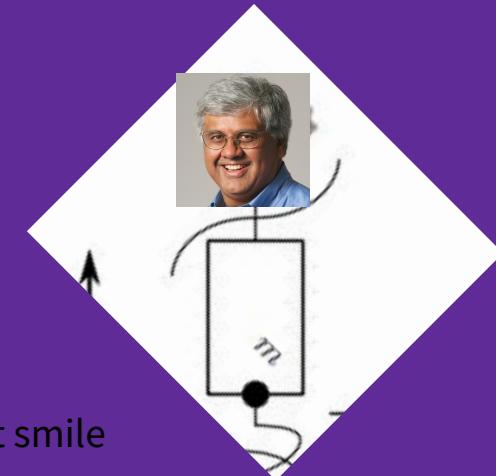


# EECS/MechE/BioE C106A: Midterm 2 Review Session

The return of Prof. Tarun Amarnath!

Look at that brilliant smile



# **Lab**

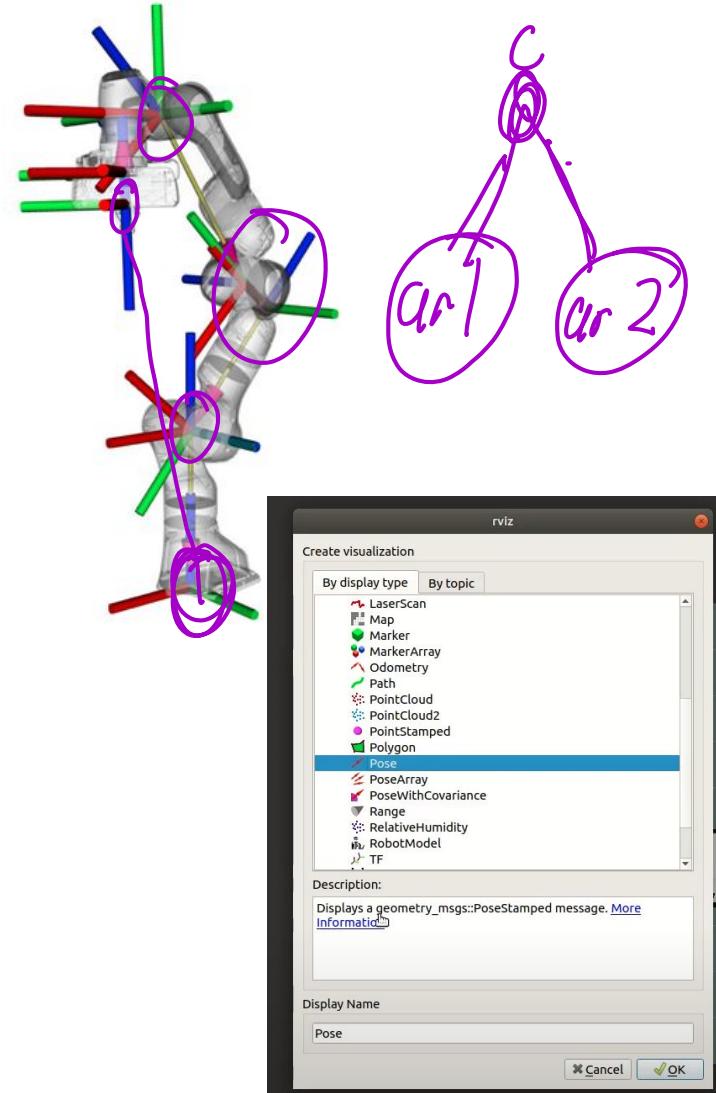
- Make sure you're familiar with the basic setup operations!
- Nodes, topics, publishers, subscribers
- Creating packages, running programs
- Work done in labs (planning, tracking, mapping, etc.)

# TF Tree, transforms, & RVIZ

- Can perform transform between any two coordinate frames in TF tree using tf2
- How to code a transform

```
tfBuffer = tf2_ros.Buffer()
tfListener = tf2_ros.TransformListener(tfBuffer)
while not rospy.is_shutdown():
    try:
        trans = tfBuffer.lookup_transform('odom', 'base_footprint', rospy.Time())
        print(trans)
        break
    except:
        pass
```

- We can display a bunch of objects in RViz
  - Image, TF, Robot Model, Point, Marker



# Labs - Fullstack Robotics

- Perception

- AR Tags - Forms a TF between camera and AR tags
- RGBD Cameras
- RGB vs HSV



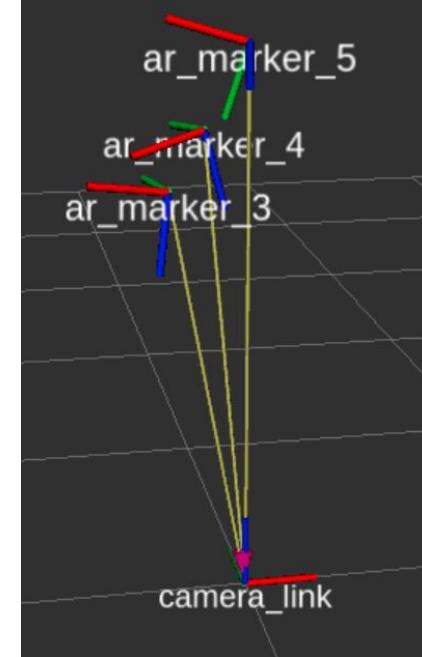
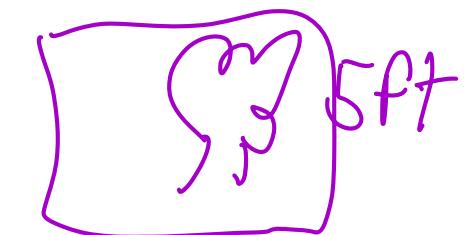
- Path Planning

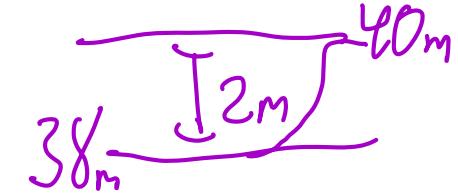
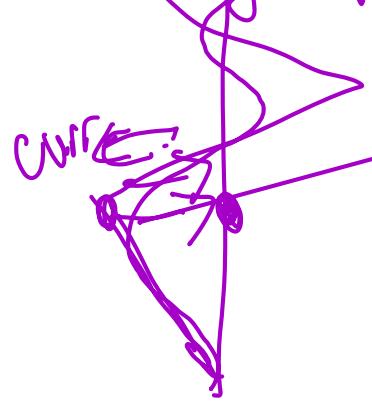
- Its ok be happy (don't worry about path planning for exam)

- Control

- Positional vs Velocity PID control
- Porptional Integral Derivative

goal po





# All the Past Content...

# Rigid Body Transformations

- Length and orientation preserving
- Represent a movement or a change in coordinate frame
- Rotations, translations, or both (screw motion)

# Homogeneous Transformation Matrices

- Compact representation
- Both rotation and translation included
- Can stack and invert

$$g = \begin{bmatrix} R & p \\ 0 & 1 \end{bmatrix} \quad g^{-1} = \begin{bmatrix} R^T & -R^T p \\ 0 & 1 \end{bmatrix}$$

# Exponential Coordinates

- **Goal:** Create rotation and homogeneous transformation matrices as a *function of time*
- Comes from solving a differential equation
- We only need information about **how** the object moves (time is a parameter that's plugged in)

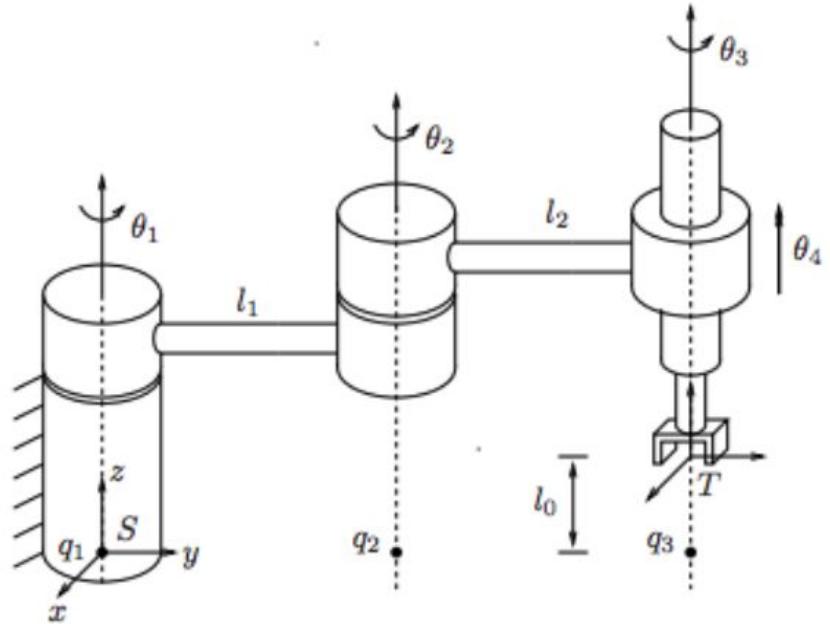
$$g(x) = e^{\hat{\xi}t} g(0)$$
$$(\xi, t)$$

$$R(t) = e^{\hat{\omega}t} R(0)$$
$$(\omega, t)$$

# Forward Kinematics

- **Goal:** Find the location of the tool after a multi-joint robot arm has moved around
- Compose exp. coords

$$g_{st}(\theta_1, \dots, \theta_n) = e^{\hat{\gamma}_1 \theta_1} \cdots e^{\hat{\gamma}_n \theta_n} g_{st}(0)$$



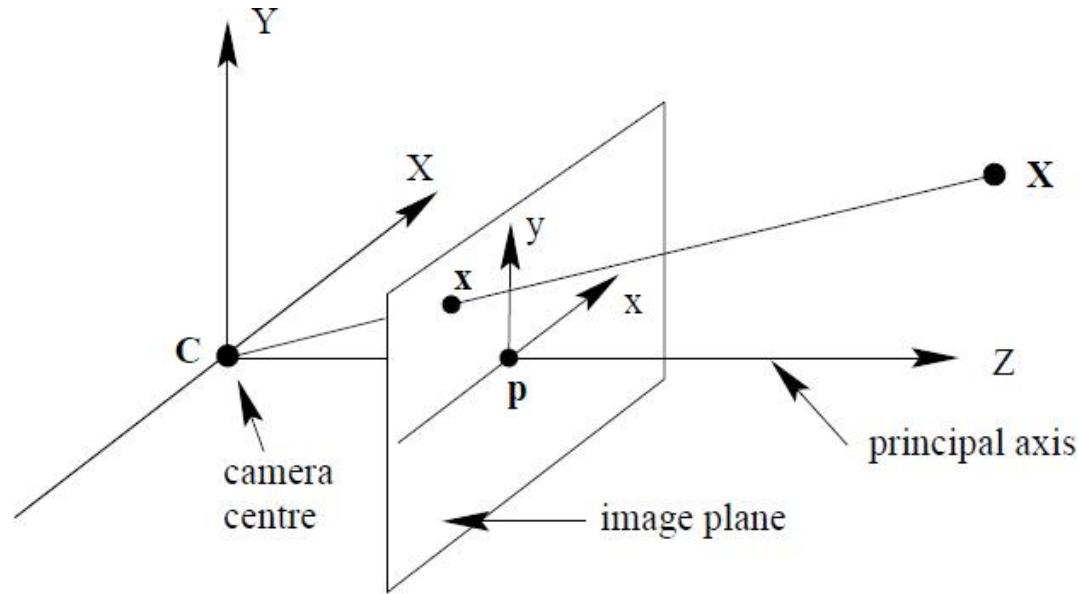
# Inverse Kinematics

*- Given: Some final position  
- Joint twists  
- Find:  $\theta_s$*

- How do we move our robot's joints to reach a desired configuration?
- Use Paden-Kahan subproblems along with tricks (reduce problem down to simpler parts)

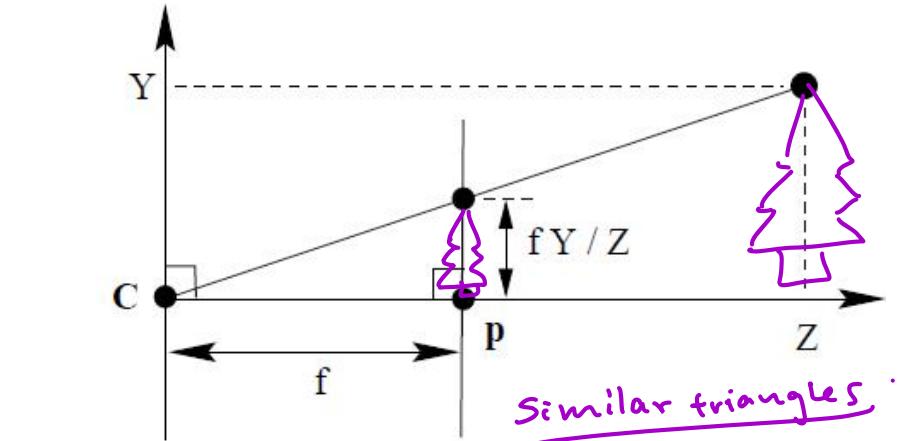
# Computer Vision

# Pinhole Camera Model



$$(X, Y, Z)^T \rightarrow \left(\frac{fX}{Z}, \frac{fY}{Z}\right)^T$$

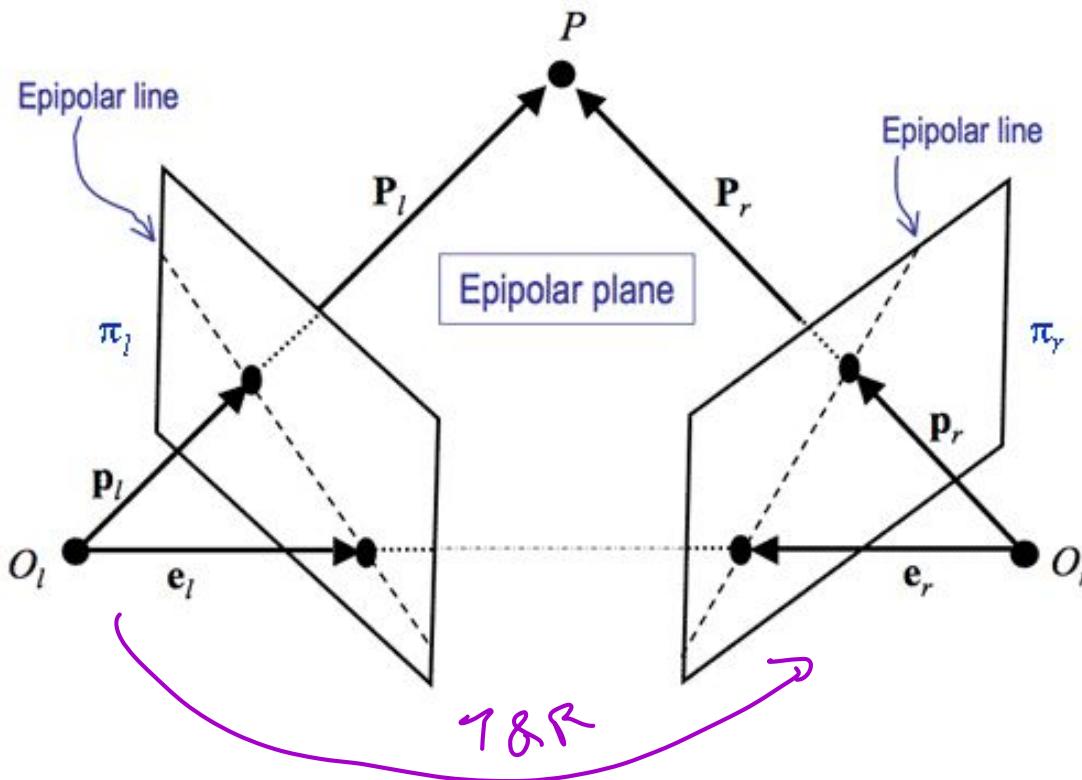
$$z \leftarrow \textcircled{2} x = K X$$



$$K = \begin{bmatrix} f_x & 0 & o_x \\ 0 & f_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$

$\frac{y}{f} \rightarrow \frac{y}{f} = \frac{Y}{Z}$   
 $\uparrow$   
 Img  $\frac{y}{f}$   
 $\frac{y}{f} = \frac{Y}{Z}$

# Two-View Geometry



$$\begin{aligned} (\mathbf{x}_2)^T E \mathbf{x}_1 = 0 \\ \text{point in img 2} \quad \uparrow \quad \text{pt in img 1} \\ = \hat{\mathbf{T}} \mathbf{R} \end{aligned}$$

# Convolutions

- Slide a kernel over some image
- Understand some information about the picture

Input image



Convolution  
Kernel

$$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

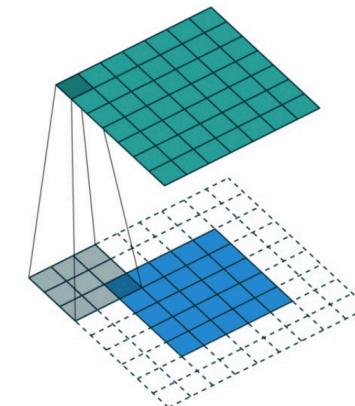
Feature map



Original image



Gaussian Blur filter applied



# Velocities

# What do we mean by them?

- Velocity in general is the rate of change with respect to some reference frame
- With robots, use a **stationary frame**
- Calculate the velocity of some point attached to the end effector wrt to the base

# Some Important Considerations

- Spatial & body velocities - **just a coordinate shift**, tells us which coordinate system to use
- Spatial and body velocities are **twists**
- Generic expressions for any point  $\rightarrow$  how robot moves
- Can apply them to a specific point to determine that point's velocity



$$\mathbf{r}_q = \mathbf{v}_{AB}^S \mathbf{P}_a$$

# Spatial Velocity

- Express our point in the **spatial frame**

→ A frame  
is spatial

$$\begin{aligned} \frac{d}{dt} \left( q_a(t) \right) &= \dot{g}_{ab} \cdot \dot{q}_b \\ \text{Switch frames} \quad \left( \dot{q}_a(t) \right) &= \dot{g}_{ab} \cdot \dot{q}_b \\ v_{q_a}(t) &= \underbrace{\dot{g}_{ab} g_{ab}^{-1}}_{\hat{v}^s} \dot{q}_a \end{aligned}$$

$$\hat{V}_{ab}^s := \dot{g}_{ab} g_{ab}^{-1} = \begin{bmatrix} \dot{R}_{ab} R_{ab}^T & -\dot{R}_{ab} R_{ab}^T p_{ab} + \dot{p}_{ab} \\ 0 & 0 \end{bmatrix} \quad V_{ab}^s = \begin{bmatrix} v_{ab}^s \\ \omega_{ab}^s \end{bmatrix} = \begin{bmatrix} -\dot{R}_{ab} R_{ab}^T p_{ab} + \dot{p}_{ab} \\ (\dot{R}_{ab} R_{ab}^T)^\vee \end{bmatrix}$$

# Body Velocity

- Point is expressed in terms of the **body frame**

$$v_{q_b}(t) = \underbrace{g_{ab}^{-1}(t)}_{\hat{v}_{ab}} \dot{g}_{ab}(t) q_b$$

$$\widehat{V}_{ab}^b := g_{ab}^{-1}(t) \dot{g}_{ab} = \begin{bmatrix} R_{ab}^T \dot{R}_{ab} & R_{ab}^T \dot{p}_{ab} \\ 0 & 0 \end{bmatrix} \quad V_{ab}^b = \begin{bmatrix} v_{ab}^b \\ \omega_{ab}^b \end{bmatrix} = \begin{bmatrix} R_{ab}^T \dot{p}_{ab} \\ (R_{ab}^T \dot{R}_{ab})^\vee \end{bmatrix}$$

# Interpreting Velocities as Twists

- Can break them apart into  $v$  and  $w$  components
- Calculate each one separately

Quantity	Interpretation
$\omega_{ab}^s$	Angular velocity of $B$ wrt frame $A$ , viewed from $A$ .
$v_{ab}^s$	Velocity of a (possible imaginary) point attached to $B$ traveling through the origin of $A$ wrt $A$ , viewed from $A$ .
$\omega_{ab}^b$	Angular velocity of $B$ wrt frame $A$ , viewed from $B$ .
$v_{ab}^b$	Velocity of origin of $B$ wrt frame $A$ , viewed from $B$ .

# Adjoints

# What are they?

- Like a g matrix for twists!
- Change coordinate frames if we have a twist
- Because velocities are also twists, we can use adjoints to switch between spatial and body velocities

$$\hat{\xi}' = g \hat{\xi} g^{-1}$$

$$v_{AB}^s = Ad_{g_{AB}} \cdot v_{AB}^b$$

$$\xi' = Ad_g \xi$$

# Formulas

$$\underbrace{\begin{bmatrix} R_{ab} & \hat{p}_{ab} R_{ab} \\ 0 & R_{ab} \end{bmatrix}}_{:= Ad_{g_{ab}}}$$

$$Ad_{g_{ab}}^{-1} = \begin{bmatrix} R_{ab}^T & -R_{ab}^T \hat{p} \\ 0 & R_{ab}^T \end{bmatrix}$$

$$V_{ac}^s = V_{ab}^s + Ad_{g_{ab}} V_{bc}^s$$

$$V_{ac}^b = Ad_{g_{bc}^{-1}} V_{ab}^b + V_{bc}^b$$

$$\begin{aligned} & \text{Ad}_{g_{ab}} \cdot \text{Ad}_{g_{bc}} \\ &= \text{Ad}_{(g_{ab} g_{bc})} \end{aligned}$$

# Jacobians and Singularities

# Motivation

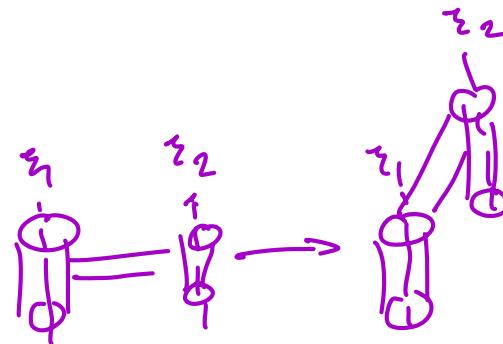
- We want to get the velocity of our **end effector**
- However, our **sensors** give us the **velocities of our links**
- Jacobian allows us to go from **link velocities** → **end effector velocity**

$$V_{st}^s = J_{st}^s(\theta) \dot{\theta}$$

*→ vector, how fast each link moves*

# Spatial Jacobian

- Gets us to the spatial velocity
- Columns of the Jacobian:
  - Twists of each of the links of the robot
  - In their *current* positions (i.e. not at 0 position, unlike FK)
  - Expressed in spatial coordinates
- Column represents derivative of end effector position wrt each of the links



# Formulas

$$\begin{aligned} J_{st}^s(\theta) &= \left[ \left( \frac{\partial g_{st}}{\partial \theta_1} \right)^\vee \quad \dots \quad \left( \frac{\partial g_{st}}{\partial \theta_n} \right)^\vee \right] \\ &= [\xi_1 \quad \xi'_2 \quad \dots \quad \xi'_n] \end{aligned}$$

$$\xi'_i = Ad_{(e^{\widehat{\xi}_1 \theta_1} \dots e^{\widehat{\xi}_{i-1} \theta_{i-1}})} \xi_i$$

$$v_{q_s} = \widehat{V}_{st}^s q_s = (J_{st}^s(\theta) \dot{\theta})^\wedge q_s$$

# Body Jacobian

*body velocity*

- Analogous to spatial Jacobian
- Gets us the body velocity, instead of the spatial velocity
- Each of the twists are represented in the body frame instead

$$J_{st}^b(\theta) = [\xi_1^\dagger \quad \xi_2^\dagger \quad \dots \quad \xi_n^\dagger]$$

$$\xi_i^\dagger = Ad_{(e^{\hat{\xi}_{i+1}\theta_{i+1}} \dots e^{\hat{\xi}_n\theta_n} g_{st}(0))}^{-1} \xi_i$$

$$v_{q_b} = \widehat{V}_{st}^b q_b = (J_{st}^b(\theta) \dot{\theta})^\wedge q_b$$

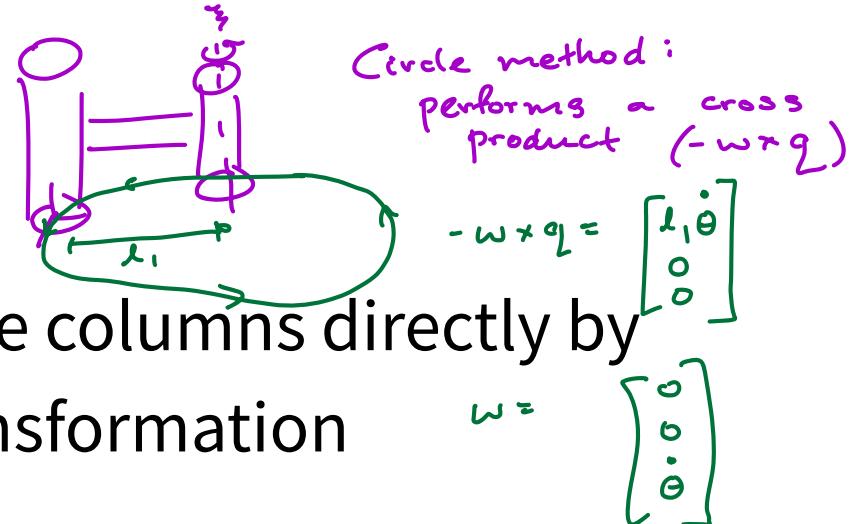
# Conversion

- Jacobians are composed of twists
- Can use the adjoint to move between them!
  - Adjoint is invertible, can go the other way as well

$$J_{st}^s(\theta) = \text{Ad}_{g_{st}(\theta)} J_{st}^b(\theta)$$

# Finding the Jacobian

- Can find the twists making up the columns directly by finding and applying adjoint transformation
- Alternatively, we can calculate the new positions of each of the  $v$  and  $w$  components that make up the twists



← Find new  $\omega, \omega'$   
→ Find new  $q_L, q_R'$

$$q = \begin{bmatrix} -\omega' \times q_L' \\ \omega' \end{bmatrix}$$

# Singularities

$$V_{st}^s = J_{st}^s(\theta) \dot{\theta}$$

- Jacobian drops in rank
  - Manipulability measure  
→ Product of singular values of Jacobian
- We can't reach all of the velocities that we *should* be able to no matter what we set each of our link velocities to
- This is a **singular configuration**
- Would prefer to avoid being in it or near it
  - Can't achieve instantaneous motion in certain directions
  - Could require significant amounts of force in certain directions around that area
  - Mess up error tracking

6 rows {  $\begin{bmatrix} z_1 & \cdots & z_7 \end{bmatrix}$  }

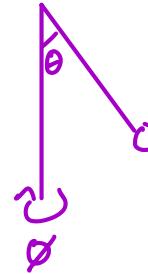


# Dynamics

# Forces!

- In real life, we're trying to control our robot by applying some force to its joints
- Need to get the **dynamics** of our system
- The forces in each direction so that we know exactly what to apply to achieve our trajectory

# Use Energy!



- Forces can be difficult
  - When there are multiple reference frames, particularly rotating ones, in play
  - End up with many complicated terms
  - Sometimes have several “imaginary” forces to balance equations’
- Energy is nice!
  - Scalars
  - Only depends on current state of the object
  - Invariant to coordinate frame - choose any one

Doesn't matter which coord frame chosen  
(stay consistent)

# Method

1. Choose state *→ Generalized coords  $q$*
2. Kinetic energy *→ Use  $q$ ,  $\dot{q}$*
3. Potential energy *↑*
4. Lagrangian  *$L = T - V$*
5. Equations of motion (convert to forces)

$$\underline{\underline{F}} = \frac{d}{dt} \frac{dL}{d\dot{q}} - \frac{dL}{dq}$$

# State

- Depends on the problem at hand
- Choose minimal representation needed or the representation that makes it easiest to determine what forces to apply
- Usually  $p$ ,  $\theta$ , or something similar

↳  $x, y$

# Kinetic Energy

- Translational

$$\frac{1}{2} m \|v\|_2^2$$

- Rotational

$$\frac{1}{2} \omega^\top I \omega$$

$$\frac{1}{2} I \dot{\theta}^2 \rightarrow \text{simpler}$$

$$\begin{matrix} & \begin{matrix} x \\ y \\ z \end{matrix} \\ \begin{matrix} x \\ y \\ z \end{matrix} & \end{matrix} \quad v^b = \begin{bmatrix} 0 \\ 0 \\ 60 \end{bmatrix}$$
$$v^s = \begin{bmatrix} 0 \\ 60 \\ 0 \end{bmatrix}$$

$$\rightarrow T = \frac{1}{2} \sqrt{v_i^b}^\top M_i v_i^b$$

in body frame

$$M_i = \begin{bmatrix} m I_3 & O \\ O & I_k \end{bmatrix}$$

inertia matrix

# Potential Energy

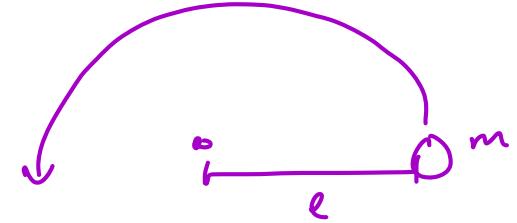
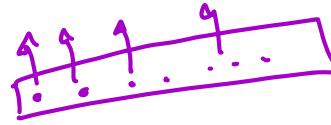
- Gravitational

$mgh$   
↳ height from  $0$  position of our chosen  
coord frame

- Spring

$$\frac{1}{2} kx^2$$

# Lagrangian



→ scalar

$$\frac{1}{2} m (\ell \dot{\theta})^2$$

$$= \frac{1}{2} \cancel{(m \ell^2)} \dot{\theta}^2$$

= Inertia

$$L = T - V = \sum T_i - \sum V_i$$

# Equations of Motion

$$\tau = \frac{d}{dt} \frac{\partial L}{\partial \dot{q}} - \frac{\partial L}{\partial q}$$

vector w/ same dims as generalized coords

Generalized forces  
External - comes from motors,  
friction, etc.

# Separation (Optional)

$$\Upsilon = M(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q)$$

mass /  
inertia  
matrix

Coriolis  
matrix

"Imaginary"  
forces from  
rotating coord frames

Constant  
forces  
ex: gravity

# Control



# Dynamical Systems

$$\dot{x} = f(x, u)$$

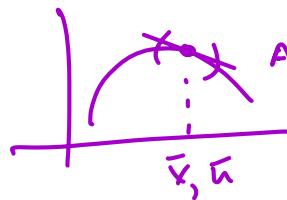
- Equations used to represent our system based on current state and input
- Often in the form of a differential equation
- Generated with knowledge of dynamics
  - State evolves as a result of forces
  - Input is the forces we add into the system

\* Equil. Point:  $\dot{x} = 0$

# Linearization

*convert to a linear system*

$$\dot{x} = f(x, u) \rightarrow \dot{x} \approx A\Delta x + B\Delta u + l(\bar{x}, \bar{u})$$



Approximate function as linear in a small area  
about the linearization point

$$f(x, u) \approx \underbrace{\frac{df}{dx} \Bigg|_{\bar{x}, \bar{u}}}_{\text{Like our slope}} (x - \bar{x}) + \frac{df}{du} \Bigg|_{\bar{x}, \bar{u}} (u - \bar{u}) + f(\bar{x}, \bar{u})$$

$Mx$       +       $B$

# LTI Systems

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

# Controllability

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

Can we get our system to want?

Assume wLOG  $x_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

$$\begin{aligned} x_1 &= Ax_0 + Bu_1 \\ &= Bu_1 \\ x_2 &= Ax_1 + Bu_2 \\ &= A(Bu_1) + Bu_2 \\ &= A^2Bu_1 + ABu_2 \\ &\quad + Bu_3 \\ &= A^2Bu_1 + ABu_2 \\ &\quad + Bu_3 \end{aligned}$$

behave the way we want

Controllability Matrix:

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

$n$  is the dim. in of our state  
 $\text{span}(AB, A^2B, \dots, B)$

Controllable within the span of  $Q$

$Q$  full rank: fully controllable

Ex.

$$Q = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$$

$$\text{span}(Q) = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

we can get system to go from

$$\begin{bmatrix} 4 \\ 5 \end{bmatrix} \rightarrow \begin{bmatrix} 7 \\ 8 \end{bmatrix}$$

we can't

$$\begin{bmatrix} 4 \\ 5 \end{bmatrix} \rightarrow \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

$x_f - x_i$  is not in  $\text{span}(Q)$

# Stability

LTI system  $\dot{x} = Ax + Bu$  stable:

$$\text{All } \operatorname{Re}(\operatorname{eig}(A)) < 0$$

Stabilizable if  $H$  is controllable in the direction of the eigenvectors w/ nonnegative eigenvalues

# PID Control

- Used to error correct and can follow trajectory to some small extent
- Model-free control - only need to know error, not system equations

$$u(t) = K_p e(t) + K_i \int_0^t e(\tau) d\tau + K_d \dot{e}(t)$$

*→ Main error correction*

*→ Fix steady-state error*

*→ Reduces oscillation*

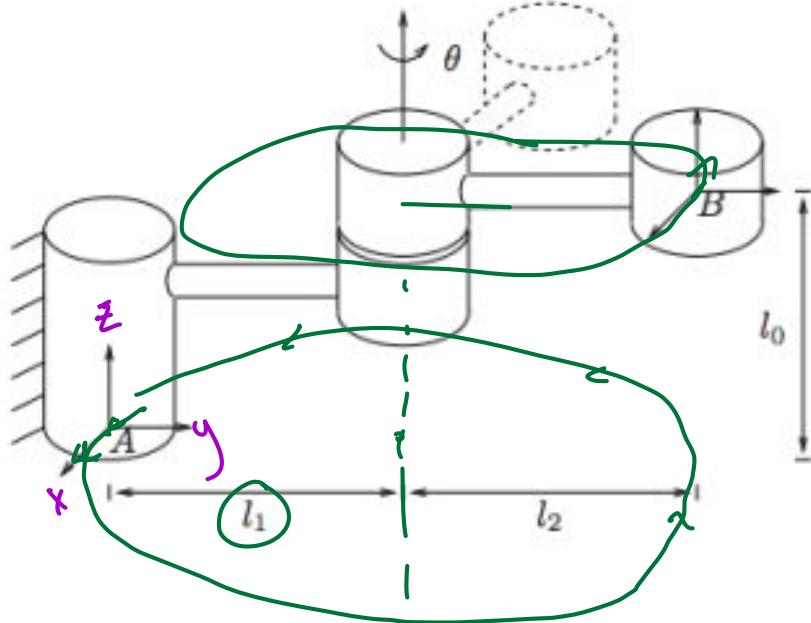
# The Terms

- Proportional
  - Workhorse
  - Applies input that pulls state towards desired trajectory
- Derivative
  - Dampens proportional response
  - Prevents oscillation and overcorrection
  - Allows for convergence
- Integral
  - Corrects steady-state error because of constant forces like  $g$

# Feedback Linearization

- Incorporate error into the input term of a linear system
- Set up the equations so that error converges to 0

# Calculate Spatial + Body Velocity



Spatial:

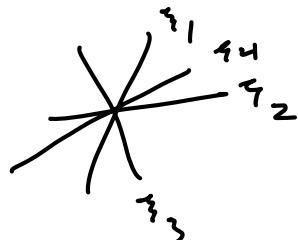
$$\begin{bmatrix} l_1 \dot{\theta} \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Body:

$$\begin{bmatrix} -l_2 \dot{\theta} \\ 0 \\ 0 \\ 0 \\ 0 \\ \dot{\theta} \end{bmatrix}$$

revolute

# Show that a manipulator with 4 intersecting joints will have a singularity.



WLOG, assume these joints are @  
origin  
 $\rightarrow -\omega \times q = -\omega \tau 0 = 0$

$$J^s = \begin{bmatrix} \ell_1 & \ell_2' & \ell_3' & \ell_4' \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ \omega_1 & \omega_2' & \omega_3' & \omega_4' \end{bmatrix}$$

Max rank of 3  
we have 4 joints though  
 $\Rightarrow$  Jacobian could be rank 4

# Calculate the dynamics

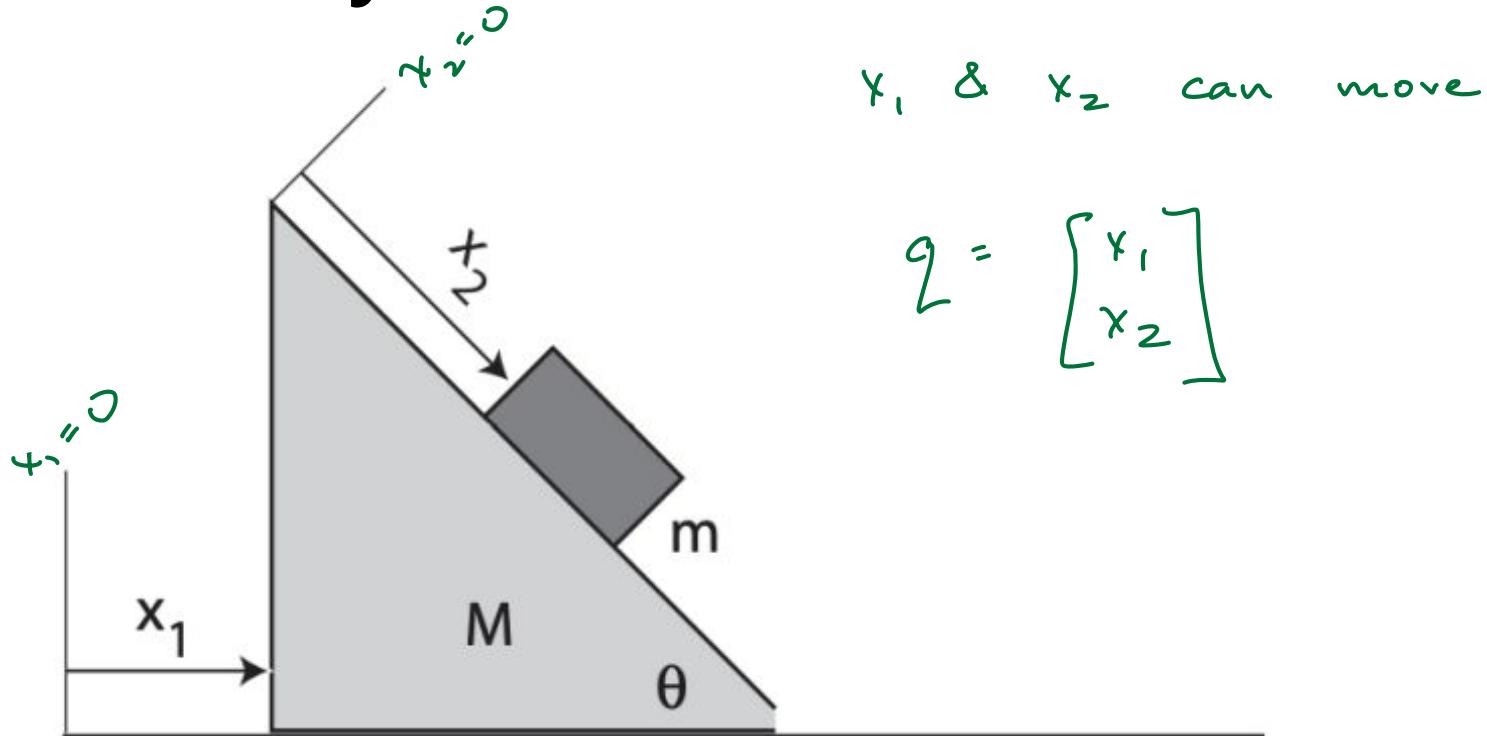


Figure 1: Image sourced from [http://www.dzre.com/alex/P441/lectures/lec\\_18.pdf](http://www.dzre.com/alex/P441/lectures/lec_18.pdf)

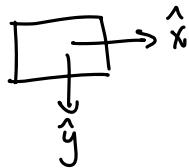
$$\textcircled{1} \quad q = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

\textcircled{2} KE

Ramp moving  $\rightarrow$

$$\frac{1}{2} M \dot{x}_1^2$$

Object moving  $\rightarrow$  velocities from  $\dot{x}_1$  &  $\dot{x}_2$



$$v_{\hat{x}} = \dot{x}_1 + \dot{x}_2 \cos \theta$$

$$v_{\hat{y}} = \dot{x}_2 \sin \theta$$

$$T_m = \frac{1}{2} m (\dot{x}_1 + \dot{x}_2 \cos \theta)^2 + \frac{1}{2} m (\dot{x}_2 \sin \theta)^2$$

Expanding expression  
 $\sin^2 + \cos^2 = 1$

$$= \frac{1}{2} m (\dot{x}_1^2 + 2\dot{x}_1 \dot{x}_2 \cos \theta + \dot{x}_2^2)$$

$$T = T_m + T_m$$

\textcircled{3} PE: box on ramp

$$V_g = -mg \underbrace{(x_2 \sin \theta)}_{\text{height}}$$

\textcircled{4} Lagrangian

$$L = T - V$$

$$= T_m + T_m - V$$

$$= \frac{1}{2} M \dot{x}_1^2 + \frac{1}{2} m (\dot{x}_1^2 + 2\dot{x}_1 \dot{x}_2 \cos \theta + \dot{x}_2^2) + mg x_2 \sin \theta$$

$$\textcircled{5} \quad \ddot{r} = \frac{d}{dt} \frac{dL}{dq} - \frac{dL}{dq}$$

$$\frac{dL}{d\dot{x}_1} = M \dot{x}_1 + m \dot{x}_1 + m \dot{x}_2 \cos \theta \rightarrow \frac{d}{dt} = M \ddot{x}_1 + m \ddot{x}_1 + m \ddot{x}_2 \cos \theta$$

$$\frac{dL}{d\dot{x}_2} = m \dot{x}_1 \cos \theta + m \dot{x}_2 \rightarrow \frac{d}{dt} = m \ddot{x}_1 \cos \theta + m \ddot{x}_2$$

$$\frac{dL}{dx_1} = 0$$

$$\frac{dL}{dx_2} = mg \sin \theta$$

$$\begin{bmatrix} I_{x_1} \\ x_{x_2} \end{bmatrix} = \begin{bmatrix} m\ddot{x}_1 + m\ddot{x}_1 + m\ddot{x}_2 \cos \theta \\ m\ddot{x}_1 \cos \theta + m\ddot{x}_2 + mg \sin \theta \end{bmatrix}$$

Say we have a system with  $x_1, x_2, u_1, u_2$ . Linearize the system at an equilibrium point of  $(0, 0)$ , and analyze the stability and controllability:

$$\dot{x}_1 = 2x_1^2 + 3x_2 + u_1^2 = f_1$$

$$\dot{x}_2 = \sin x_1 + x_2 = f_2$$

$$f(x, u) - f(\bar{x}, \bar{u}) = \delta f \approx \frac{\partial f}{\partial x} \Big|_{\substack{x=\bar{x} \\ u=\bar{u}}} (x - \bar{x}) + \frac{\partial f}{\partial u} \Big|_{\substack{x=\bar{x} \\ u=\bar{u}}} (u - \bar{u})$$

$$\left. \frac{\partial f}{\partial x} \right|_{\substack{x=\bar{x} \\ u=\bar{u}}} = \underbrace{\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}}_{\substack{x=\bar{x} \\ u=\bar{u}}} = \begin{bmatrix} 4x_1 \\ \cos x_1 \end{bmatrix} \Bigg|_{\substack{x=\bar{x} \\ u=\bar{u}}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

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

$$\frac{\partial f}{\partial u} = \begin{bmatrix} \frac{\partial f_1}{\partial u_1} & \frac{\partial f_1}{\partial u_2} \\ \frac{\partial f_2}{\partial u_1} & \frac{\partial f_2}{\partial u_2} \end{bmatrix} \Bigg|_{\substack{x=\bar{x} \\ u=\bar{u}}} = \begin{bmatrix} 2u_1 & 0 \\ 0 & 0 \end{bmatrix} \Bigg|_{\substack{x=\bar{x} \\ u=\bar{u}}} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\dot{x}_1 = 2x_1^2 + 3x_2 + 2u_1 = f_1$$

$$\dot{x}_2 = \sin x_1 + x_2 = f_2$$

$$f(x, u) - f(\bar{x}, \bar{u}) = \delta f \approx \frac{\partial f}{\partial x} \Bigg|_{\substack{x=\bar{x} \\ u=\bar{u}}} (x - \bar{x}) + \frac{\partial f}{\partial u} \Bigg|_{\substack{x=\bar{x} \\ u=\bar{u}}} (u - \bar{u})$$

$$\frac{\partial f}{\partial x} \Bigg|_{\substack{x=\bar{x} \\ u=\bar{u}}} = \underbrace{\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}}_{\substack{x=\bar{x} \\ u=\bar{u}}} = \begin{bmatrix} 4x_1 & 3 \\ \cos x_1 & 1 \end{bmatrix} \Bigg|_{\substack{x=\bar{x}=0 \\ u=\bar{u}=0}} = \begin{bmatrix} 0 & 3 \\ 1 & 1 \end{bmatrix}$$

$$\frac{\partial f}{\partial u} \Bigg|_{\substack{x=\bar{x} \\ u=\bar{u}}} = \begin{bmatrix} \frac{\partial f_1}{\partial u_1} & \frac{\partial f_1}{\partial u_2} \\ \frac{\partial f_2}{\partial u_1} & \frac{\partial f_2}{\partial u_2} \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} \Bigg|_{\substack{x=\bar{x}=0 \\ u=\bar{u}=0}} = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$$

FINAL LINEARIZATION:

$$\begin{aligned} \bar{f}_1 &= 0 + 0 + 0 \rightarrow \text{Evaluated @ linearization point} \\ \bar{f}_2 &= 0 + 0 \end{aligned}$$

$$\delta f = f - \bar{f} = f \approx \begin{bmatrix} 0 & 3 \\ 1 & 1 \end{bmatrix} (x - \bar{x}) + \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} (u - \bar{u})$$

$$\begin{bmatrix} x_1 - \bar{x}_1 \\ x_2 - \bar{x}_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

## Stability

$$(0 - 2)(1 - 2) - (3)(1) = 0$$

$$0 - \lambda + \lambda^2 - 3 = 0$$

$$(\lambda - 2)(\lambda + 1) = 0$$

$\lambda = \{2, -1\} \rightarrow$  Unstable b/c we have a nonnegative eigenvalue

## Controllability

$n = 2$  (we have 2 states)

$$Q = \begin{bmatrix} B & AB \end{bmatrix}$$

$$B = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$$

$$AB = \begin{bmatrix} 0 & 3 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 2 & 0 \end{bmatrix}$$

$$Q = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 \end{bmatrix}$$

$$\text{Rank} = 2$$

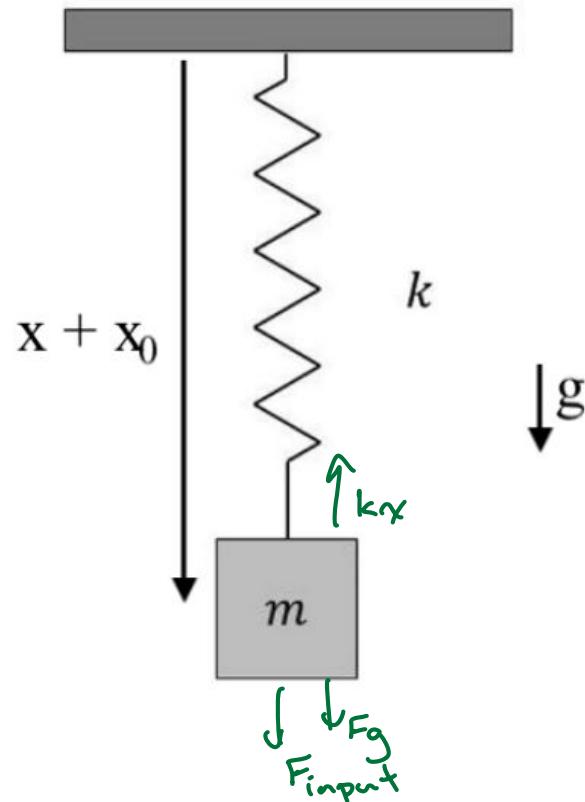
Fully rank  $\Rightarrow$  Fully controllable !

## Stabilizability

Can we stabilize in the direction of the eigenvector(s) w/ positive eigenvalue(s)?  $\rightarrow$  Can apply control input to keep system stable

$\rightarrow$  YES b/c we are fully controllable ( $Q$  is full rank)

# Calculate a control law that causes error to go to 0



$$m\ddot{x} = mg - kx + F_{\text{input}}$$

Solve for input achieves  $\ddot{x}_d$

Add feedback term for error correction

$$F_{\text{input}} = \ddot{x}_d - mg + kx + m(k_p e + k_d \dot{e})$$

$$\ddot{x} = mg - kx + (\ddot{x}_d - mg + kx + m(k_p e + k_d \dot{e}))$$

$$= m\ddot{x}_d - m\dot{x} + m(K_p e + K_d \dot{e})$$

$$0 = m(\ddot{e} + K_p e + K_d \dot{e})$$



error will converge to 0 ✓