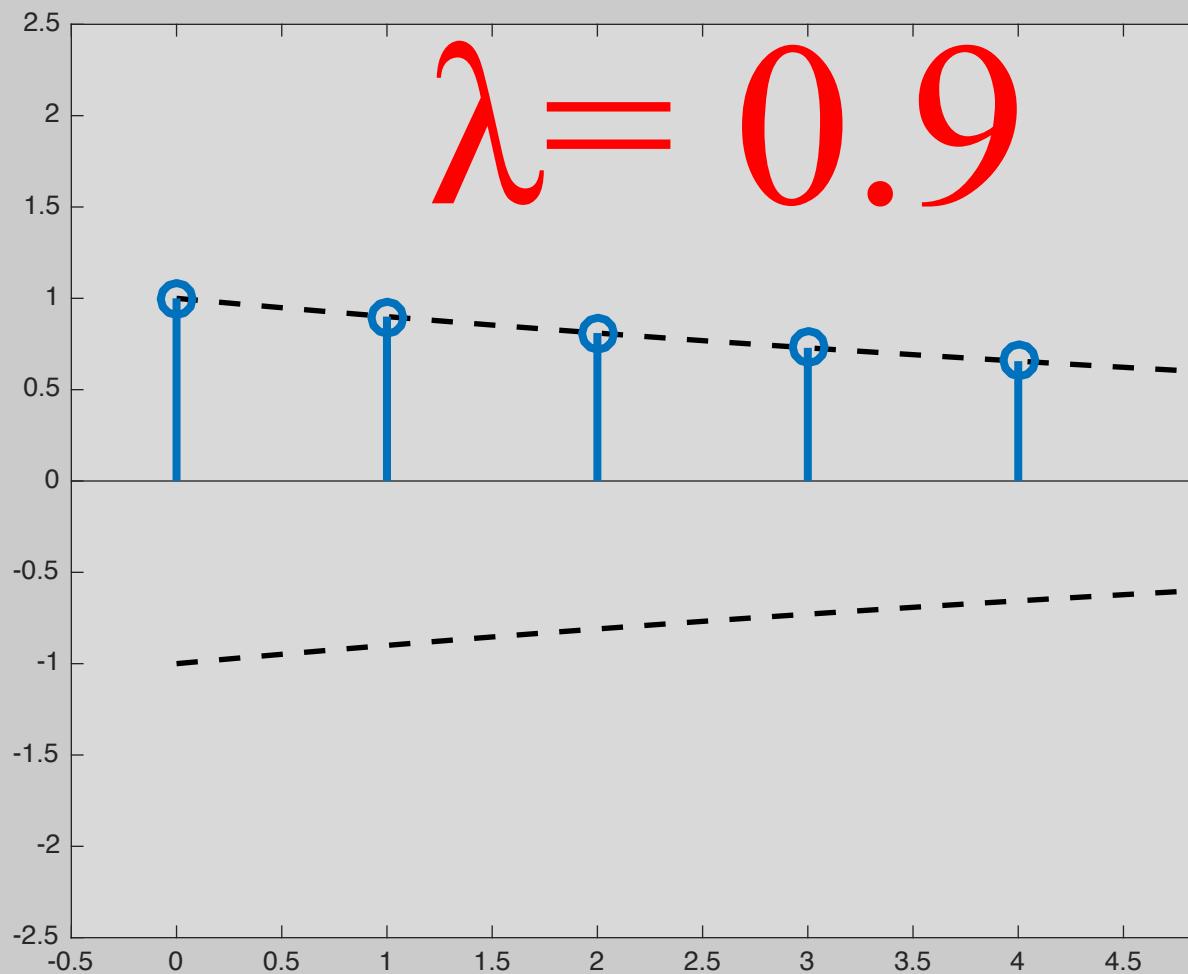
Predicting System Behavior

Discrete Time

$$z(t+1) = \lambda_i z(t)$$

Soln : $\lambda_i^t z(0)$

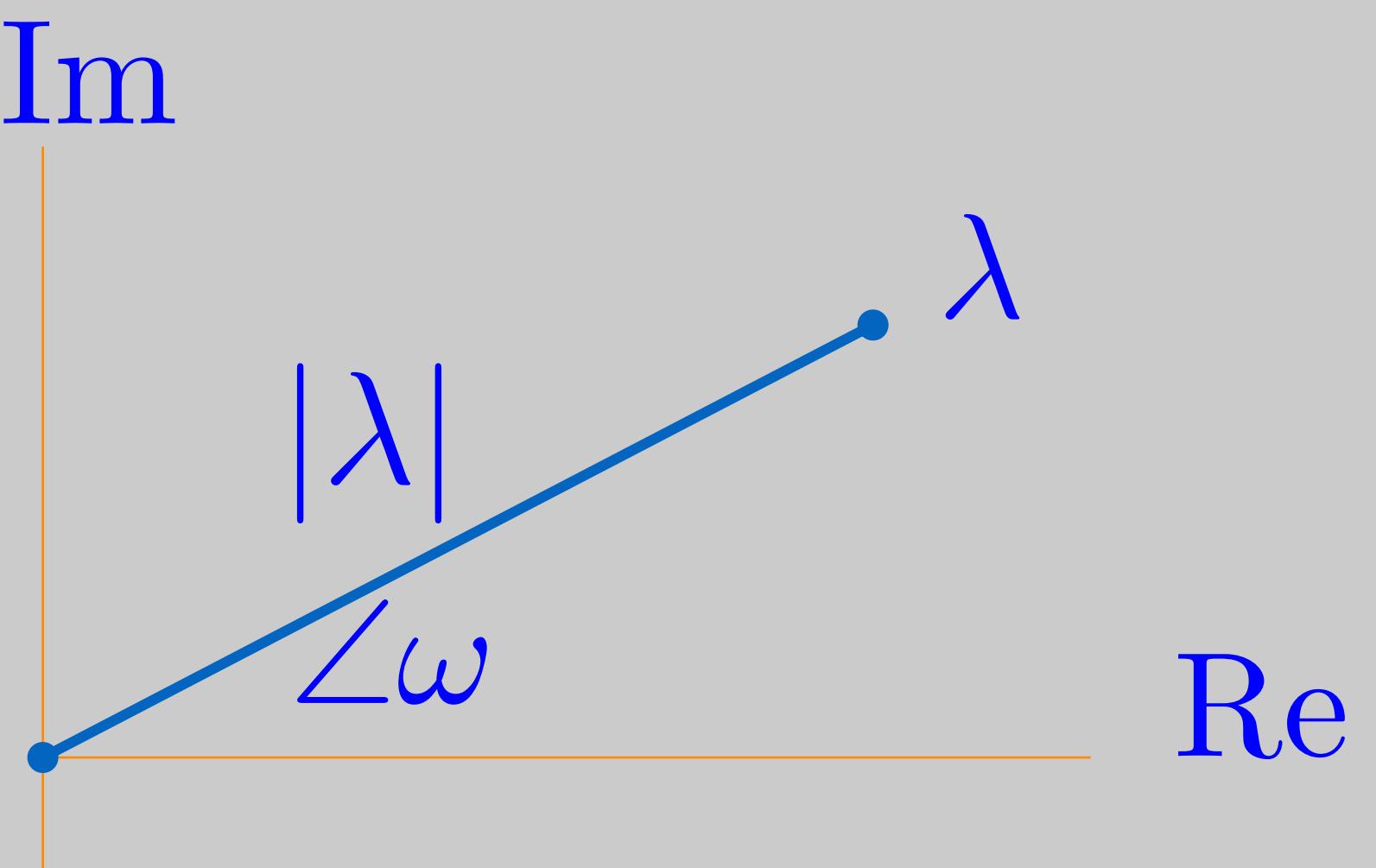
λ^t



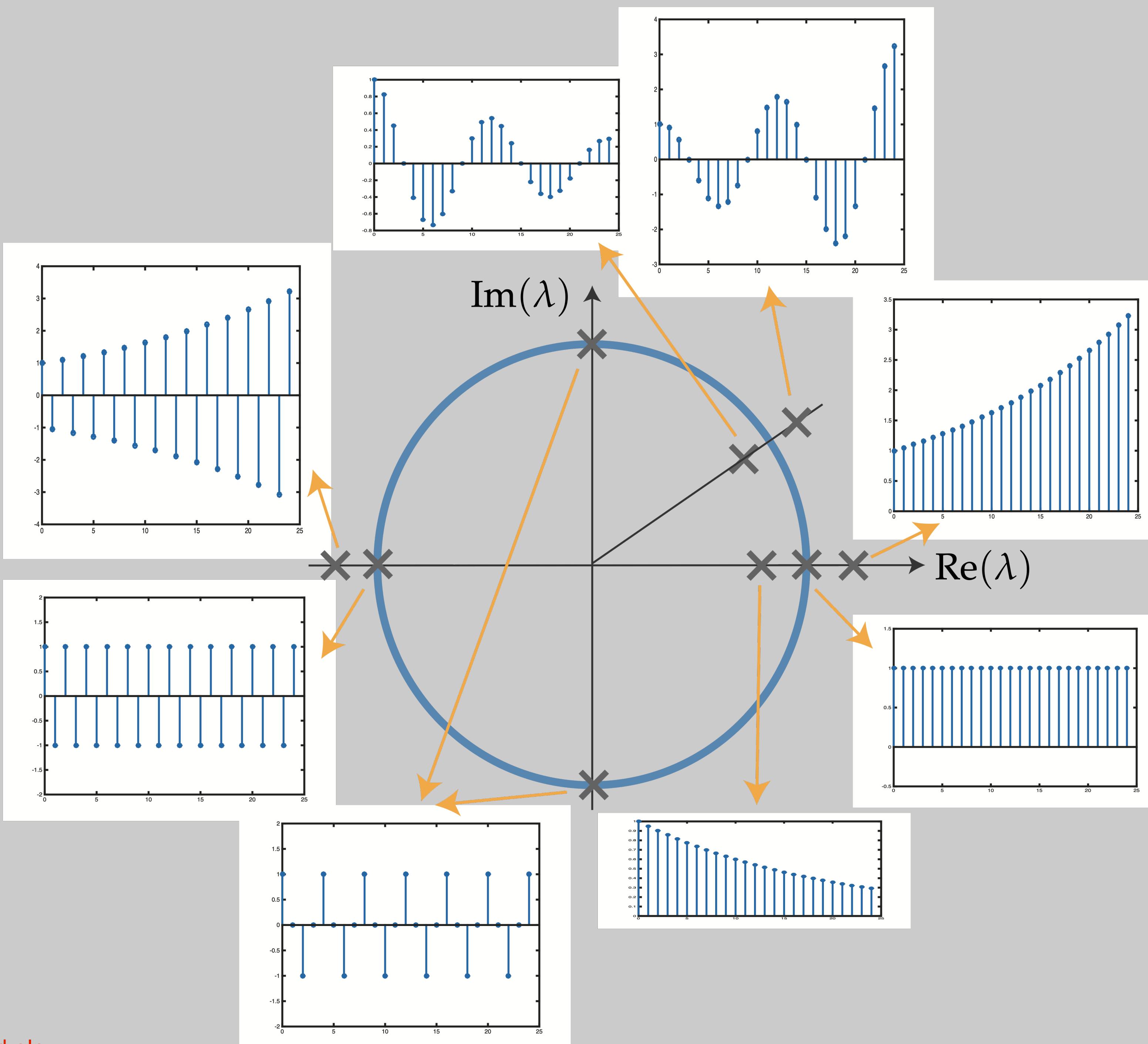
- If λ is complex

$$\lambda^t = (|\lambda|e^{j\omega})^t$$

$$= |\lambda|^t e^{j\omega t}$$



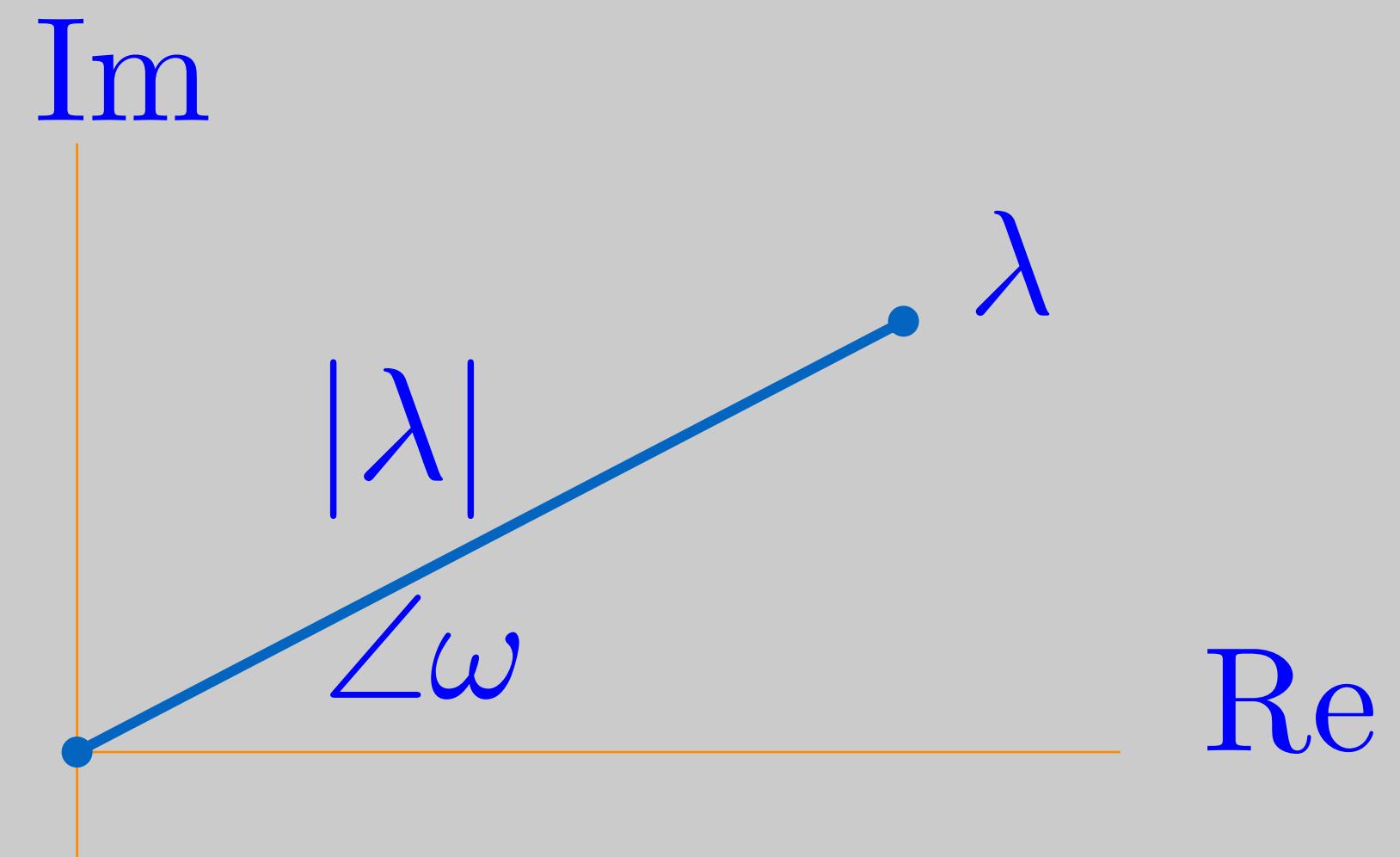
$$\lambda = |\lambda|e^{j\omega} \text{ (Euler)}$$



- If λ is complex

$$\lambda^t = (|\lambda|e^{j\omega})^t$$

$$= |\lambda|^t e^{j\omega t}$$



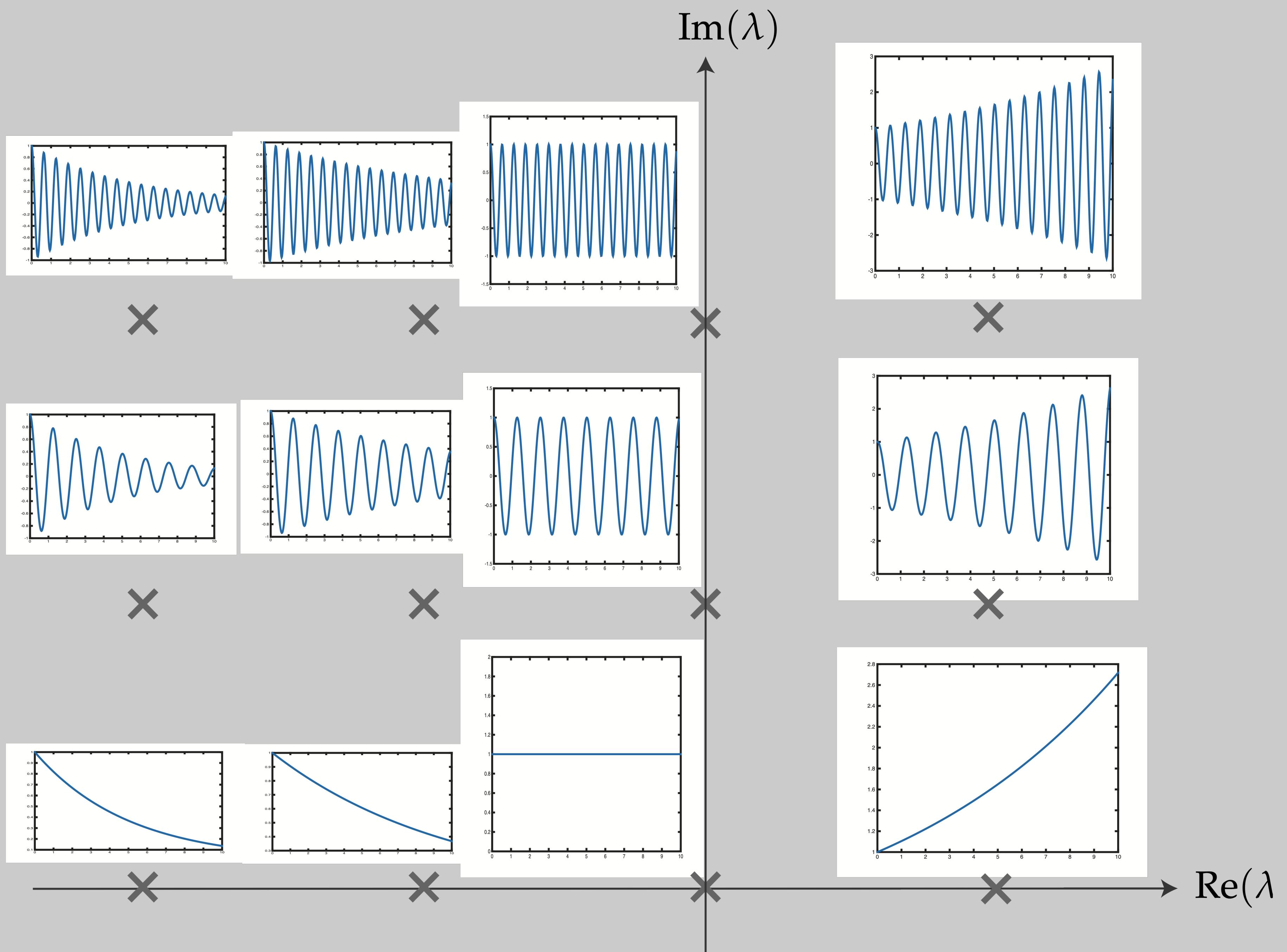
$$\lambda = |\lambda|e^{j\omega} \text{ (Euler)}$$

- Continuous time:

$$\frac{d}{dt} Z_i(t) = \lambda_i Z_i(t) \Rightarrow e^{\lambda_i t} Z_i(0)$$

Q) What does $e^{\lambda t}$ look like for different choices of λ ?

A) $\lambda = v + j\omega \Rightarrow e^{\lambda t} = e^{vt} e^{j\omega t}$



Big Picture

- State space modeling is powerful
- Linear state-space are awesome
 - We can say a lot about them!
 - We can approximate non-linear as linear at eq. points
- What can we say about linear systems?
 - We can tell if they are stable – we have a test!
 - We can predict system behaviour for initial conditions!
- What about controls?
 - Can test if the system can be controlled to reach all states (controllability)
 - We can control a system to move to a certain state (open loop)
 - We can control a system to stay around a state (feedback)

Controllability

Discrete-time: $\vec{x}(t+1) = A\vec{x}(t) + Bu(t)$

From last time:

$$\vec{x}(t) = A^t \vec{x}(0) + \sum_{k=0}^{t-1} A^{t-1-k} Bu(k)$$

$$= A^t \vec{x}(0) + A^{t-1} Bu(0) + A^{t-2} Bu(1) + \cdots + Bu(t-1)$$

$$\vec{x}(t) - A^t \vec{x}(0) = \begin{bmatrix} & & & ? \\ & & & \\ & & & \\ & & & \end{bmatrix} \begin{bmatrix} u(0) \\ u(1) \\ \vdots \\ u(t-1) \end{bmatrix}$$

Controllability

$$\vec{x}(t) - A^t \vec{x}(0) = A^{t-1} B u(0) + A^{t-2} B u(1) + \cdots + B u(t-1)$$

$$\vec{x}(t) - A^t \vec{x}(0) = \left[\begin{array}{c} \\ \\ \\ \end{array} \right] \left[\begin{array}{c} u(0) \\ u(1) \\ \vdots \\ u(t-1) \end{array} \right]$$

Controllability

$$\vec{x}(t) - A^t \vec{x}(0) = A^{t-1} B u(0) + A^{t-2} B u(1) + \cdots + B u(t-1)$$

$$\vec{x}(t) - A^t \vec{x}(0) = \begin{bmatrix} A^{t-1} B & A^{t-2} B & \cdots & AB & B \end{bmatrix} \begin{bmatrix} u(0) \\ u(1) \\ \vdots \\ u(t-1) \end{bmatrix}$$

R_t

Q) Given any $x(0)$, can we find $u(t)$ s.t. $x(t) = x_{\text{target}}$ for some t ?

A) Depends if it is in the span of R_t

Controllability

$$R_t = \begin{bmatrix} A^{t-1}B & A^{t-2}B & \cdots & AB & B \end{bmatrix}$$

Q) For $t < n$? A) No

Q) At $t=n$, If columns are independent? A) Absolutely!

Q) If not independent, does increasing t helps? A) No!

Cayley-Hamilton Theorem: If A is $n \times n$, then A^n can be written as a linear combination of $A^{n-1}, \dots, A, 1$

$$A^n = \alpha_{n-1}A^{n-1} + \cdots + \alpha_1A + \alpha_01$$

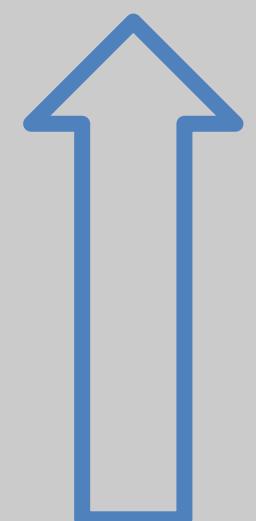
So does: $A^n B = \alpha_{n-1}A^{n-1}B + \cdots + \alpha_1AB + \alpha_0B$

Controllability

$$R_n = \begin{bmatrix} A^{n-1}B & A^{n-2}B & \cdots & AB & B \end{bmatrix}$$

What about R_{n+1} ?

$$R_{n+1} = \begin{bmatrix} A^nB & A^{n-1}B & A^{n-2}B & \cdots & AB & B \end{bmatrix}$$



$$A^nB = \alpha_{n-1}A^{n-1}B + \cdots + \alpha_1AB + \alpha_0B$$

Controllability Test

If R_t doesn't have n independent columns at $t=n$, it never will for $t > n$ either!

Therefore, we need only to examine R_n for controllability:

$$R_n = \begin{bmatrix} A^{n-1}B & A^{n-2}B & \dots & AB & B \end{bmatrix}$$

Conclusion: $\vec{x}(t+1) = A\vec{x}(t) + Bu(t)$

is controllable if and only if

$$\text{rank}\{R_n\} = n$$

Example 1:

$$\vec{x}(t+1) = \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix} \vec{x}(t) + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u(t)$$

A

B

$$R_2 = [AB \quad B] = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$$

$$R_3 = [A^2B \quad AB \quad B] = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

Rank {R2} = 1, < n=2 \Rightarrow Not controllable!

State equations: $x_1(t+1) = x_1(t) + x_2(t) + u(t)$

$x_2(t+1) = 2x_2(t)$ (not stable)

Can not control x_2 , not with u and not with x_1

Example 2:

$$p(t+T) = p(t) + Tv(t) + \frac{1}{2}T^2u(t)$$

$$v(t+T) = v(t) + Tu(t)$$

$$\begin{bmatrix} p(t+T) \\ v(t+T) \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & T \\ 0 & 1 \end{bmatrix}}_A \begin{bmatrix} p(t) \\ v(t) \end{bmatrix} + \underbrace{\begin{bmatrix} \frac{1}{2}T^2 \\ T \end{bmatrix}}_B u(t)$$

$$R_2 = [AB \quad B] = \begin{bmatrix} \frac{3}{2}T^2 & \frac{1}{2}T^2 \\ T & T \end{bmatrix}$$

Rank = 2 \Rightarrow Controllable!

Continuous Time (no derivation here)

- The continuous-time system

$$\frac{d}{dt} \vec{x}(t) = A\vec{x}(t) + Bu(t)$$

is controllable if and only if

$$R_n = \begin{bmatrix} A^{n-1}B & A^{n-2}B & \cdots & AB & B \end{bmatrix}$$

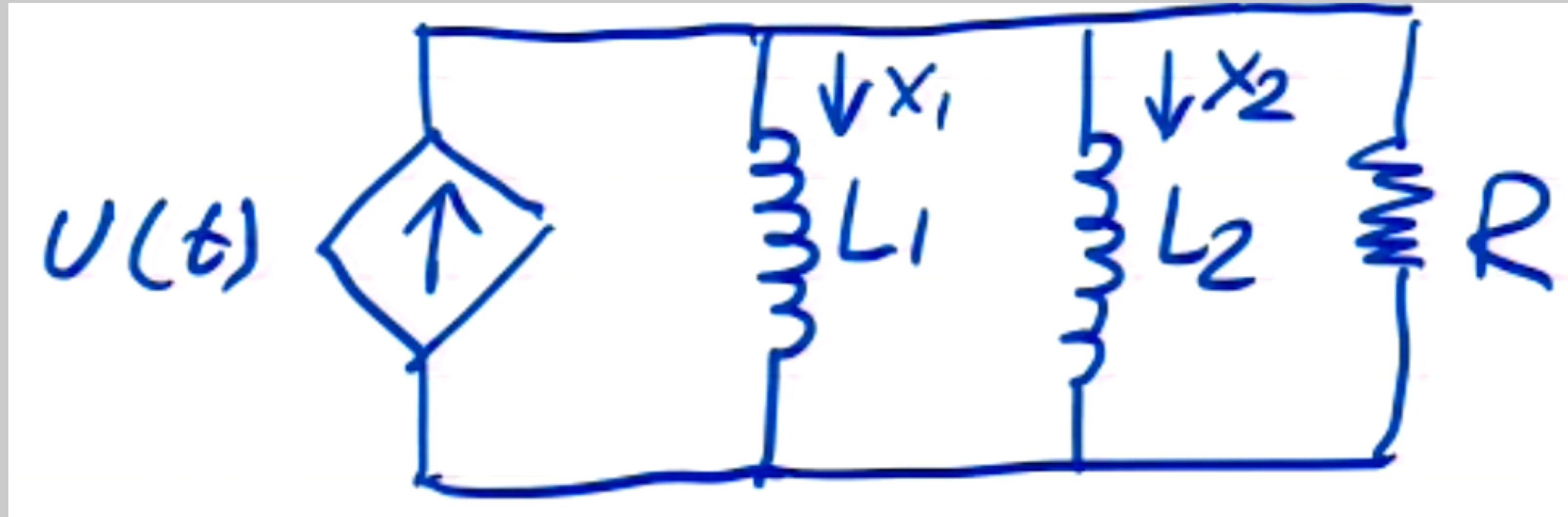
has rank = n

Example 3 + Quiz

- Write the state model:

$$\frac{d}{dt} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} \quad & \quad \\ \quad & \quad \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} \quad \\ \quad \end{bmatrix} u(t)$$

For:

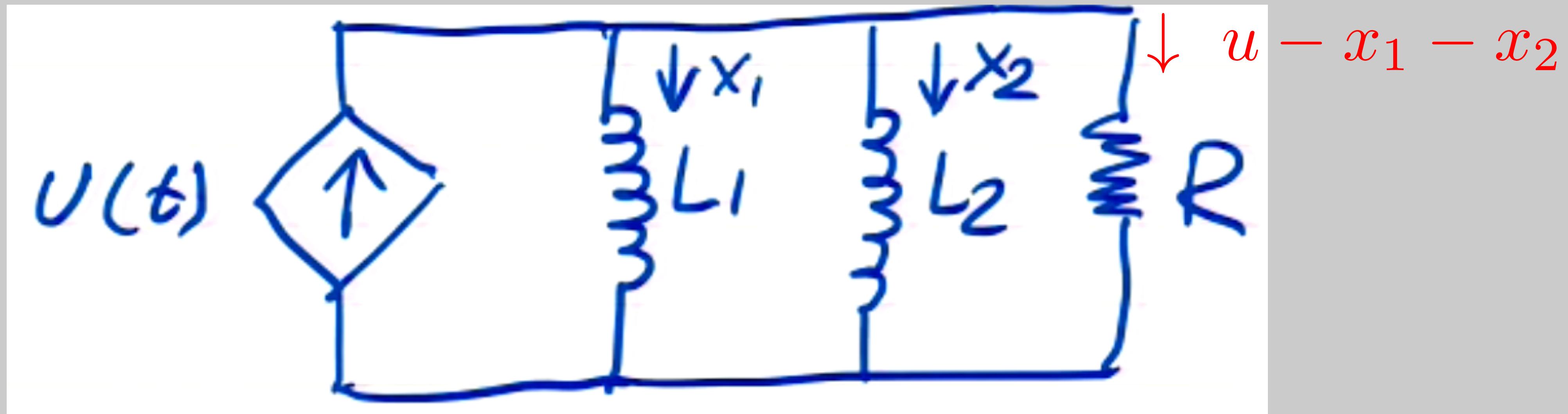


Quiz

- Write the state model:

$$\frac{d}{dt} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = \begin{bmatrix} -\frac{R}{L_1} & -\frac{R}{L_1} \\ -\frac{R}{L_2} & -\frac{R}{L_2} \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} \frac{R}{L_1} \\ \frac{R}{L_2} \end{bmatrix} u(t)$$

For:



$$V_r = R(u - x_1 - x_2) = L_1 \dot{x}_1 = L_2 \dot{x}_2$$

Example 3

- Controllability:

$$B = \begin{bmatrix} \frac{R}{L_1} \\ \frac{R}{L_2} \end{bmatrix}$$

$$AB = \begin{bmatrix} -\frac{R}{L_1} \left(\frac{R}{L_1} + \frac{R}{L_2} \right) \\ -\frac{R}{L_2} \left(\frac{R}{L_1} + \frac{R}{L_2} \right) \end{bmatrix}$$
$$R = [AB \ B]$$

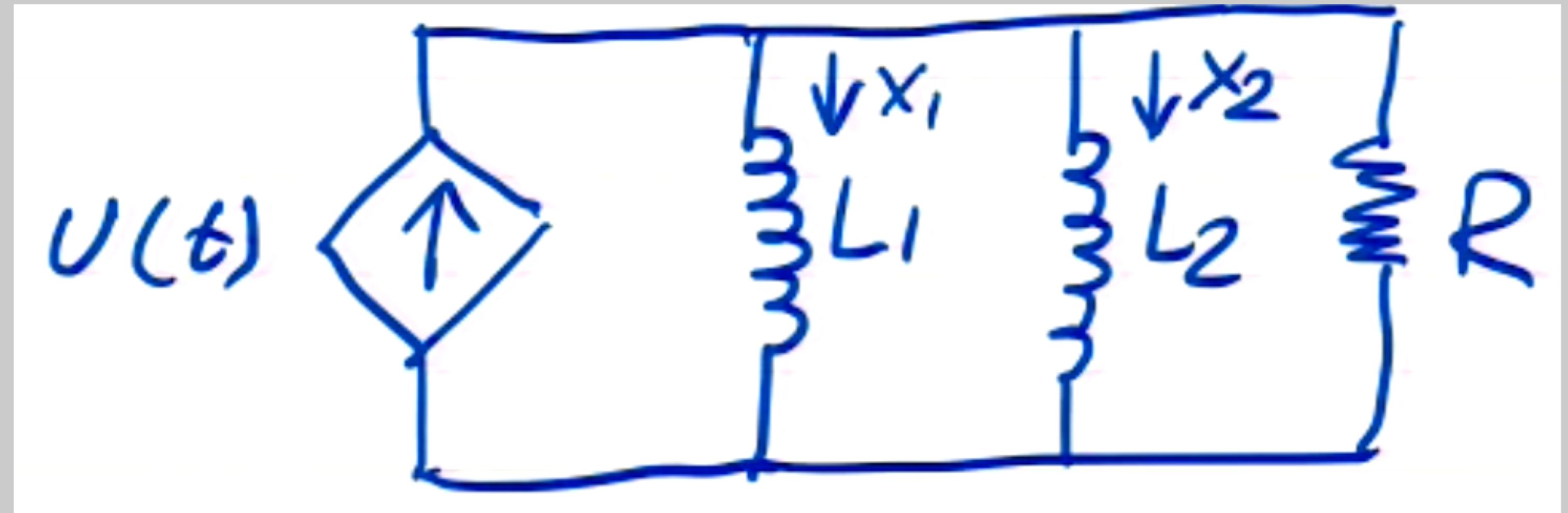
$$AB = \left(\frac{R}{L_1} + \frac{R}{L_2} \right) B$$

Rank = 1 !

Not controllable

Physical explanation

- Why can't I drive the currents x_1 and x_2 freely using $U(t)$?



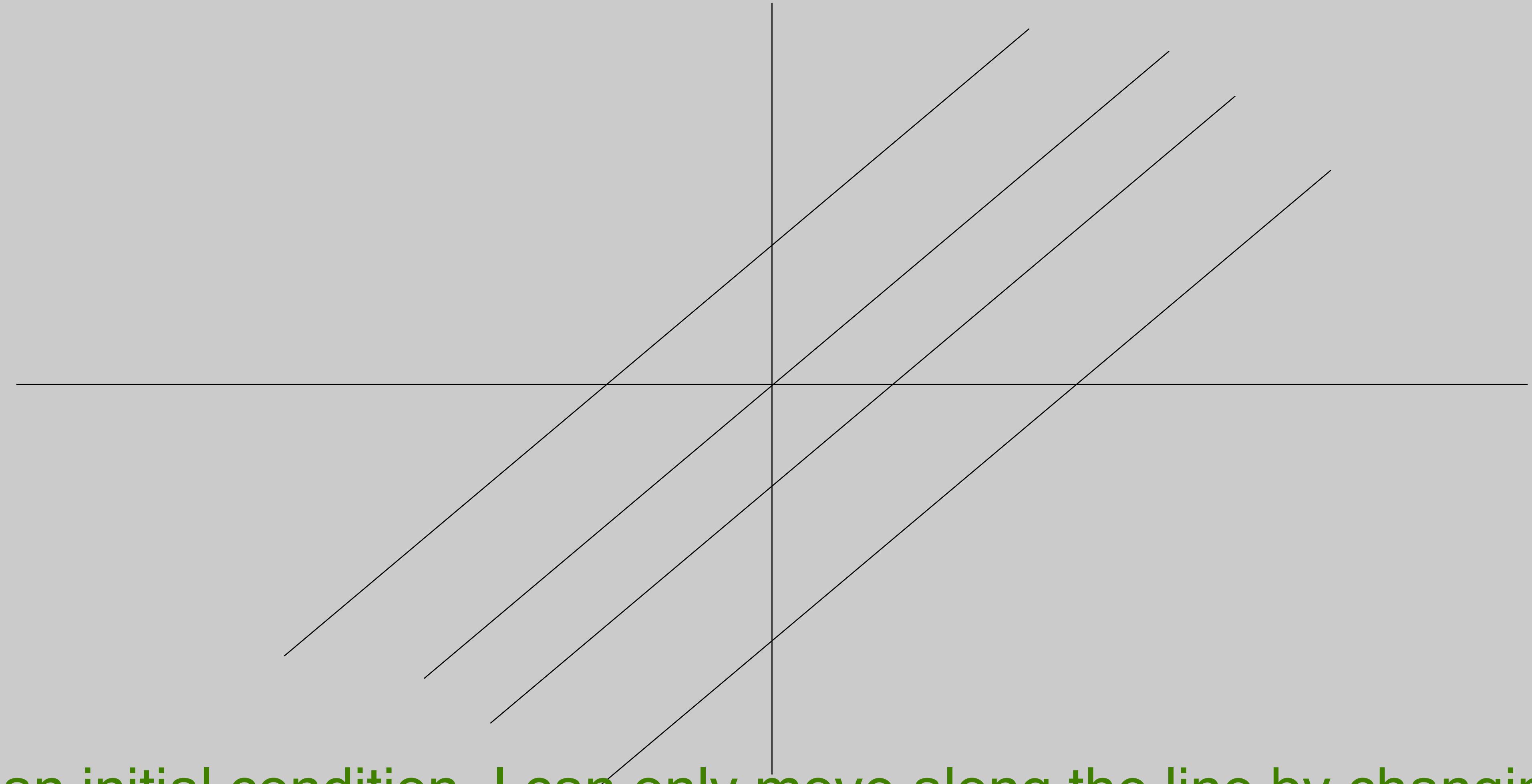
$$L_1 \frac{dx_1}{dt} = L_2 \frac{dx_2}{dt} = R \cdot i_R = V_R \quad \Rightarrow \quad L_1 \frac{dx_1}{dt} - L_2 \frac{dx_2}{dt} = 0$$

$$\frac{d}{dt} (L_1 x_1 - L_2 x_2) = 0 \quad \Rightarrow \quad (L_1 x_1 - L_2 x_2) = \text{Const}$$

$$(L_1 x_1 - L_2 x_2) = \text{Const}$$

x_2

x_1



Given an initial condition, I can only move along the line by changing U

Q) What if $A = 0$? Can the system be controllable?

$$\frac{d}{dt} \vec{x}(t) = Bu(t)$$

$$R = [0 \ 0 \ 0 \ 0 \ \cdots \ B]$$

A) Only if $u(t)$ is a vector with the same number of elements as the number of states

Summary

- Described and derived conditions for controllability of linear state models.
 - Rank of R_n for both discrete and continuous
- Showed examples of controllable and non-controllable systems
- Next time:
 - Open loop and state feedback control
 - Controllers to make systems do what we want!