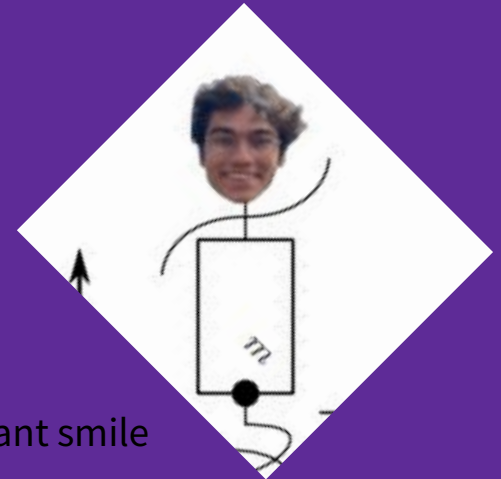


# EECS/MechE/BioE C106A: Midterm 2 Review Session

The return of Prof. Tarun Amarnath!



Look at Sunay's brilliant smile

**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)

$$g = \begin{bmatrix} R & P \\ 0 & 1 \end{bmatrix}$$

# 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)

Twists  
Axis of rotation

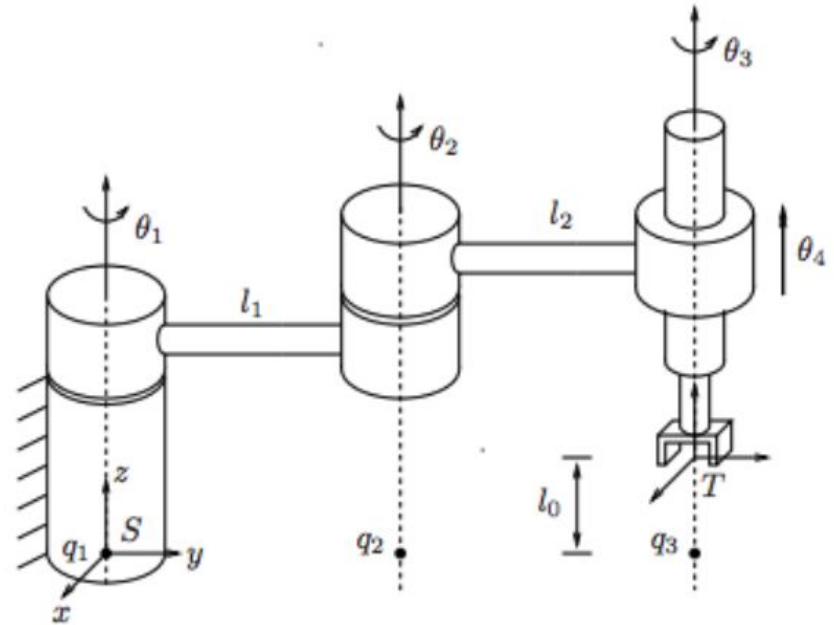
$$g(t) = e^{\hat{\xi}t}$$

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

# Forward Kinematics

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

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



# Inverse Kinematics

→ Given some final position  $g_{st}(t)$   
→ Find  $\theta$

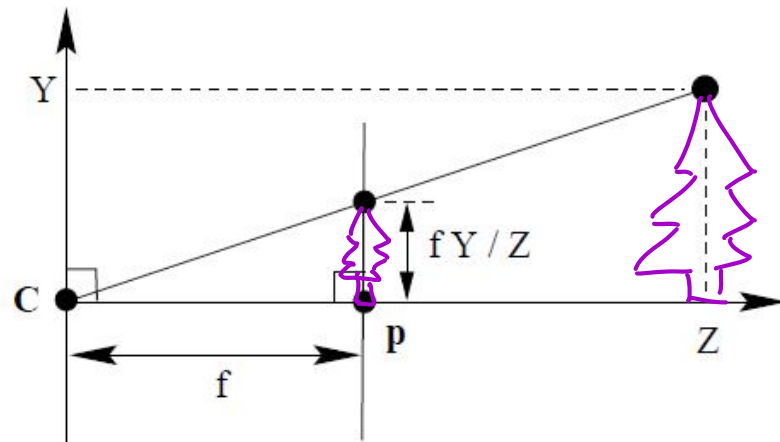
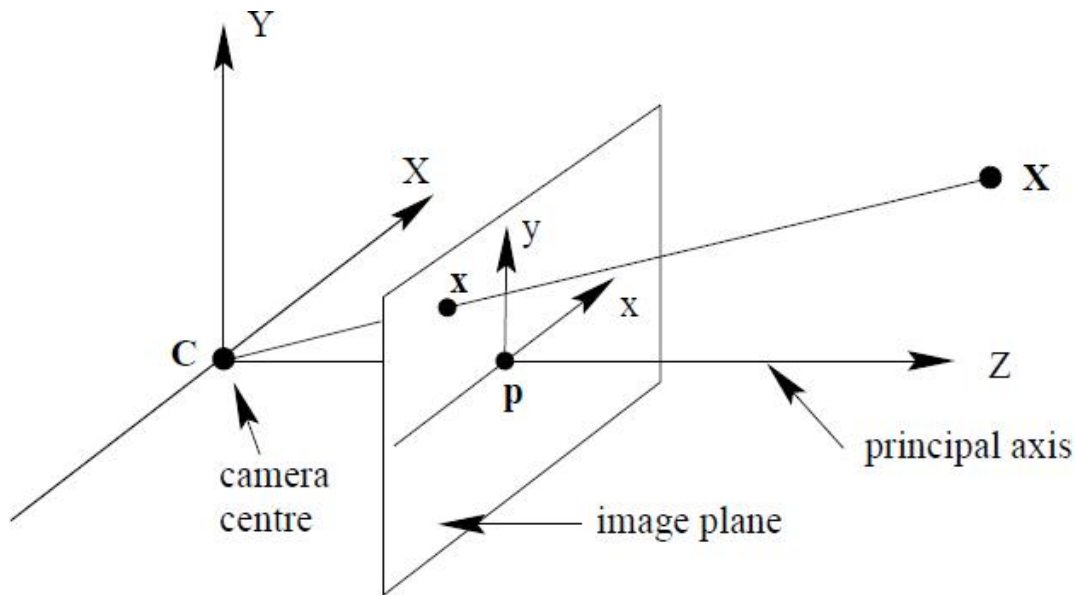
- How do we move our robot's joints to reach a desired configuration?
- Use Paden-Kahan subproblems along with tricks (reduce problem down to simpler parts)

# Computer Vision



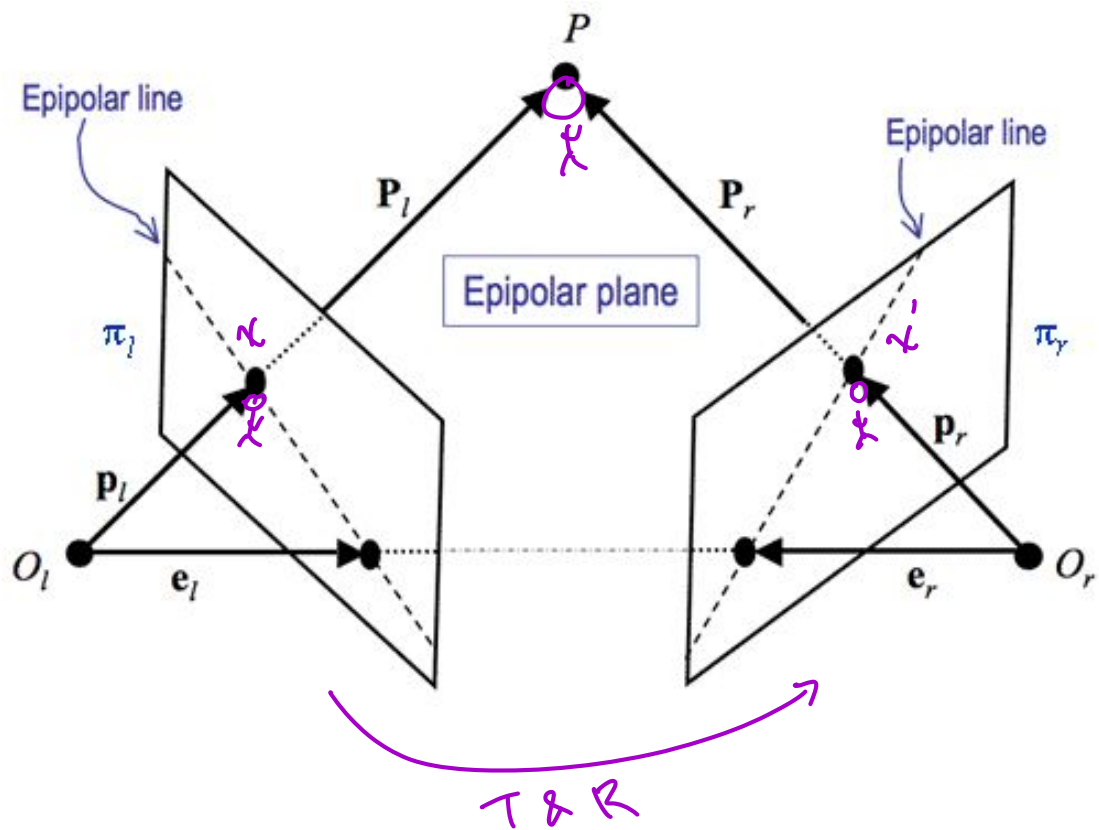


# Pinhole Camera Model



$$(X, Y, Z)^T \longrightarrow \left(\frac{fX}{Z}, \frac{fY}{Z}\right)^T$$
$$K = \begin{bmatrix} f_x & 0 & o_x \\ 0 & f_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$

# Two-View Geometry



$$\begin{pmatrix} x \\ y \end{pmatrix}^T E x = 0$$

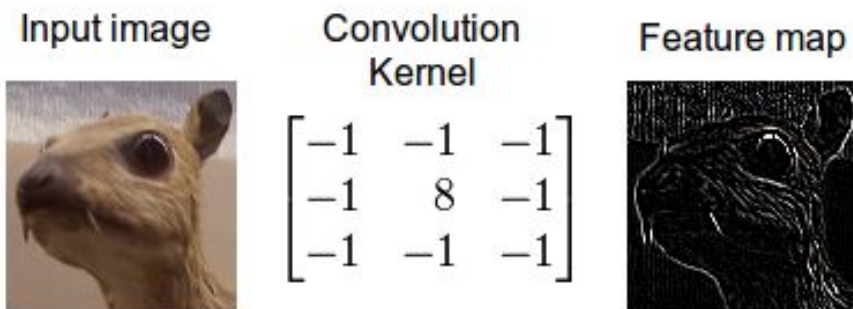
Point location in img 2 in img coords

$$= \hat{T} R$$

- Can determine depth info

# Convolutions

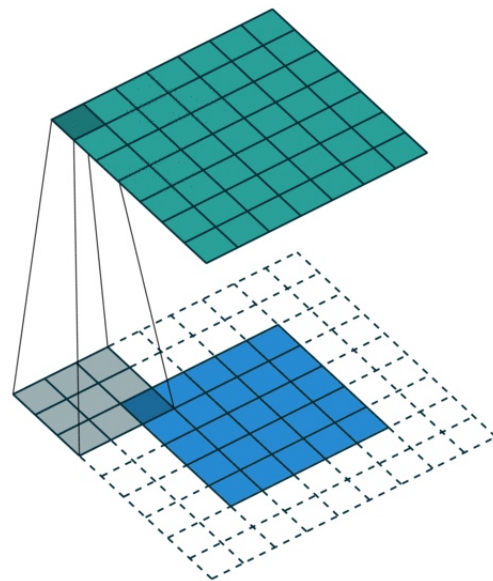
- Slide a kernel over some image
- Understand some information about the picture



Original image



Gaussian Blur filter applied



# Homography

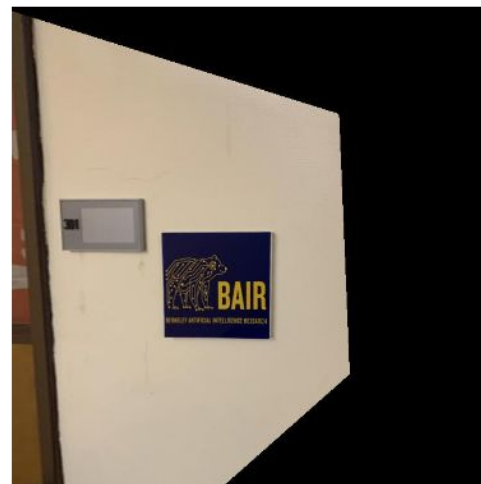
- Apply some kind of affine transformation to an image
- Change perspectives - for example, can straighten a picture
- Need at least 4 pairs of points to do this



Original sign



Sign points



Straightened sign

# 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
- Can apply them to a specific point to determine that point's velocity

$$\rightarrow v^S = \begin{bmatrix} v \\ \omega \end{bmatrix}$$

$$v_q = \hat{V}_{ab}^S \cdot P_a$$

# Spatial Velocity

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

$\frac{d}{dt}$

$q_a(t) = g_{ab} \cdot q_b$

$\dot{q}_a(t) = \dot{g}_{ab} \cdot q_b$

Switch frames

$\dot{q}_a(t) = \underbrace{\dot{g}_{ab} \cdot g_{ab}^{-1}}_{\hat{V}_{ab}^s} \cdot q_a$

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



# Body Velocity

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

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

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

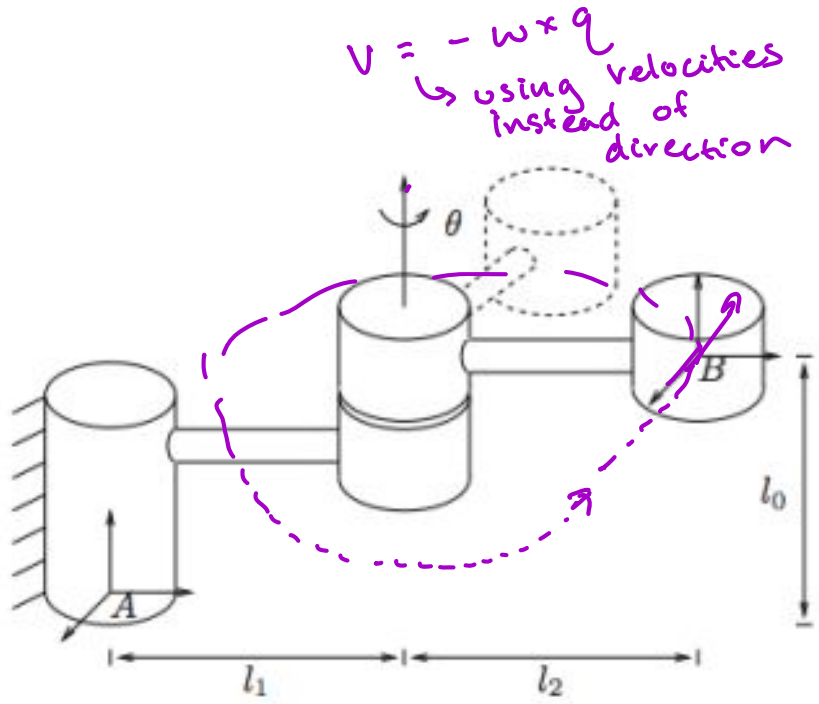
$$V_{ab}^b = \begin{bmatrix} v_{ab}^b \\ \omega_{ab}^b \end{bmatrix} = \begin{bmatrix} R_{ab}^T \dot{p}_{ab} \\ (R_{ab}^T \dot{R}_{ab})^\vee \end{bmatrix}$$

# 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 $B$ .

# Review Example



$v = -w \times q$   
 using velocities instead of direction

$$v^S = \begin{bmatrix} v \\ w \end{bmatrix} = \begin{bmatrix} l_1 \dot{\theta} \\ 0 \\ 0 \\ 0 \\ 0 \\ \dot{\theta} \end{bmatrix}$$

linear speed of origin attached to body frame

angular speed

$$v^B = \begin{bmatrix} v \\ w \end{bmatrix} = \begin{bmatrix} -l_2 \dot{\theta} \\ 0 \\ 0 \\ 0 \\ 0 \\ \dot{\theta} \end{bmatrix}$$

linear speed of body (in body frame)

angular speed

$v_q = \begin{bmatrix} v \\ w \end{bmatrix} \cdot q$

# 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

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

$$V_{ab}^s = \text{Ad}_{g_{ab}} \cdot V_{ab}^b$$

$$\xi' = \text{Ad}_g \xi$$

# Formulas

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

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

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

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

$$\begin{aligned} & Ad_{g_{ab}} \cdot Ad_{g_{bc}} \\ & = Ad_{(g_{ab} g_{bc})} \end{aligned}$$

# 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**

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



# 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

$$\begin{aligned} J_{st}^s(\theta) &= \left[ \left( \frac{\partial g_{st}}{\partial \theta_1} \right)^\vee \quad \dots \quad \left( \frac{\partial g_{st}}{\partial \theta_n} \right)^\vee \right] \rightarrow \text{Partial derivatives} \\ &= [\xi_1 \quad \xi_2' \quad \dots \quad \xi_n'] \rightarrow \text{Twists} \end{aligned}$$

$$\xi_i' = \text{Ad}_{(e^{\hat{\xi}_1 \theta_1} \dots e^{\hat{\xi}_{i-1} \theta_{i-1}})} \xi_i$$

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

# Body Jacobian

- 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

$$J_{st}^s(\theta) = Ad_{g_{st}(\theta)} J_{st}^b(\theta)$$

# 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

# Singularities

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


- Jacobian drops in rank
- 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

# 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

# Method

1. Choose state  $\rightarrow q$
2. Kinetic energy
3. Potential energy
4. Lagrangian
5. Equations of motion (convert to forces)
6. Separate into matrices

# 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

# Kinetic Energy

- Translational

$$\frac{1}{2} m \cdot v^2 \rightarrow \dot{q}$$

- Rotational

$$\frac{1}{2} \cdot I \cdot \dot{\theta}^2$$

$$\rightarrow T = \frac{1}{2} V_i^b M_i \cdot V_i^b$$

*6D twist*

$$M_i = \begin{bmatrix} m \cdot I_3 & 0 \\ 0 & I \end{bmatrix}$$

*Identity*

*Inertia matrix  $\rightarrow 3 \times 3$*

# Potential Energy

- Gravitational

$$mgh$$

height from our 0 position

- Spring

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

# Lagrangian

★ No matrices or vectors  
→ scalar

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

# Equations of Motion

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

End up with vectors

$\Upsilon$  - vector w/ the same dimension as state  
- Amt. of force applied on each state variable

# Separation

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

Mass/inertia  
matrix

Coriolis

imaginary  
spinning

forces from  
reference frames

Constant  
forces



**Control**



# Trajectories

- Define how we want our robot to move
- Precomputed

$q, \dot{q}$

# Realistic Motion

- Apply some control input ( $u$ ) to follow trajectory
  - Feedforward control
- Friction, inefficiencies, and other real world issues create problems
- Adjust input to fix errors
  - Feedback

# Systems

- Equations used to represent relationships between state variables
- Also incorporate control input
- Generated with knowledge of dynamics

$$\dot{x} = f(x) + g(x) \cdot u$$

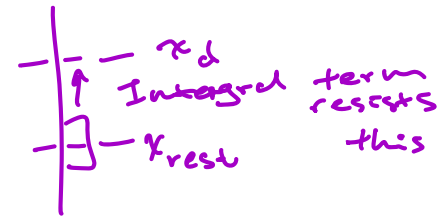
# PID Control

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

$$u(t) = \underbrace{K_p e(t)}_{\text{Proportional}} + \underbrace{K_i \int_0^t e(\tau) d\tau}_{\text{Integral}} + \underbrace{K_d \dot{e}(t)}_{\text{Derivative}}$$

# 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  $g$



↳ Supplies force to stay at 0 error

# Model-Based Control

- Uses system dynamics
- Much better inputs to control state
  - PIDs might estimate error with position
  - But input might be acceleration - not ideal
- Feedforward control - determine beforehand what the input should be

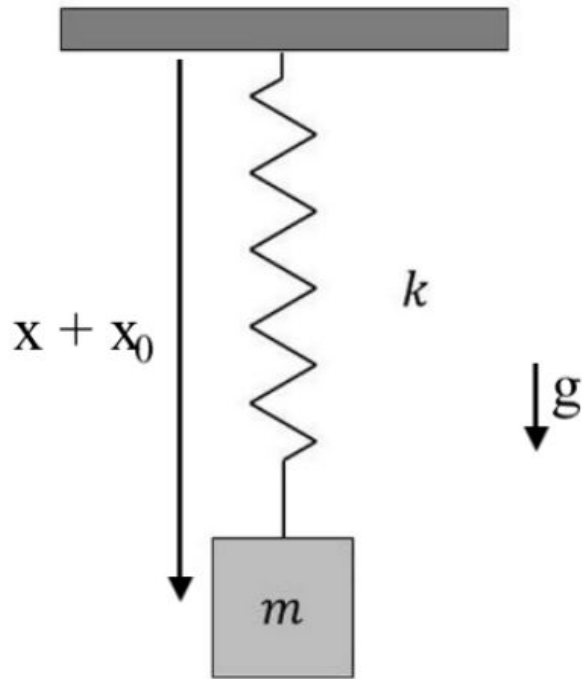
$$u = u_{ff} + u_{fb}$$

Feedforward

Feedback

PIDs, some error correction

# An Example



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

↓  
Solve for input force  
achieve  $\ddot{x}_d$

↓  
Add a feedback term  
using PID control

$$m \ddot{x}_d = m \cdot g - kx + F_{\text{input}}$$

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

$$u = F_{\text{input}} + K_p \cdot e + K_d \dot{e} + K_i \int e dt$$



# Feedback Linearization

→ Feedforward control  
even though its called  
feedback

- Setup input in such a way that we can directly plug in our trajectory as a control input

$$\dot{x} = f(x) + g(x) \cdot u$$
$$u = -g^{-1}(x) \cdot f(x) + g^{-1}(x) \cdot \dot{x}_d$$

Input is a function  
of desired  
trajectory

$$\dot{x} = \dot{x}_d$$

**Lab**



Goal: Prep for job interviews!  
Work in industry!

- Make sure you're familiar with the basic setup operations!
- Sourcing, making, etc.
- Nodes, topics, publishers, subscribers
- Creating packages, running programs
- Types of communication protocols — server/client
- ROS parameter server
- Bashrc
- Work done in labs (planning, tracking, mapping, etc.)