

CIS 3990

# Mobile and IoT Computing

<https://penn-waves-lab.github.io/cis3990-24spring>

## Lecture 7: Inertial Sensors & Inertial Sensing

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# Sensing Modalities

- Radio signals (EM waves): GPS, Cellular, Bluetooth, WiFi
- Ultrasound signals (mechanical waves): smart speakers
- Visual sensors: cameras, LIDAR
- Inertial sensors

→ **Focus of this lecture**



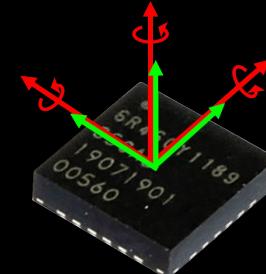
Radio



Acoustic



Visual



Inertial

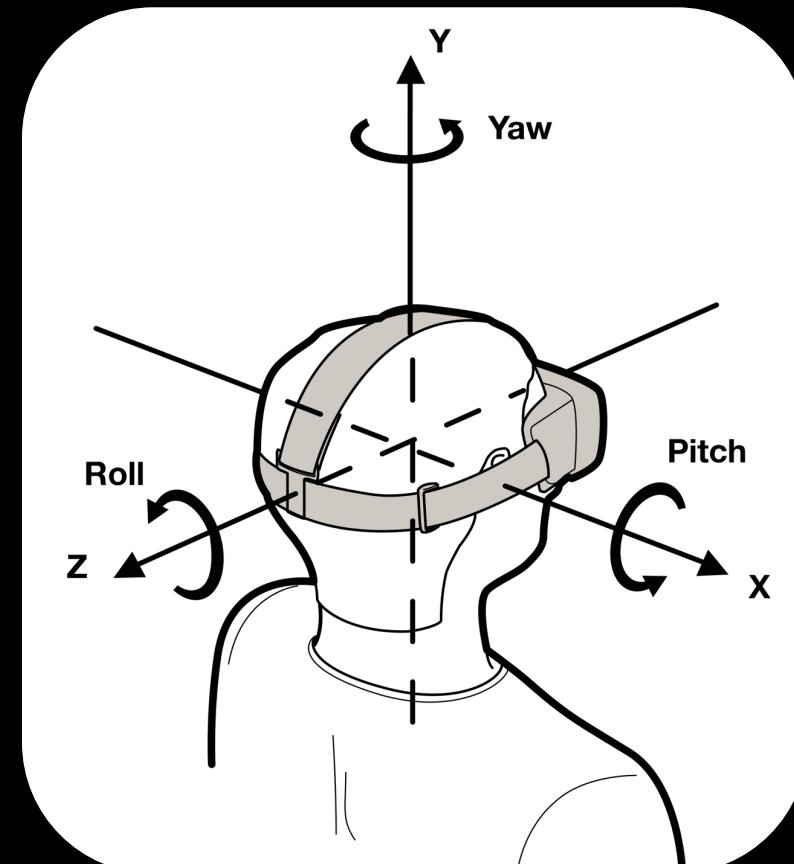
# Objectives of This Module

**Learn how foundational sensing technologies can be used to extract diverse and meaningful insights**

1. What are important application areas of Mobile and IoT sensing?
2. What are the **foundational sensing mechanisms** and how are they related to localization?
3. What processing algorithms can be used to transform raw sensor data?
4. Example sensing systems/solutions with **real-world case studies**.

# Let's understand inertial sensing in the context of VR

- **Goal:** track location and orientation of head or other device
- **Coordinates:** Six degrees of freedom:
  - Cartesian frame of reference (x, y, z)
  - Rotations represented by Euler angles (yaw, pitch roll)

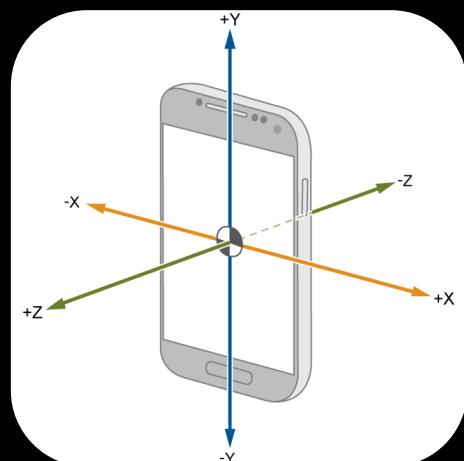


# What are Inertial Sensors?

# Inertial Sensors: Accelerometer

- **Newton's first law of motion (aka The Principle of Inertia):**
  - + Every object perseveres in its state of rest, or of uniform motion in a right line, except insofar as it is compelled to change that state by forces impressed thereon.
  - + Conservation of momentum in the absence of external forces.
- **Inertial sensors** capture the changes:

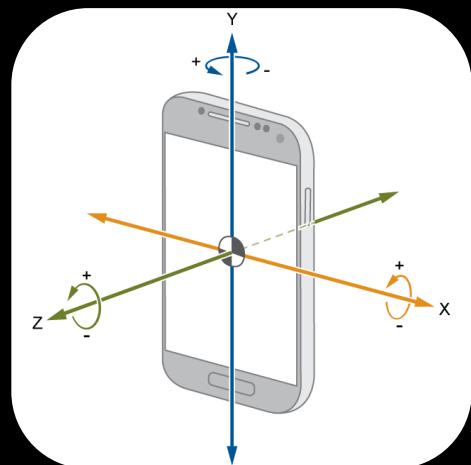
**Accelerometer** measures linear acceleration  $\mathbf{a}$  in  $\text{m/s}^2$



# Inertial Sensors: Gyroscope

- **Newton's laws for rotation (analogous to laws for motion)**
  - + A spinning object tends to spin with a constant angular velocity.
  - + Conservation of angular momentum.
- **Inertial sensors** capture the changes:

**Gyroscope** measures angular velocity  $\omega$  in degrees/s



# Inertial Sensors: Magnetometer

- A magnetometer's operation is based on the interaction between magnetic fields and electric charges, which is described by Maxwell's equations – fundamental laws of electromagnetism.
- These sensors detect magnetic fields, including the Earth's geomagnetic field, allowing us to determine direction relative to the magnetic North Pole.

**Magnetometer** measures magnetic field strength  $\mathbf{m}$  in  $\mu\text{T}$  (micro-Teslas).

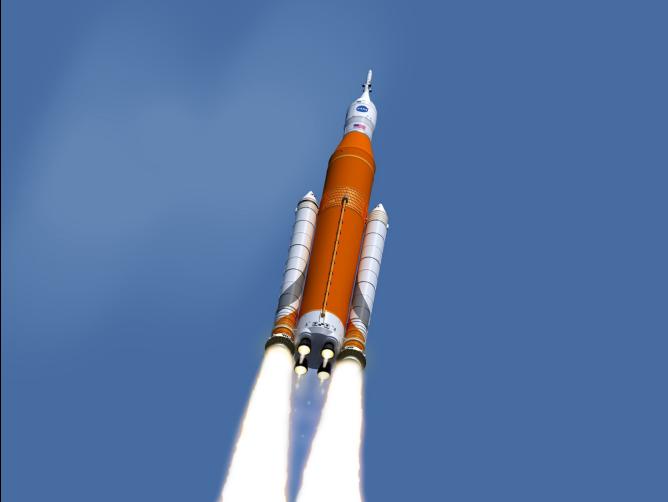


# IMU: Inertial Measurement Unit

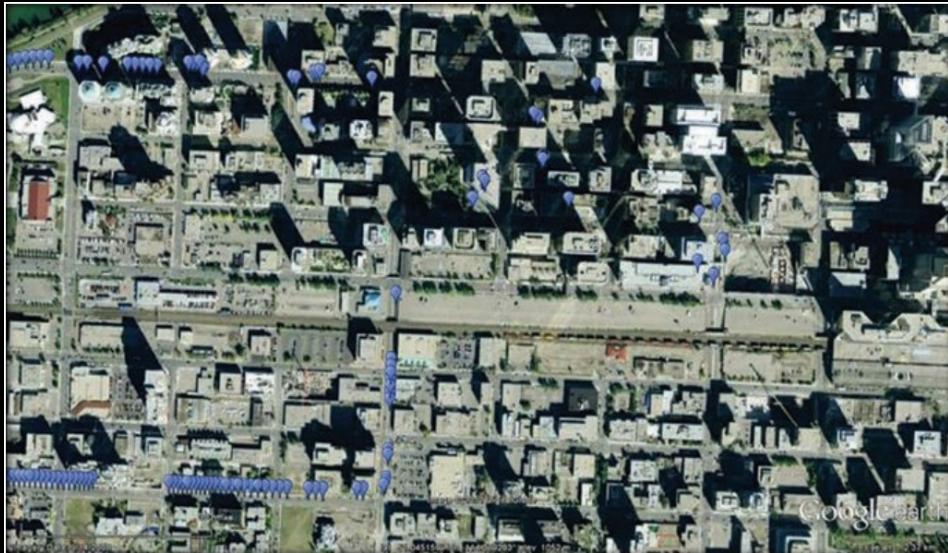
IMU consists of:

- **Gyroscope** measures angular velocity  $\omega$  in degrees/s
- **Accelerometer** measures linear acceleration  $a$  in m/s<sup>2</sup>
- **Magnetometer** measures magnetic field strength  $m$  in  $\mu\text{T}$  (micro-Teslas).

# Where are IMUs used today?



# Example Application: Inertial Navigation



GPS only

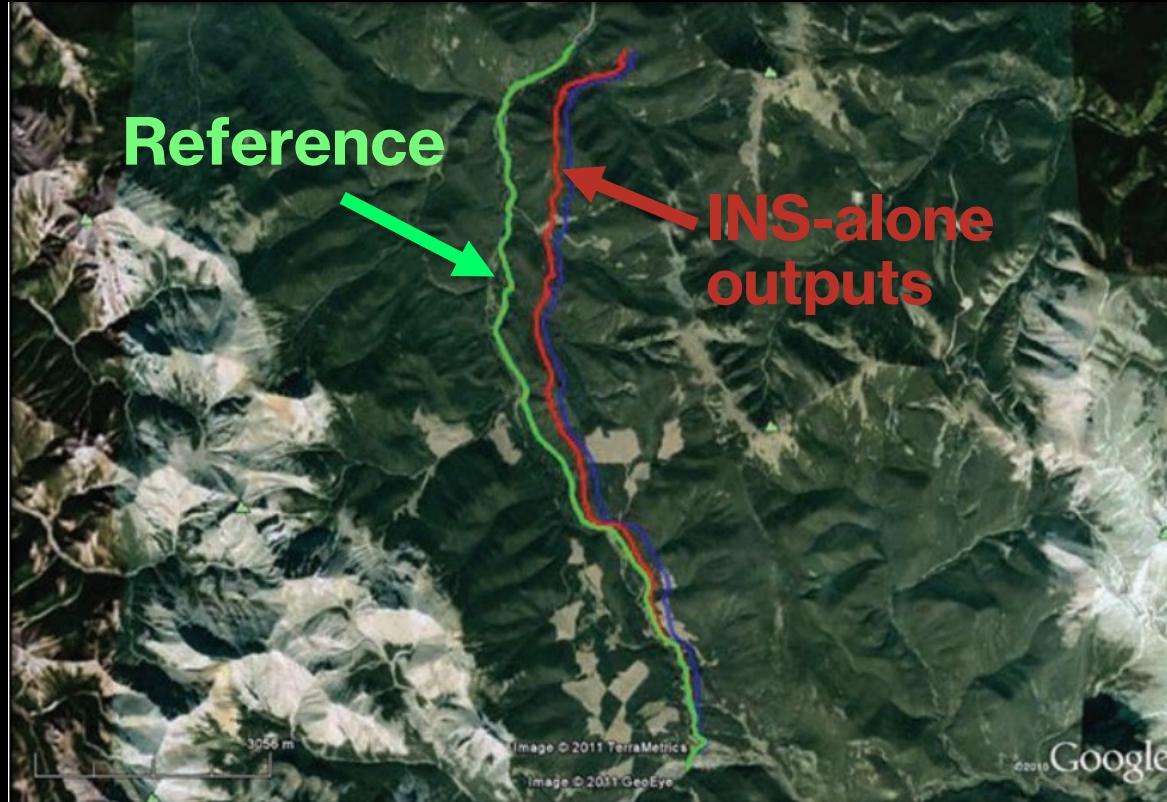


GPS+INS

Key Idea: Integrate acceleration data over time to discover location (Inertial Sensing)

# Inertial Sensing alone is not enough for accurate positioning

Errors accumulate over time



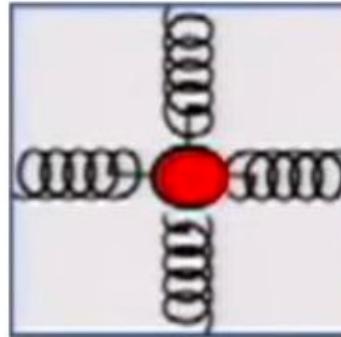
Key Idea: Fuse Data from Multiple Sensors  
(Sensor Fusion)

# Rest of this Lecture

- Basic principles of operation of different IMU sensors
- Understanding Sources of Errors
- Dead reckoning by fusing multiple sensors
- Example system: Pothole Patrol

# Accelerometer

Mass on spring



Gravity  
 $1g = 9.8m/s^2$

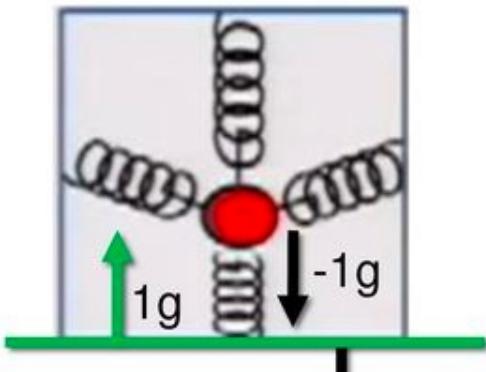
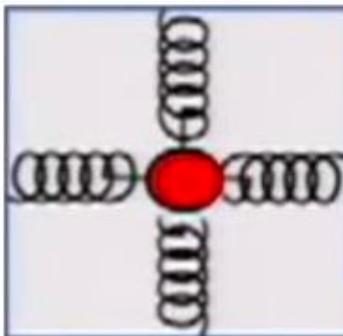
Free Fall

Linear Acceleration

Linear  
Acceleration  
plus gravity

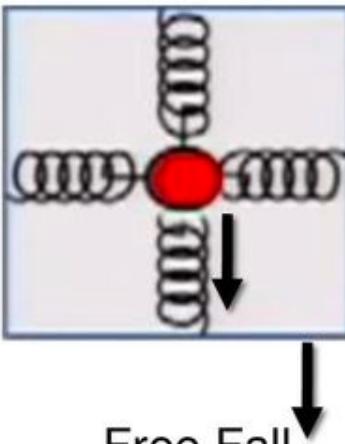
# Accelerometer

Mass on spring

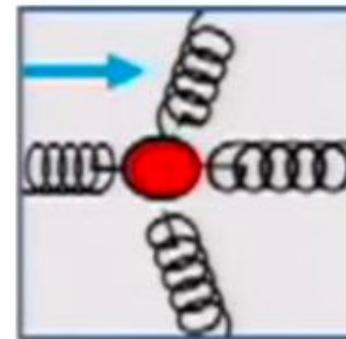


Gravity

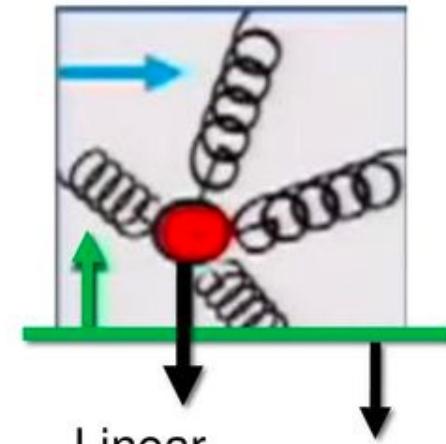
$$1g = 9.8 \text{m/s}^2$$



Free Fall

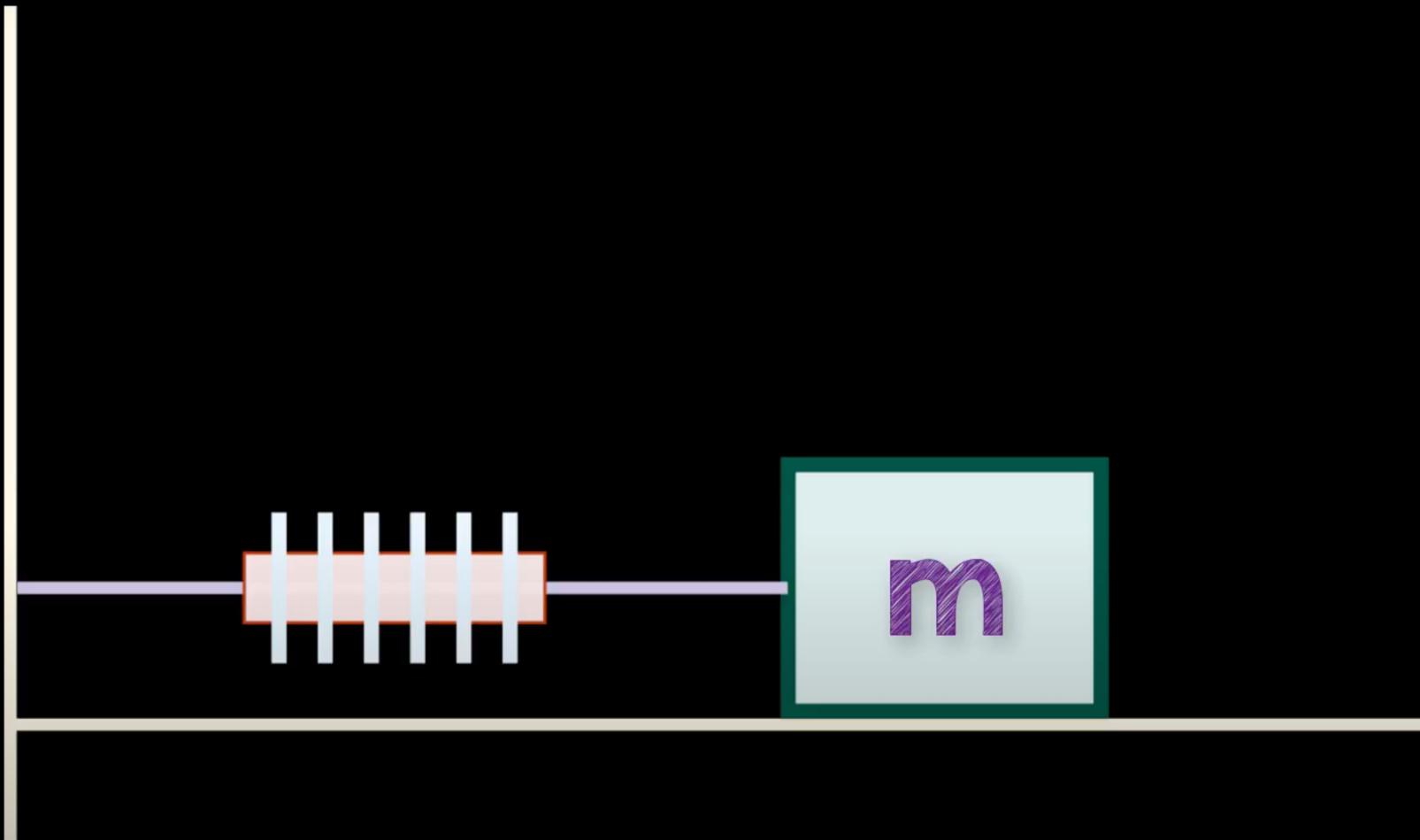


Linear Acceleration

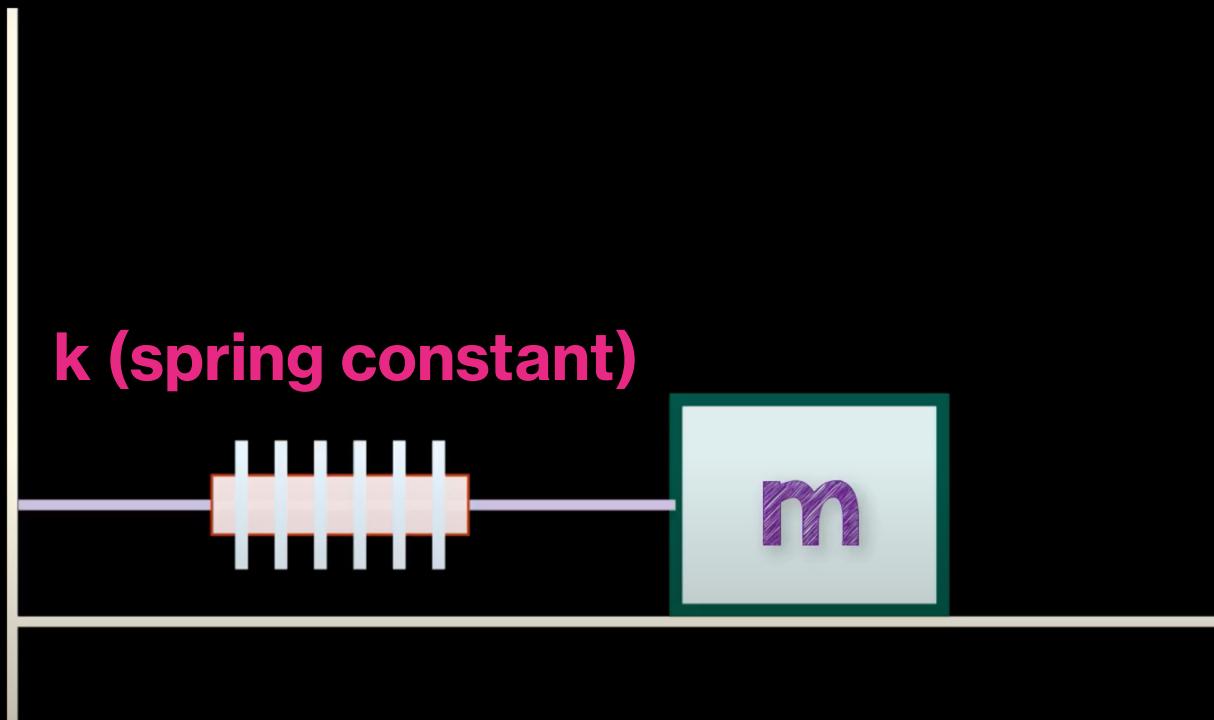


Linear Acceleration  
plus gravity

# How Accelerometers Work



What matters is the displacement



Hooke's Law

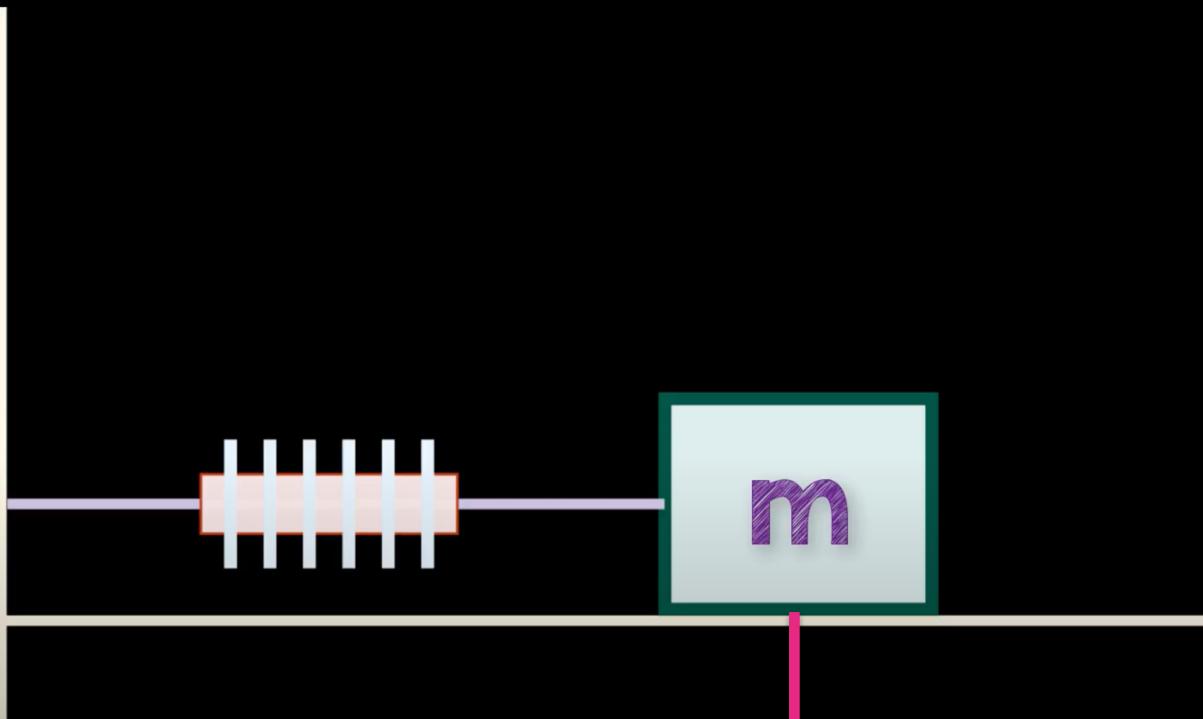
$$F = kx$$

$$\Rightarrow a = \frac{k}{m}x$$

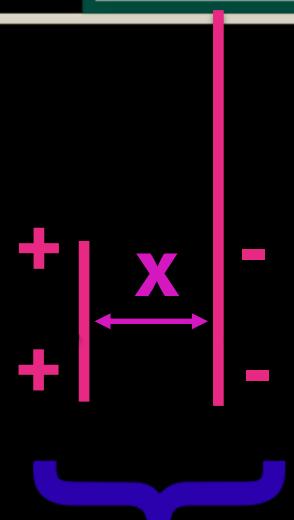
Newton's Law

$$F = ma$$

Why not simply use displacement to measure displacement?



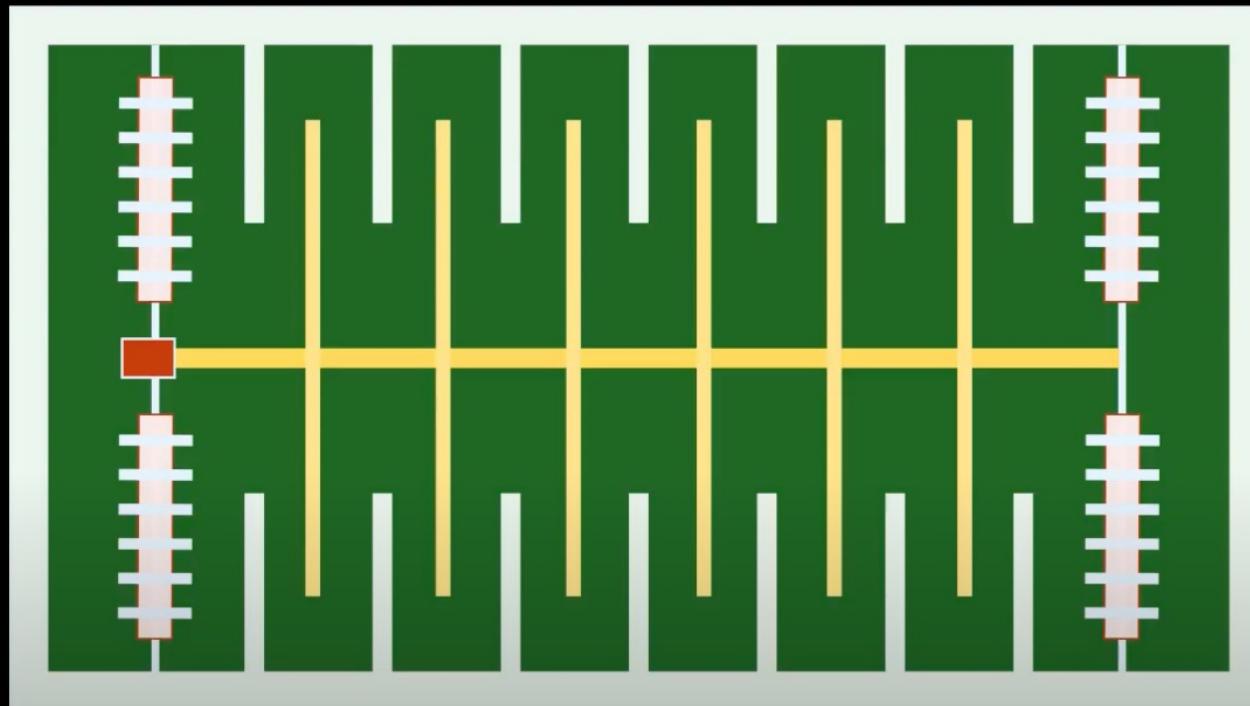
Capacitor



$$C = \epsilon \frac{\text{Area}}{x}$$

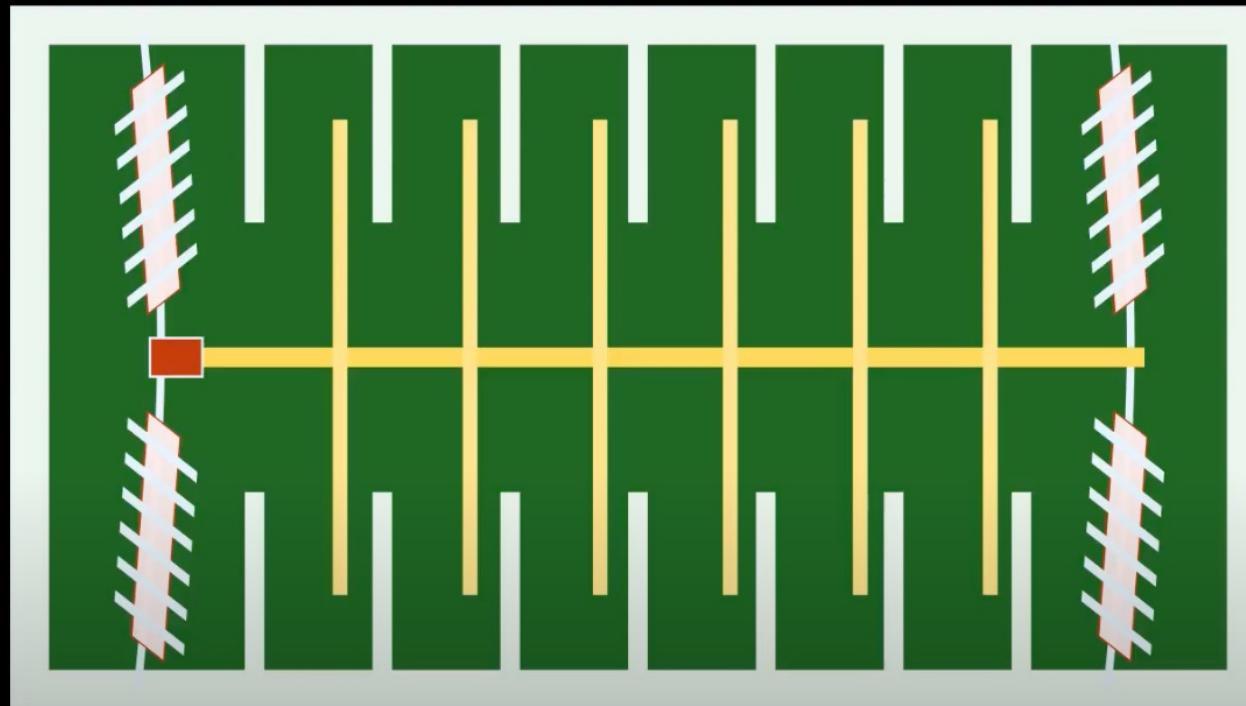
# Measuring Displacement

- How do we measure displacement?
- Most common approach is to use capacitance and MEMS (Micro electro-mechanical systems)

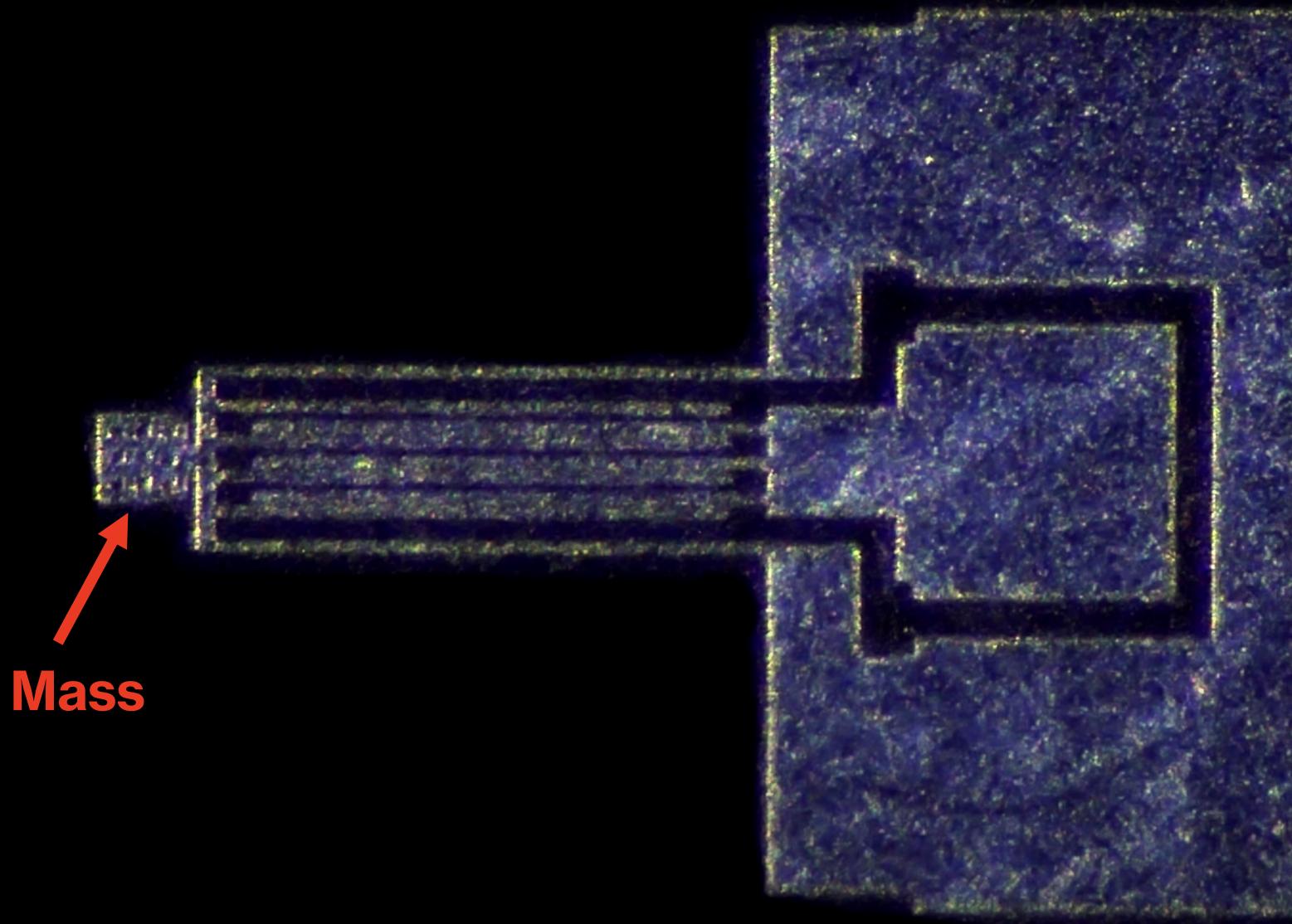


# Measuring Displacement

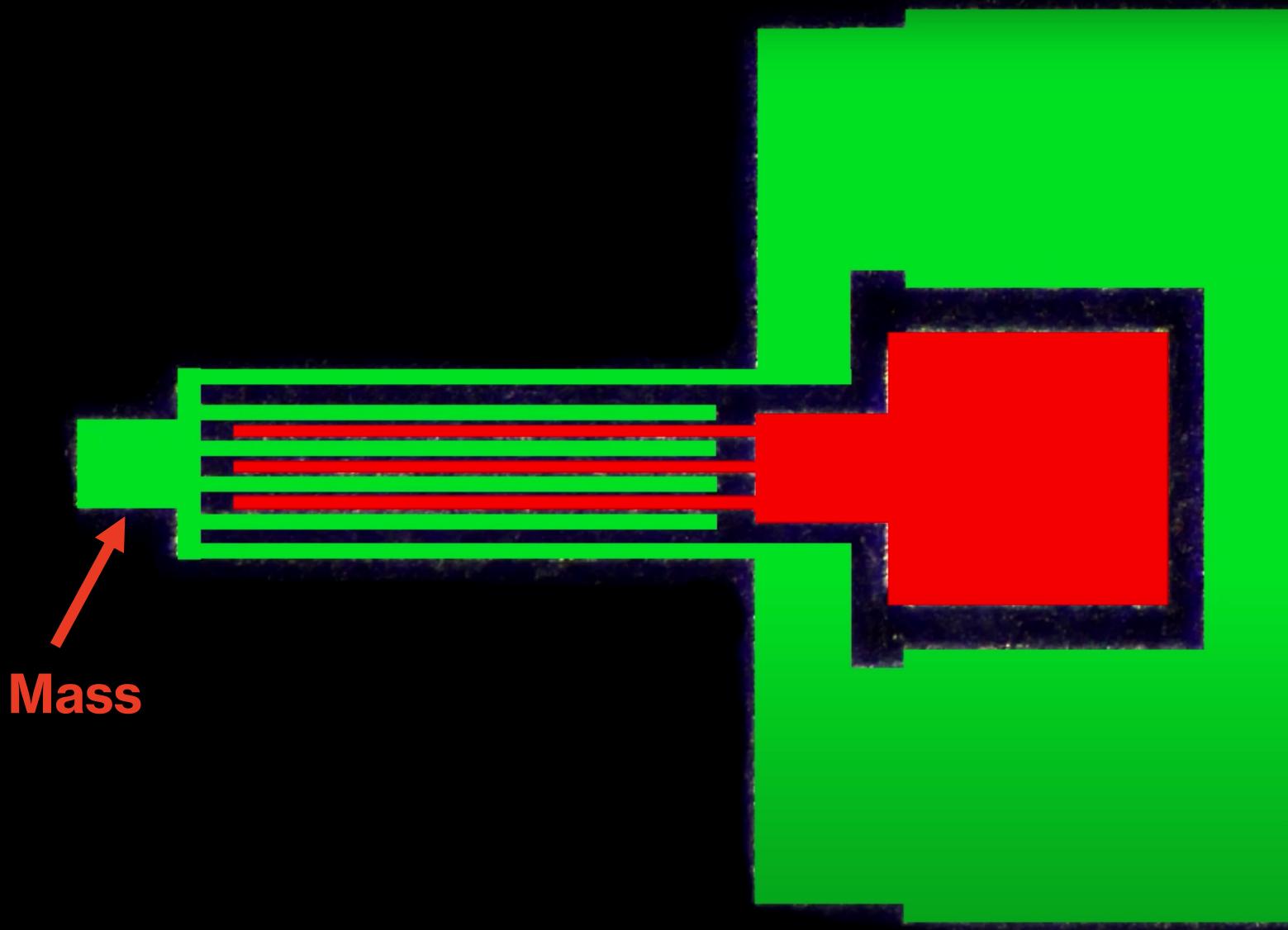
- How do we measure displacement?
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# MEMS Accelerometer

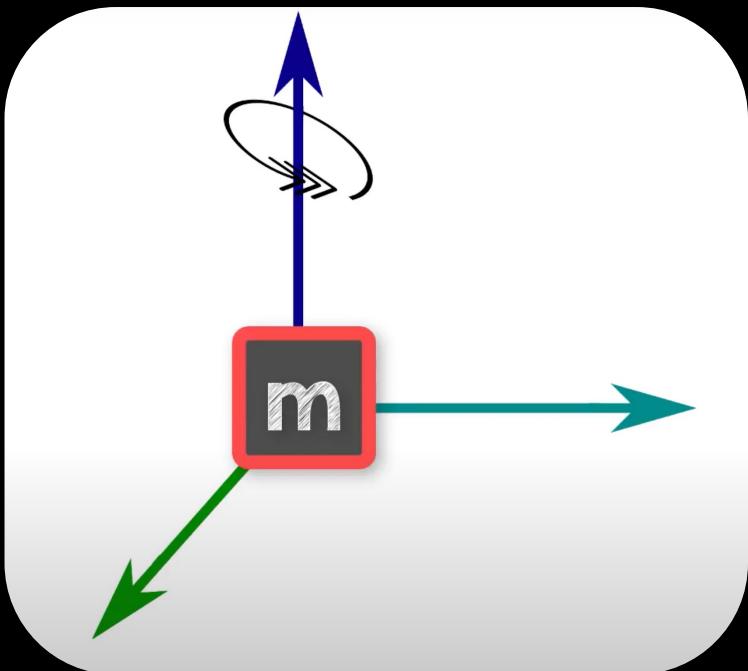


# MEMS Accelerometer



# How Gyroscopes Work?

## The Coriolis Effect

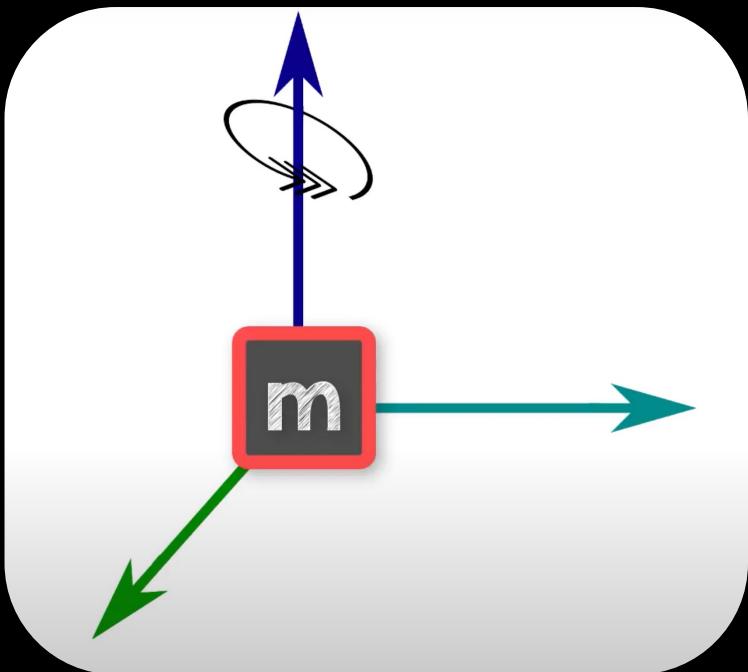


- Assume  $V_x$
- Apply  $\omega$
- Experiences a fictitious force  $F(\omega, V_x)$  following right hand rule

# The Coriolis Effect

# How Gyroscopes Work?

## The Coriolis Effect

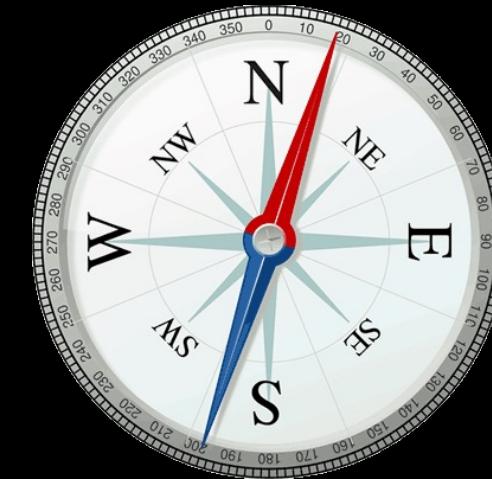
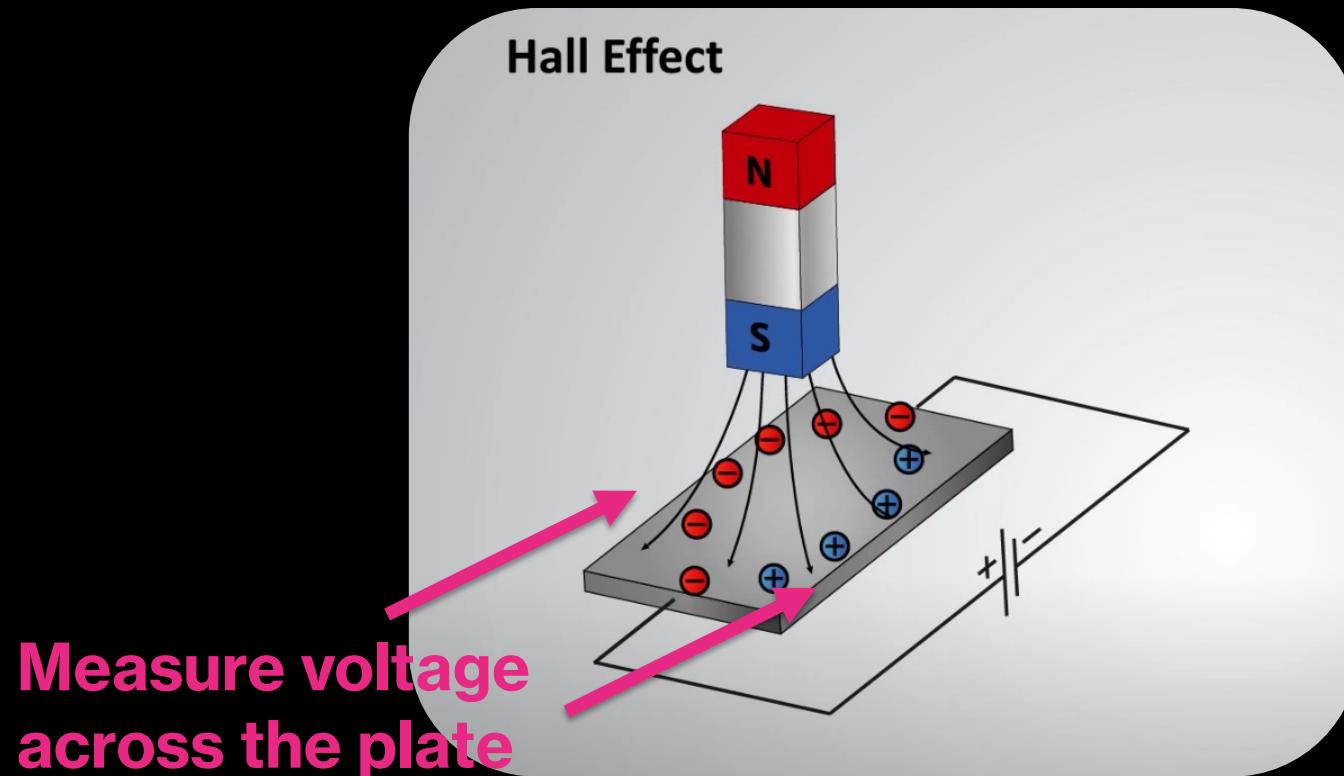


- Assume  $V_x$
- Apply  $\omega$
- Experiences a fictitious force  $F(\omega, V_x)$  following right hand rule

Can measure  $F$  in a similar fashion and use it to recover  $\omega$

# How Magnetometers Work

- E.g., Compass
- Measure Earth's magnetic field



# Rest of this Lecture

- Basic principles of operation of different IMU sensors
- **Understanding Sources of Errors**
- Dead reckoning by fusing multiple sensors
- Example system: Pothole Patrol

# Gyro Integration

Angle (degrees)

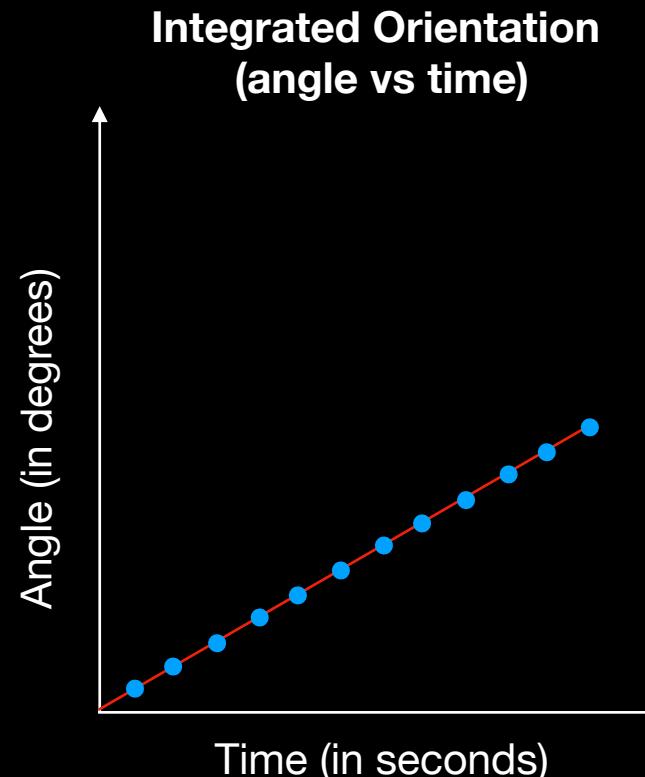
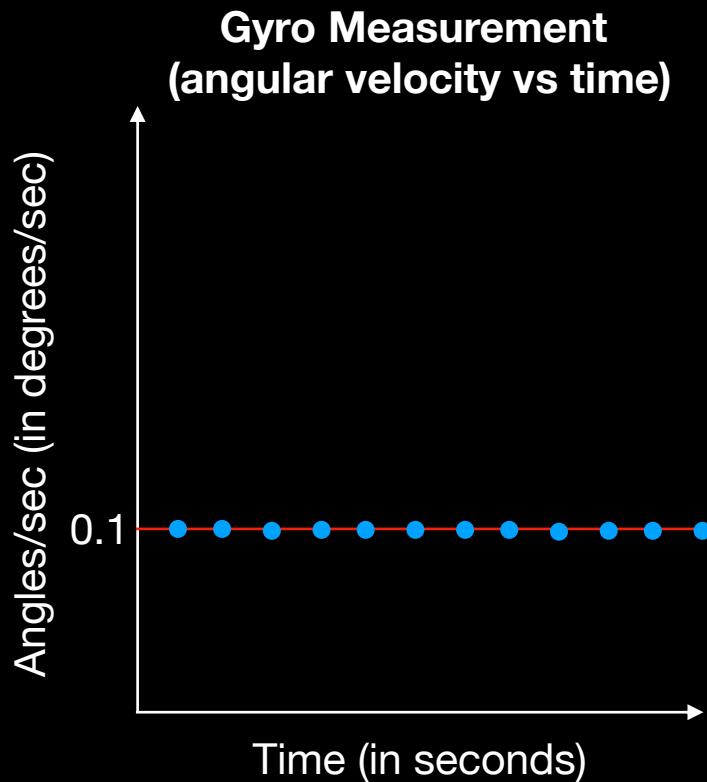


- Let's plot this for gyro measurement and for orientation
- Let's include ground truth and measured data for each

Consider:

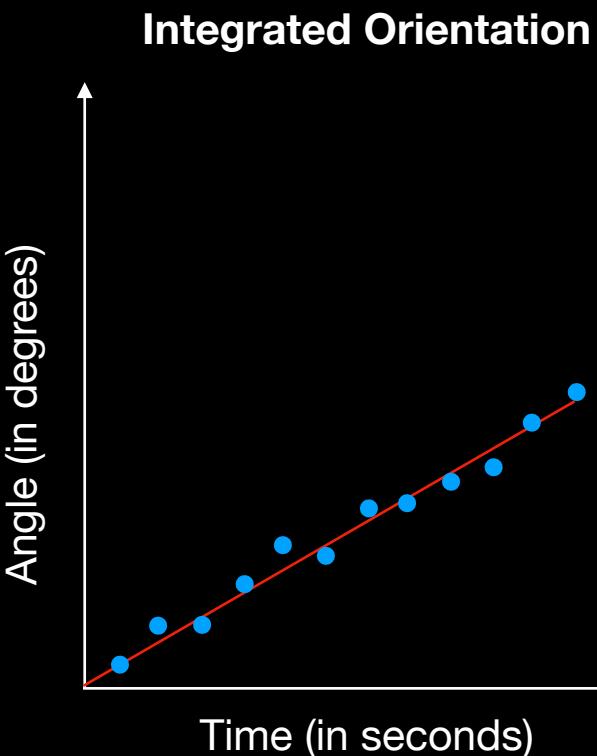
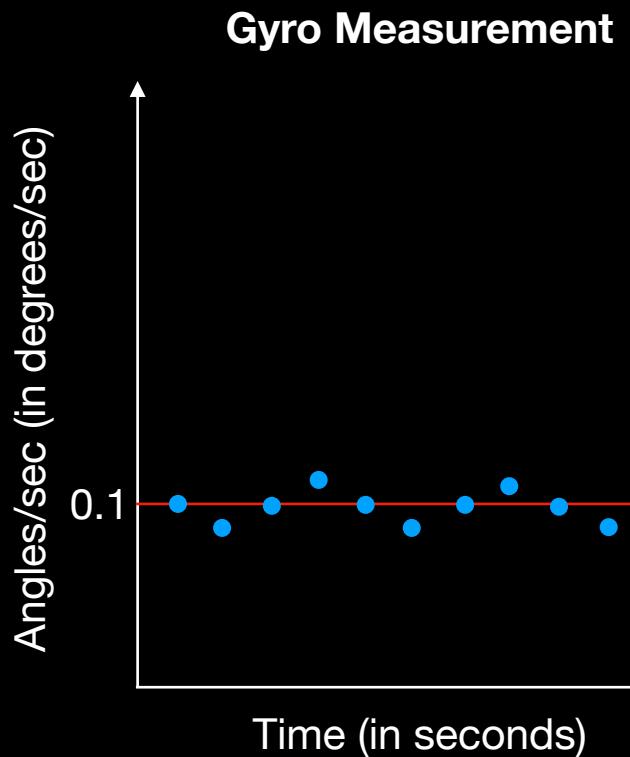
- linear (angular) motion, no noise, no bias
- linear (angular) motion, with noise, no bias
- linear (angular) motion, no noise, bias
- nonlinear motion, no noise, no bias

# Gyro integration: linear motion, no noise, no bias

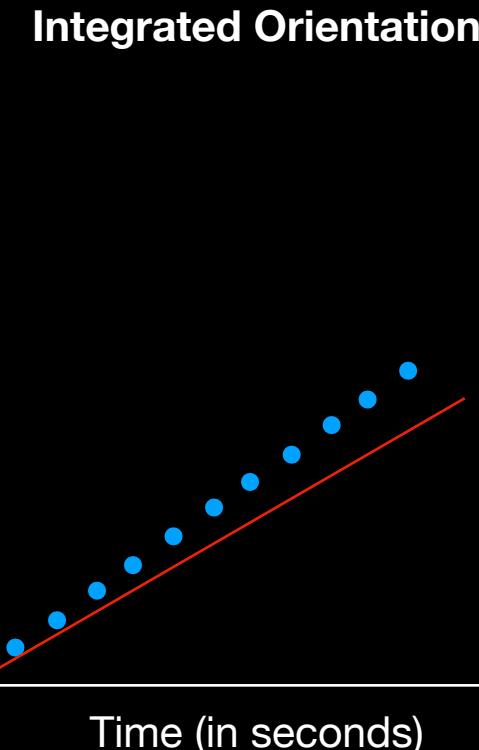
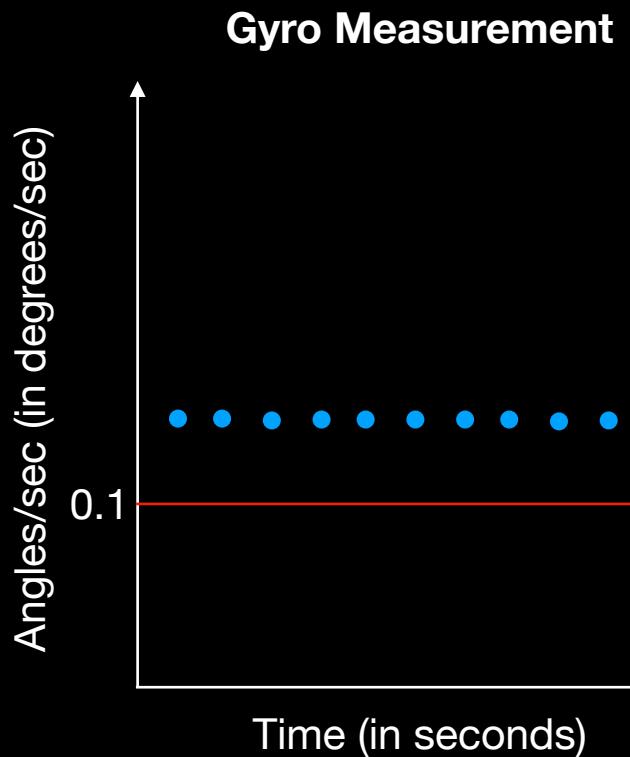


- Ground truth
- Measured/estimated angle

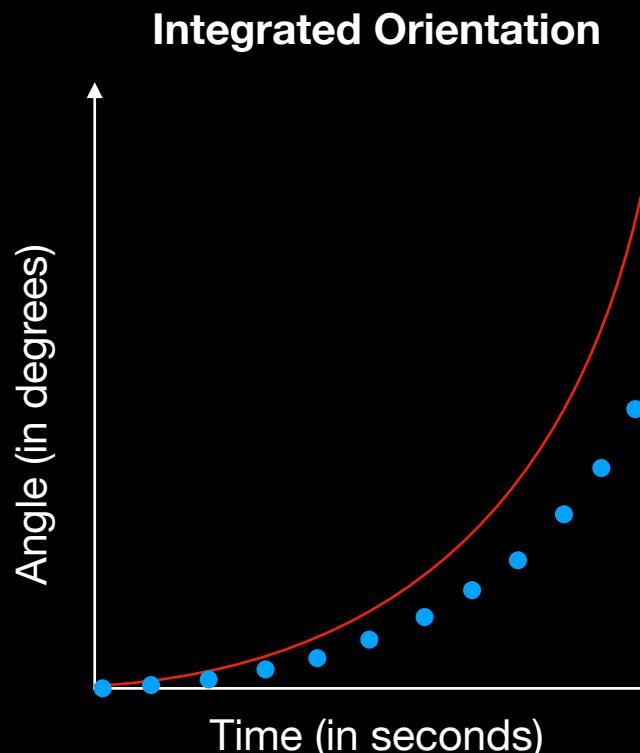
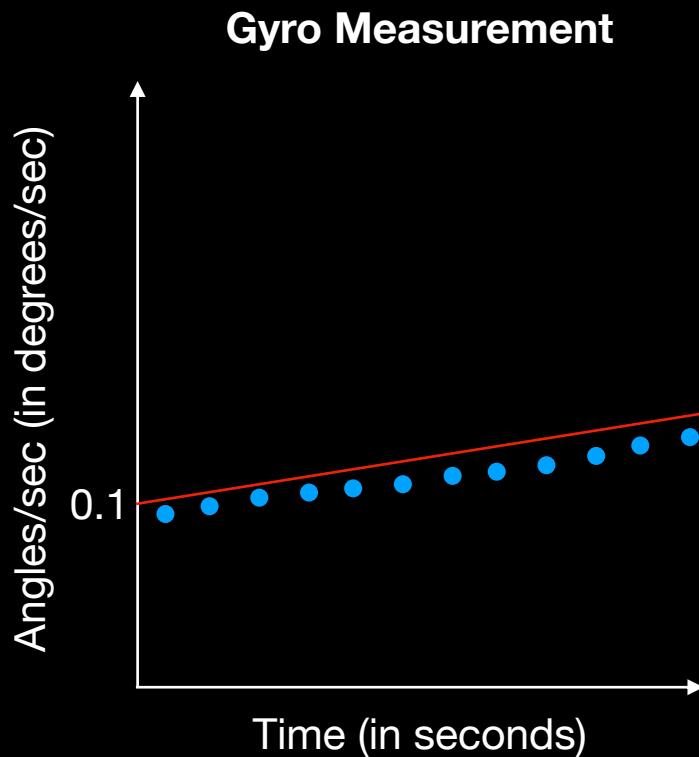
# Gyro integration: linear motion, noise, no bias



# Gyro integration: linear motion, no noise, bias



# Gyro integration: nonlinear motion, no noise, no bias



# Gyro Integration aka *Dead Reckoning*

- Works well for linear motion, no noise, no bias = unrealistic
- If bias is unknown and noise is zero -> drift (from integration)
- Bias and noise variance can be estimated, other sensor measurements used to correct for drift (sensor fusion)
- Accurate in short term, but not reliable in long term due to drift

# Rest of this Lecture

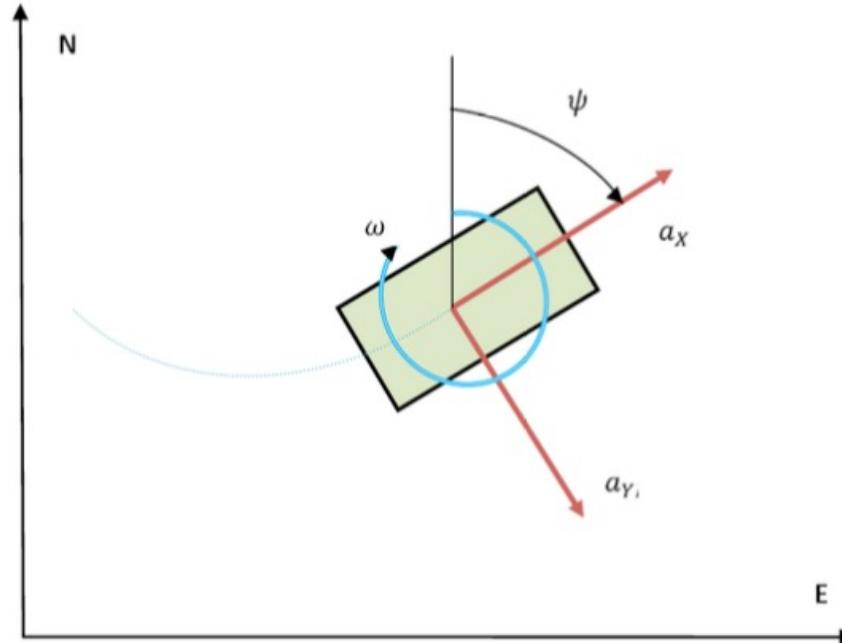
- Basic principles of operation of different IMU sensors
- Understanding Sources of Errors
- **Dead reckoning by fusing multiple sensors**
- Example system: Pothole Patrol



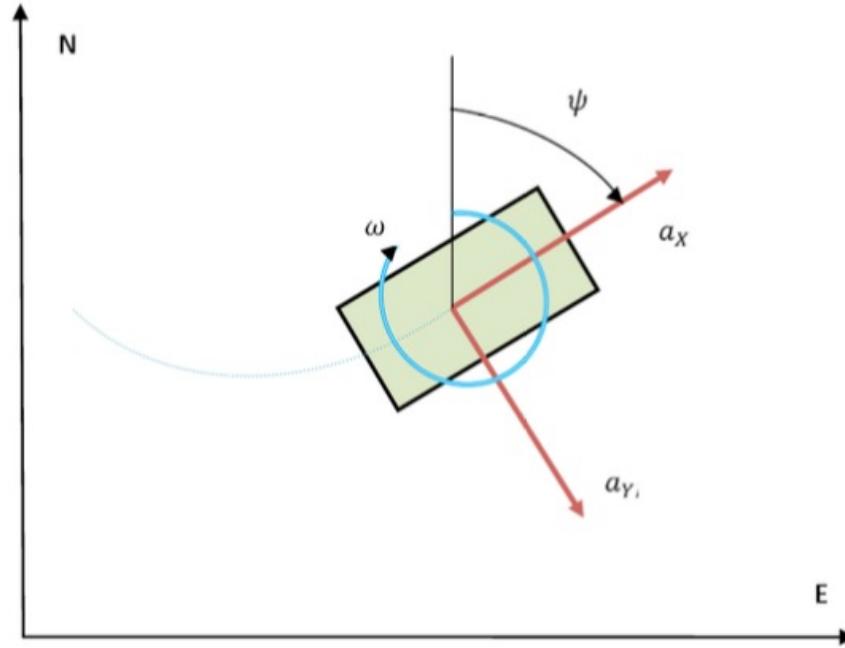
# Dead Reckoning

- The process of calculating one's current position by using a previously determined position, and advancing that position based upon known or estimated speeds over elapsed time and course
- Key things to keep in mind:
  - Frames of reference
  - Orientation change

# 2D Inertial Navigation in Strapdown System



# 2D Inertial Navigation in Strapdown System



$$\begin{bmatrix} a_N \\ a_E \end{bmatrix} = \begin{bmatrix} \cos \psi & -\sin \psi \\ \sin \psi & \cos \psi \end{bmatrix} \begin{bmatrix} a_x \\ a_y \end{bmatrix}$$

$$a_N(t) = \cos \psi \cdot a_x(t) - \sin \psi \cdot a_y(t)$$
$$a_E(t) = \sin \psi \cdot a_x(t) + \cos \psi \cdot a_y(t)$$

# 2D Inertial Navigation in Strapdown System

**Acceleration:**  $a_N(t) = \cos \psi \cdot a_x(t) - \sin \psi \cdot a_y(t)$

$$a_E(t) = \sin \psi \cdot a_x(t) + \cos \psi \cdot a_y(t)$$

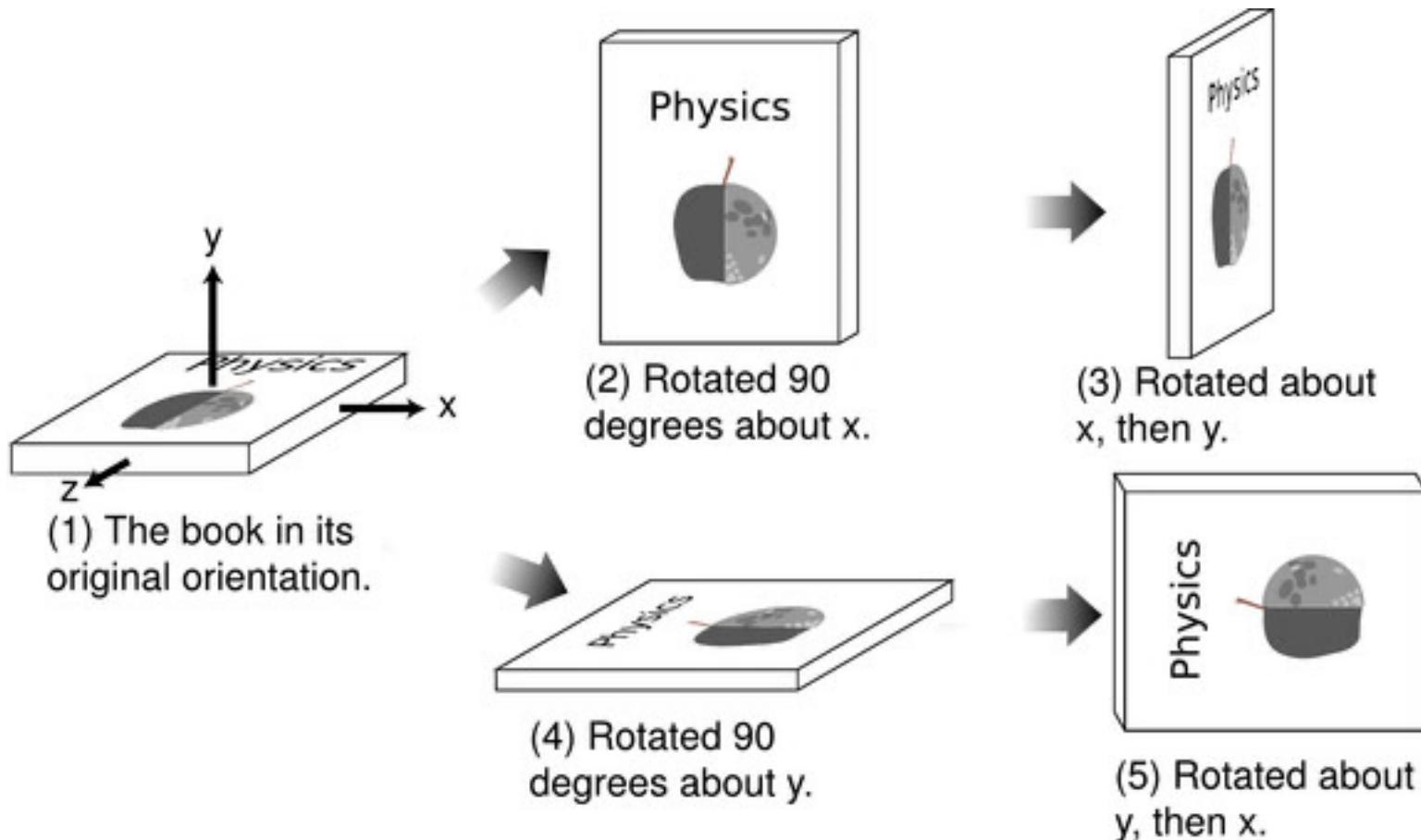
**Velocity:**  $V_N(t) = V_N(t_0) + \int_{t_0}^t a_N(\tau) d\tau$

$$V_E(t) = V_E(t_0) + \int_{t_0}^t a_E(\tau) d\tau$$

**Position:**  $X_N(t) = X_N(t_0) + \int_{t_0}^t V_N(\tau) d\tau$

$$X_E(t) = X_E(t_0) + \int_{t_0}^t V_E(\tau) d\tau$$

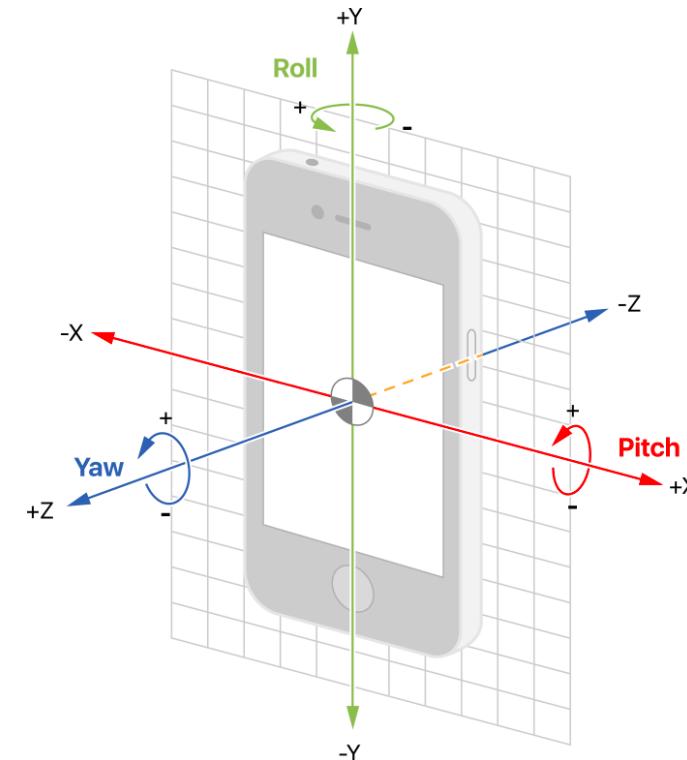
# How about 3D Rotations?



Non-commutative = order matters!

# 3D Rotation Representations

- Rotation Matrix
  - + 3 orthonormal vectors = 9 numbers
- Euler Angles (roll, pitch, yaw)
  - + Symmetry problem, Gimbal lock
- Axis-angle
- Quaternions



# Quaternions

- 4-dimensional number

Complex number	Quaternion	$i^2 = \cancel{j^2} = \cancel{k^2} = -1$
$3.14 + 1.59i$	$0.00 + \underbrace{8.46i + 2.64j + 3.38k}_{\substack{\text{Scalar} \\ \text{part}}} + \underbrace{\text{“Vector”}}_{\text{part}}$	$\cancel{ij} = -\cancel{ji} = k$
		$ki = -ik = j$
		$jk = -kj = i$

- *Unit* quaternions represent 3D rotations

$$q = \cos\left(\frac{\theta}{2}\right) + \sin\left(\frac{\theta}{2}\right) (u_x i + u_y j + u_z k)$$

$$\cancel{p} \rightarrow q \cdot \cancel{p} \cdot q^{-1}$$

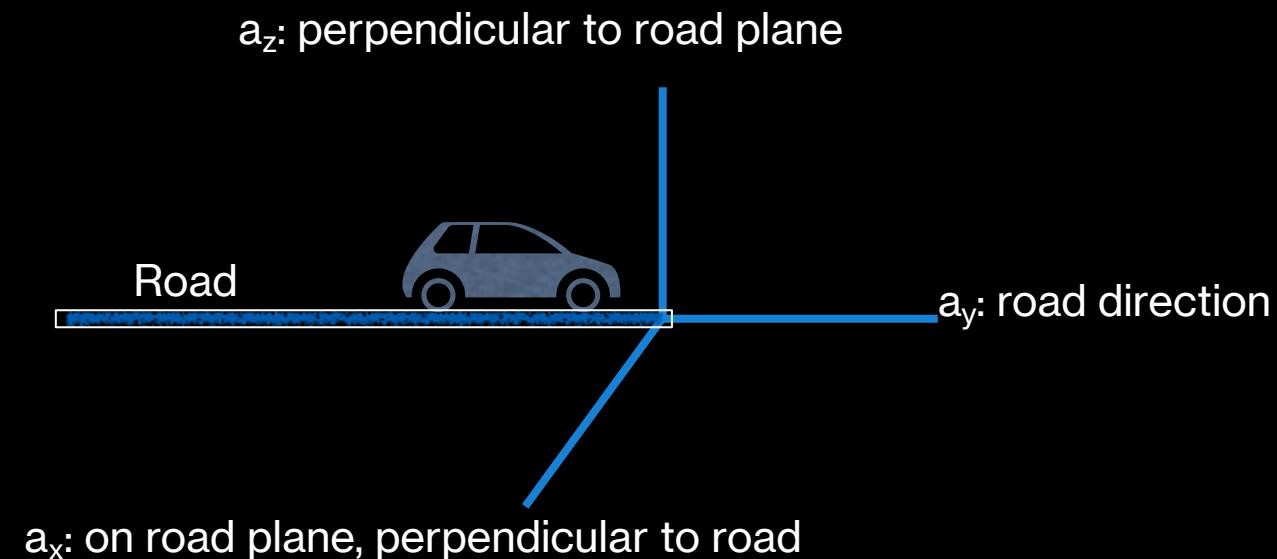
# Rest of this Lecture

- Basic principles of operation of different IMU sensors
- Understanding Sources of Errors
- Dead reckoning by fusing multiple sensors
- **Example system: Pothole Patrol**

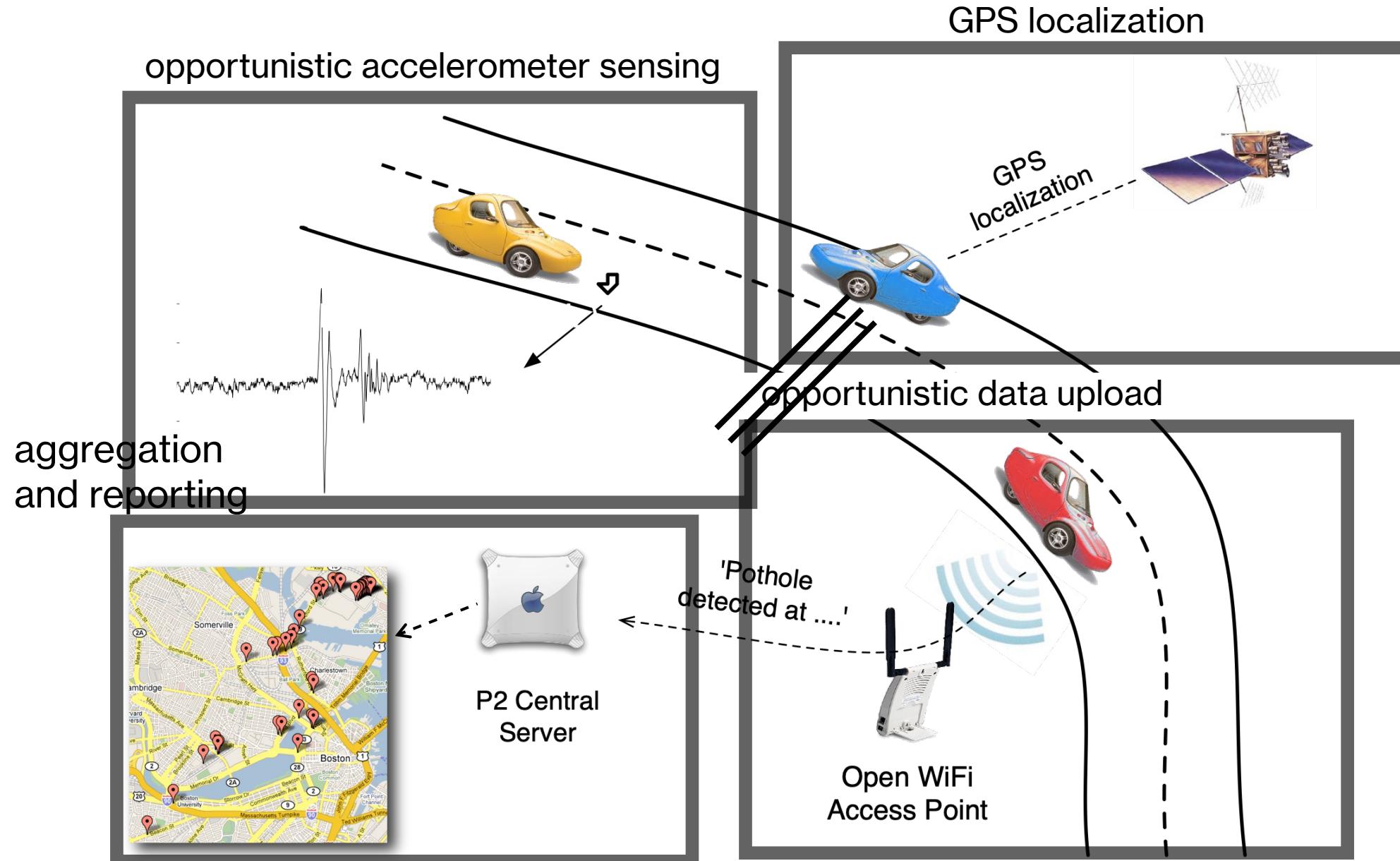


- road decay unavoidable, hard to predict
- current monitoring methods costly/ineffective

# Acceleration vector



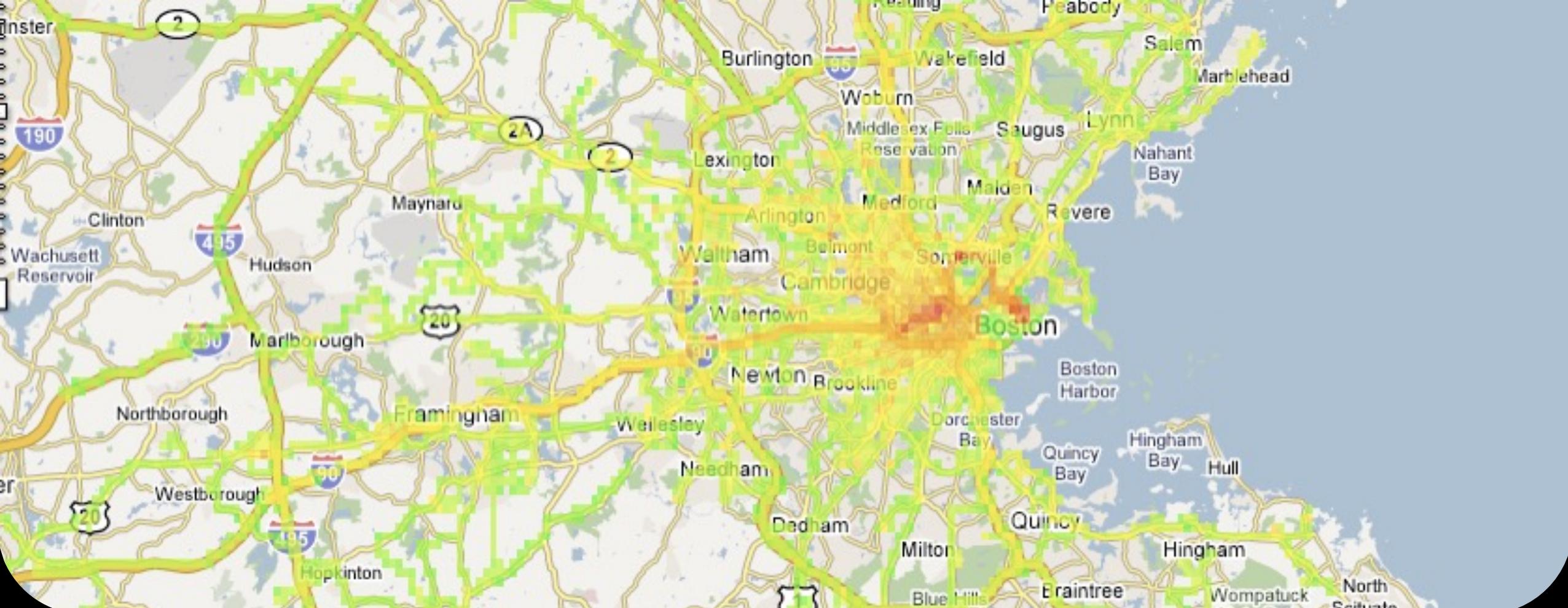
# Pothole Patrol System Overview



# Experimental Platform

- 7 Boston/Cambridge taxis
- small computer in glove box
- 380 Hz 3-axis accelerometer
- 802.11a/b/g wireless interface
- GPS receiver on roof
- <time,location,heading,speed,ax,ay,az>





# Wide-area Sensing & Crowdsensing

# Sensor Placement

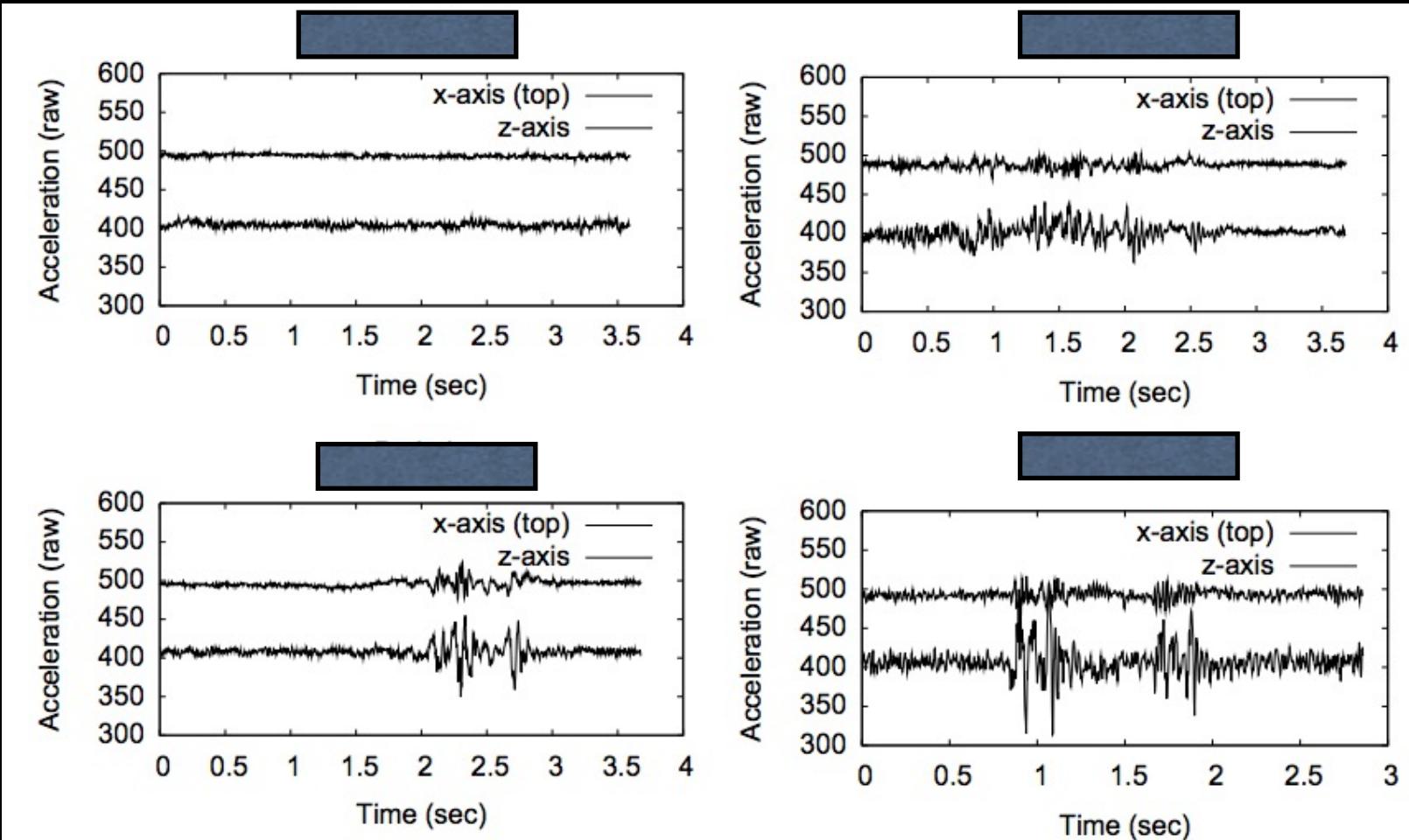


Pros? Cons?

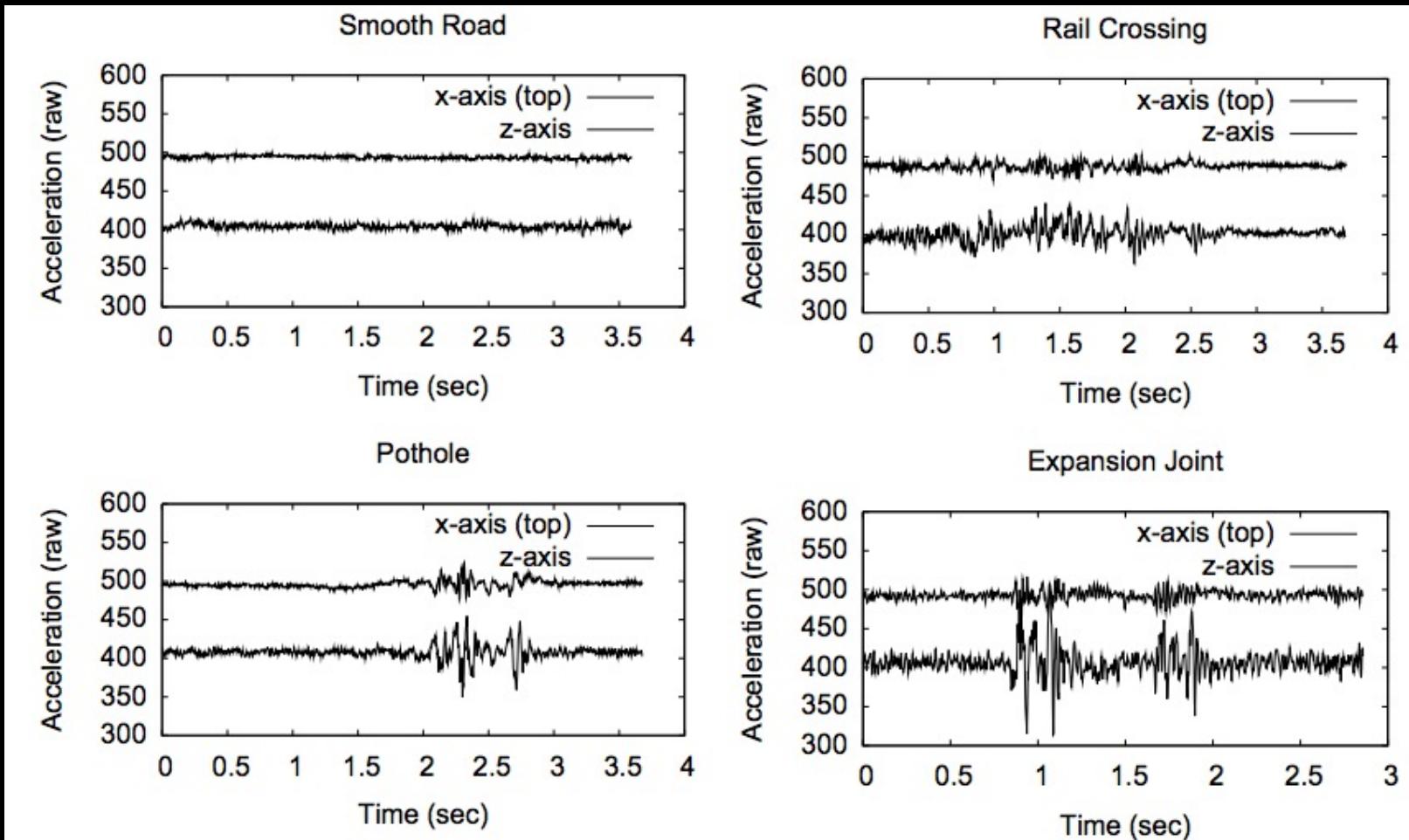
# Pothole & False Alarms



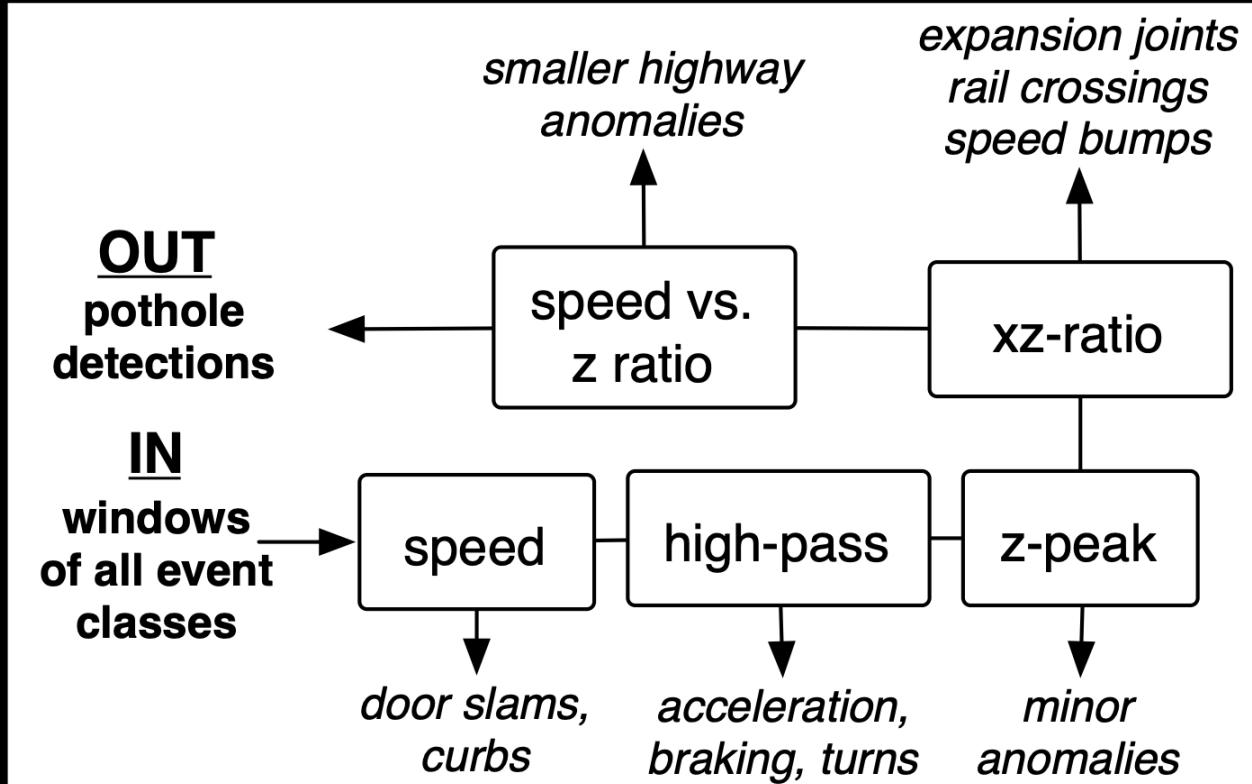
# Challenge: Pothole vs “Not Pothole”



# Challenge: Pothole vs “Not Pothole”



# P<sup>2</sup> detector



Events usually of much shorter duration than 256 samples

Need to learn threshold parameters

# iOS Lab 1 is out

- **Topic:** Develop a location app and explore the power drain vs accuracy trade-off
- **Due:** Mon Feb 19<sup>th</sup>, 11:59 pm

# Next Lecture

- **Time:** Wed Feb 14<sup>th</sup>
- **Topic:** Applied ML for Mobile and IoT Sensing