



6.808: Mobile and Sensor Computing aka IoT Systems

<http://6808.github.io>

Lecture 7: Introduction to Inertial Sensing & Sensor Fusion

Some material adapted from Gordon Wetzstein (Stanford) and Sam Madden (MIT)

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Announcements

- 1- PSet 1 due Feb 28
- 2- Lab 2 due March 2

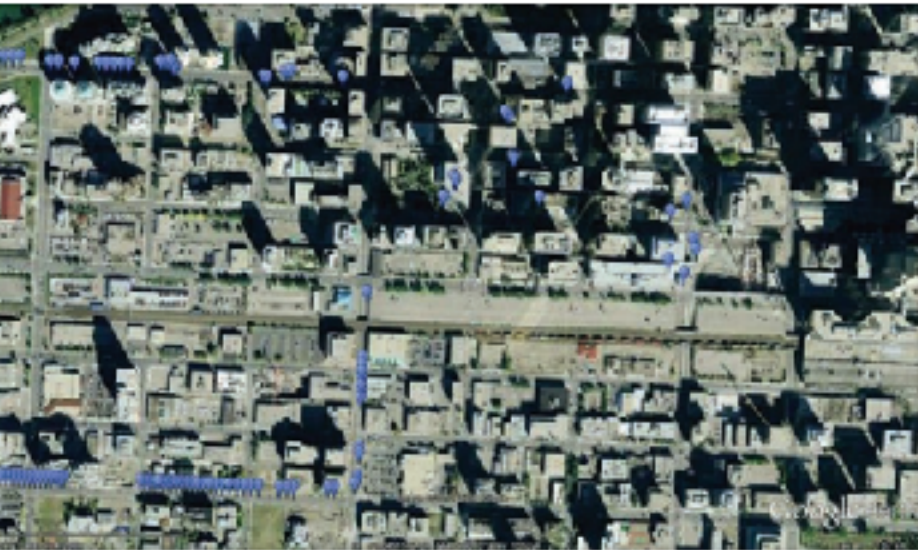
Overview: the next two lectures

A deep-dive on **IoT sensing modalities**

this lecture

1. What sensing modalities are relevant for location inference?
2. How can we leverage physics and mathematics fundamentals to build reliable, microscopic sensors?
3. Case-study based application of inertial sensing:
Pothole patrol
4. Practical approaches to accounting for sensory noise in real-world settings.

Example Application: Inertial Navigation



GPS only

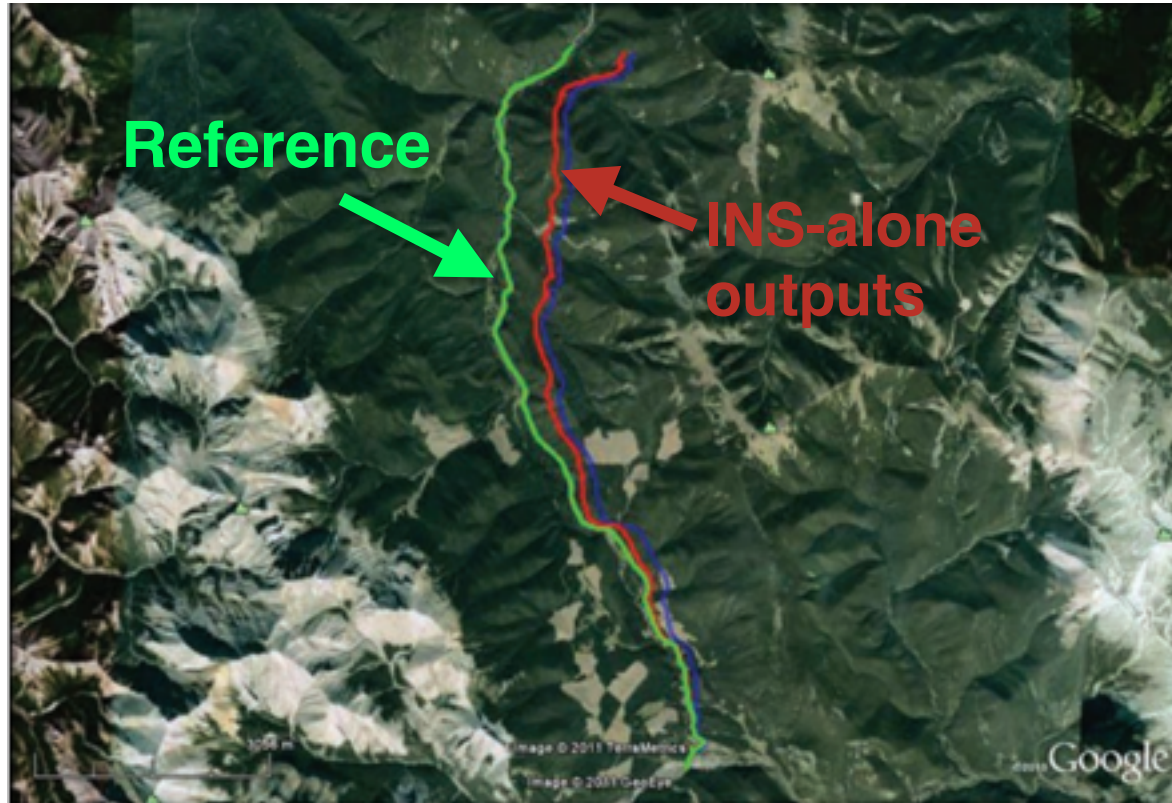


GPS+INS

Key Idea #1: Integrate acceleration data over time to discover location (Inertial Sensing)

Inertial Sensing alone is not enough for accurate positioning

- Errors accumulate over time



Source: INS Face Off
MEMS versus FOGs

Key Idea #2: Fuse Data from Multiple Sensors
(Sensor Fusion)

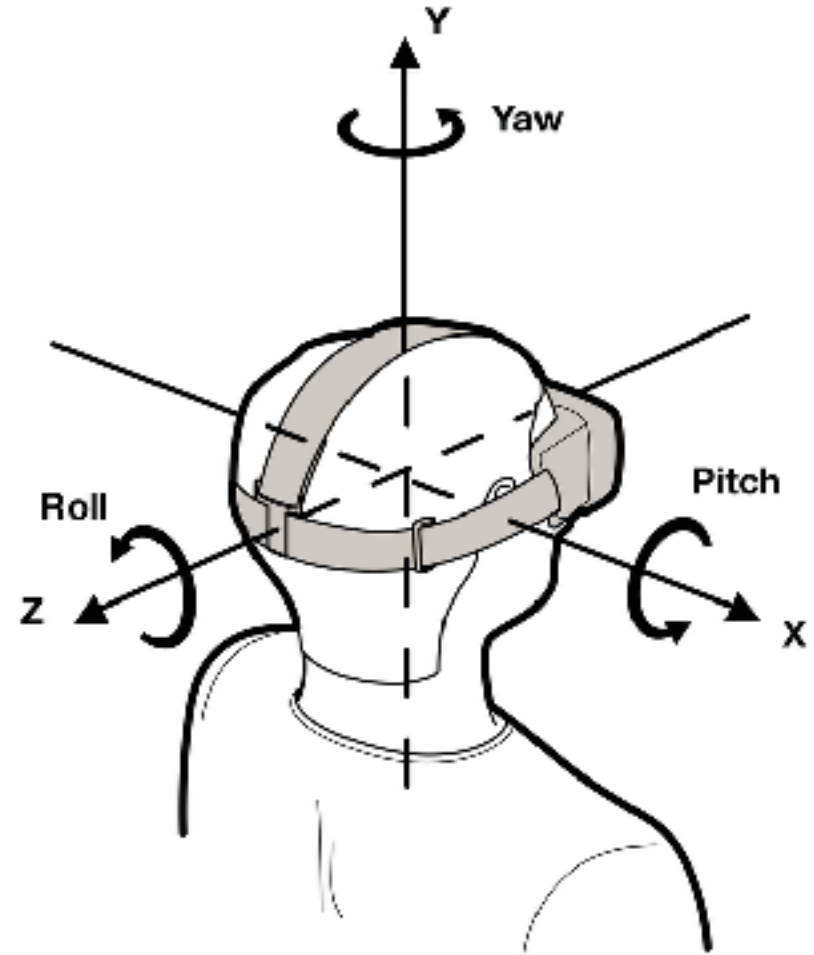
This Lecture

Key Idea #1: Integrate acceleration data over time to discover location (Inertial Sensing)

Key Idea #2: Fuse Data from Multiple Sensors (Sensor Fusion)

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)

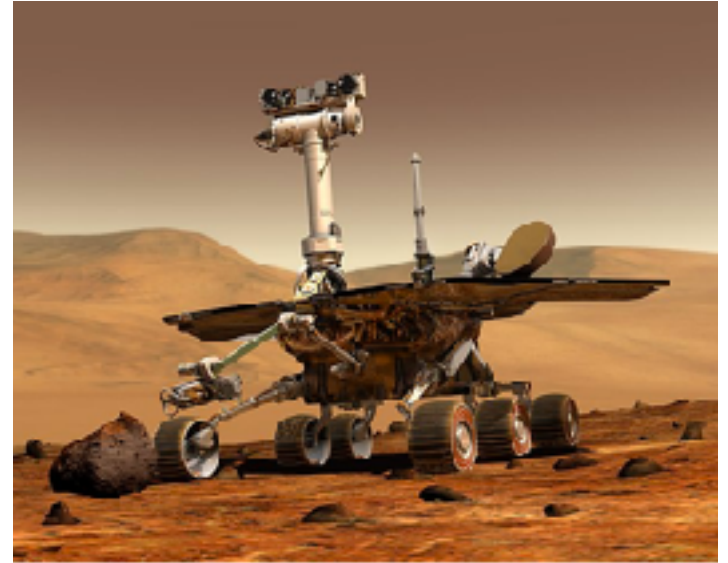


What does an IMU consist of? (Inertial Measurement Unit)

- **Gyroscope** measures angular velocity ω in degrees/s
- **Accelerometer** measures linear acceleration \mathbf{a} in m/s^2
- **Magnetometer** measures magnetic field strength \mathbf{m} in μT (micro-Teslas).

Why is it called IMU?

Where are IMUs used today?

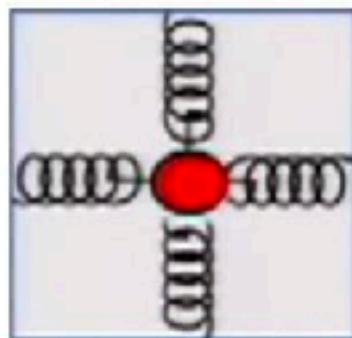


Rest of this Lecture

- Basic principles of operation of different IMU sensors: accelerometer, gyroscope, magnetometer
- Understanding Sources of Errors
- Dead reckoning by fusing multiple sensors

Accelerometer

Mass on spring



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

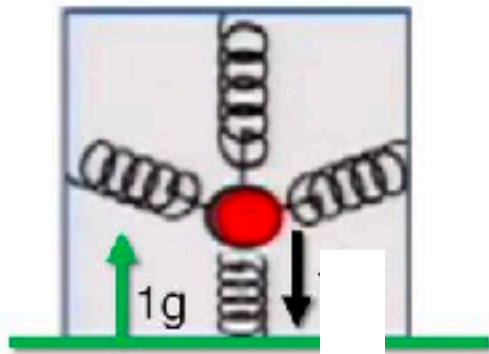
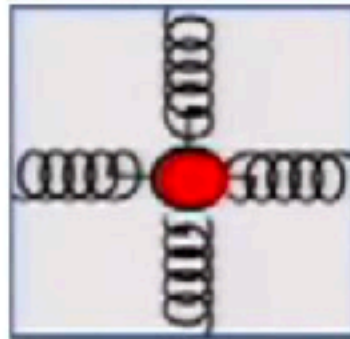
Free Fall

Linear Acceleration

Linear
Acceleration
plus gravity

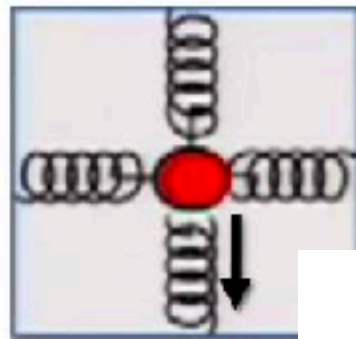
Accelerometer

Mass on spring

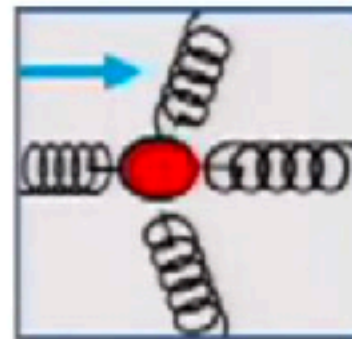


Gravity

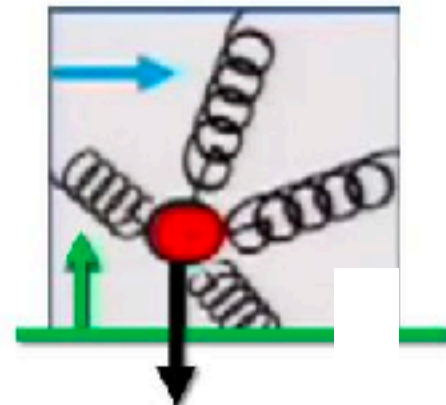
$$1g = 9.8\text{m/s}^2$$



Free Fall

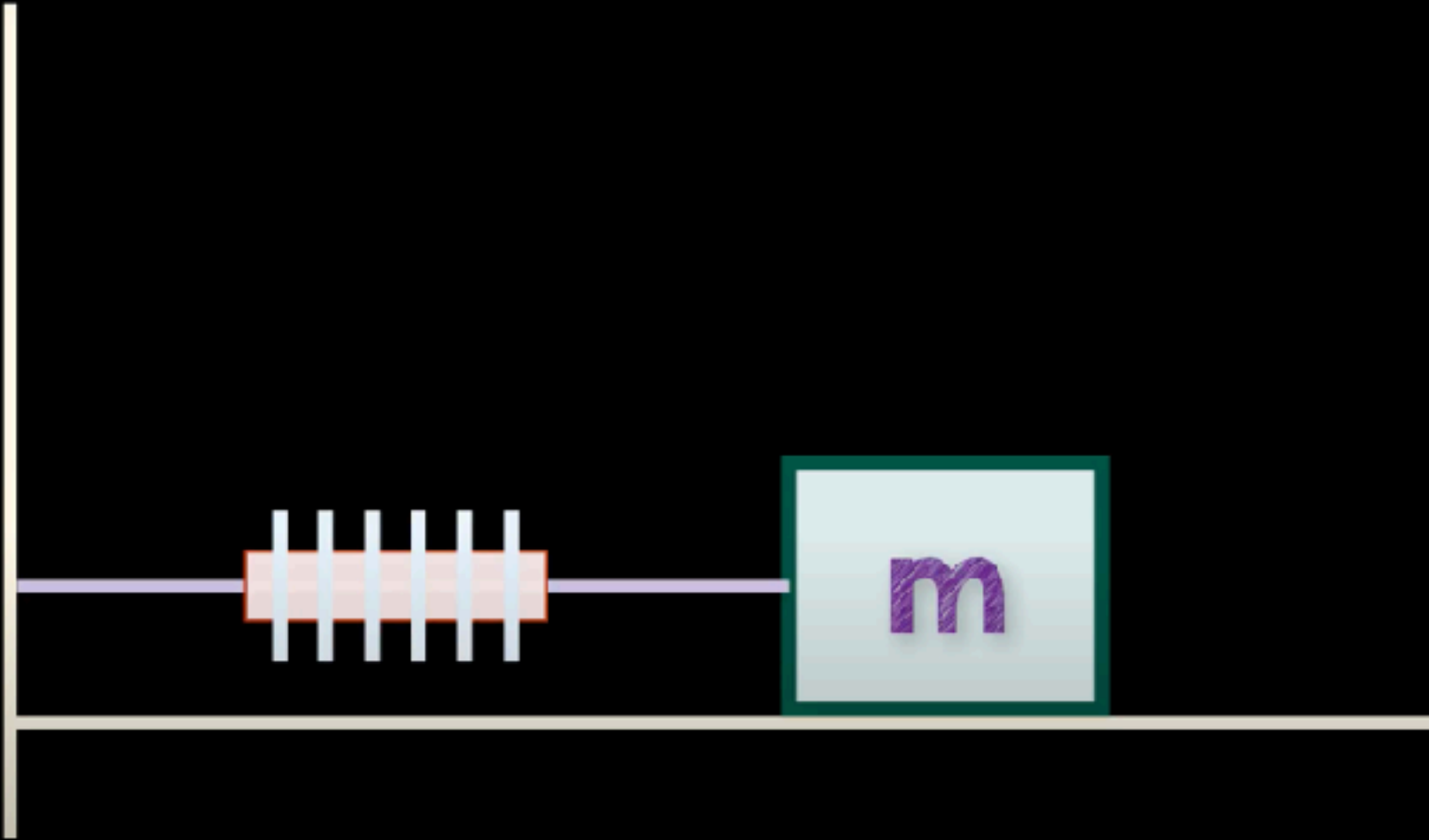


Linear Acceleration

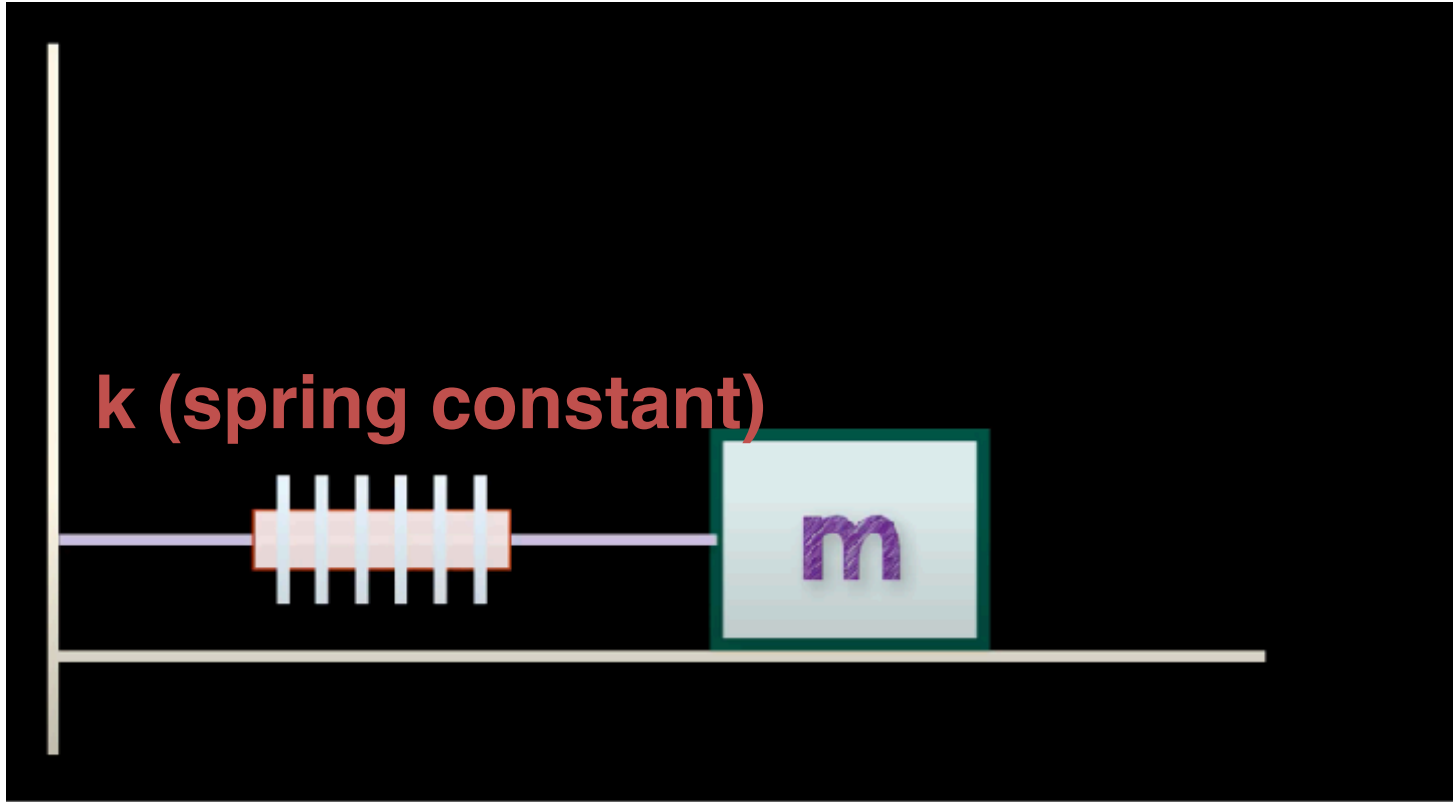


Linear
Acceleration
plus gravity

How Accelerometers Work



What matters is the displacement



Hooke's Law

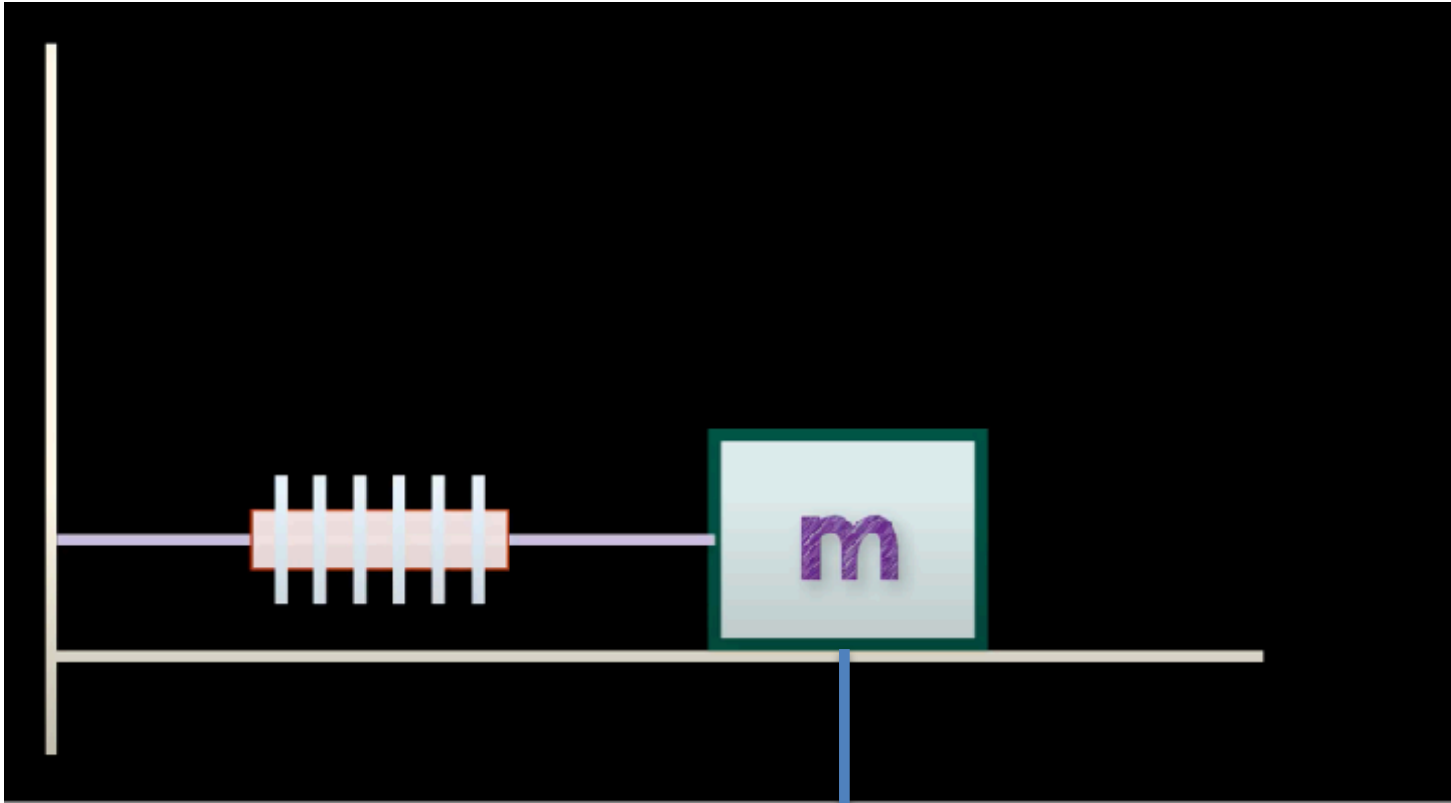
$$F = kx$$

Newton's Law

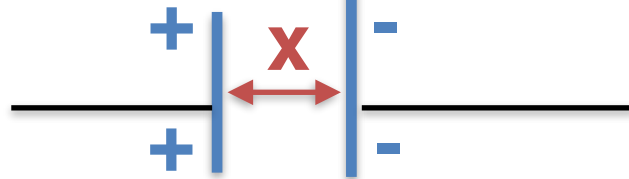
$$F = ma$$

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

Why not simply use displacement to measure displacement?



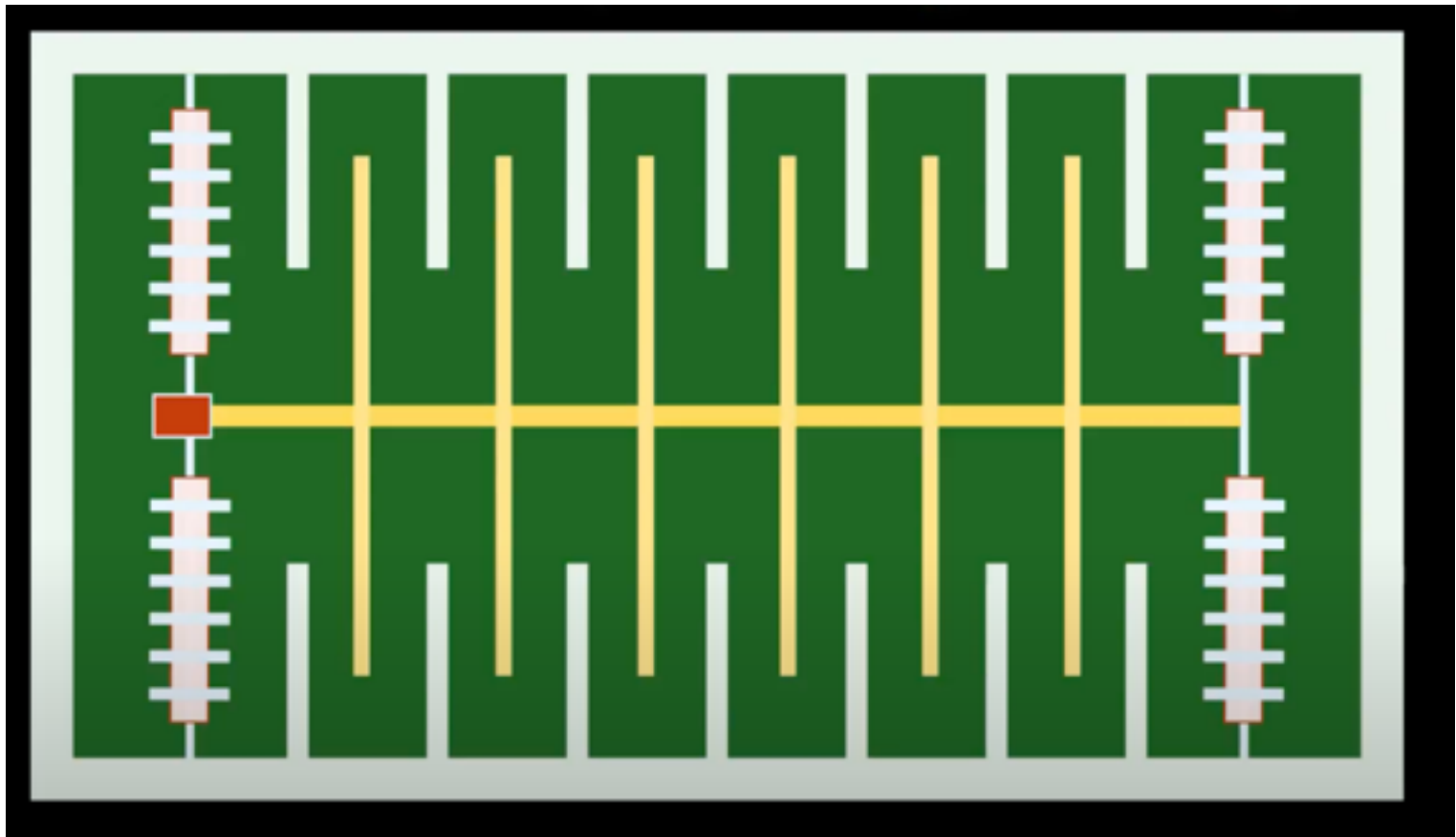
Capacitor



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

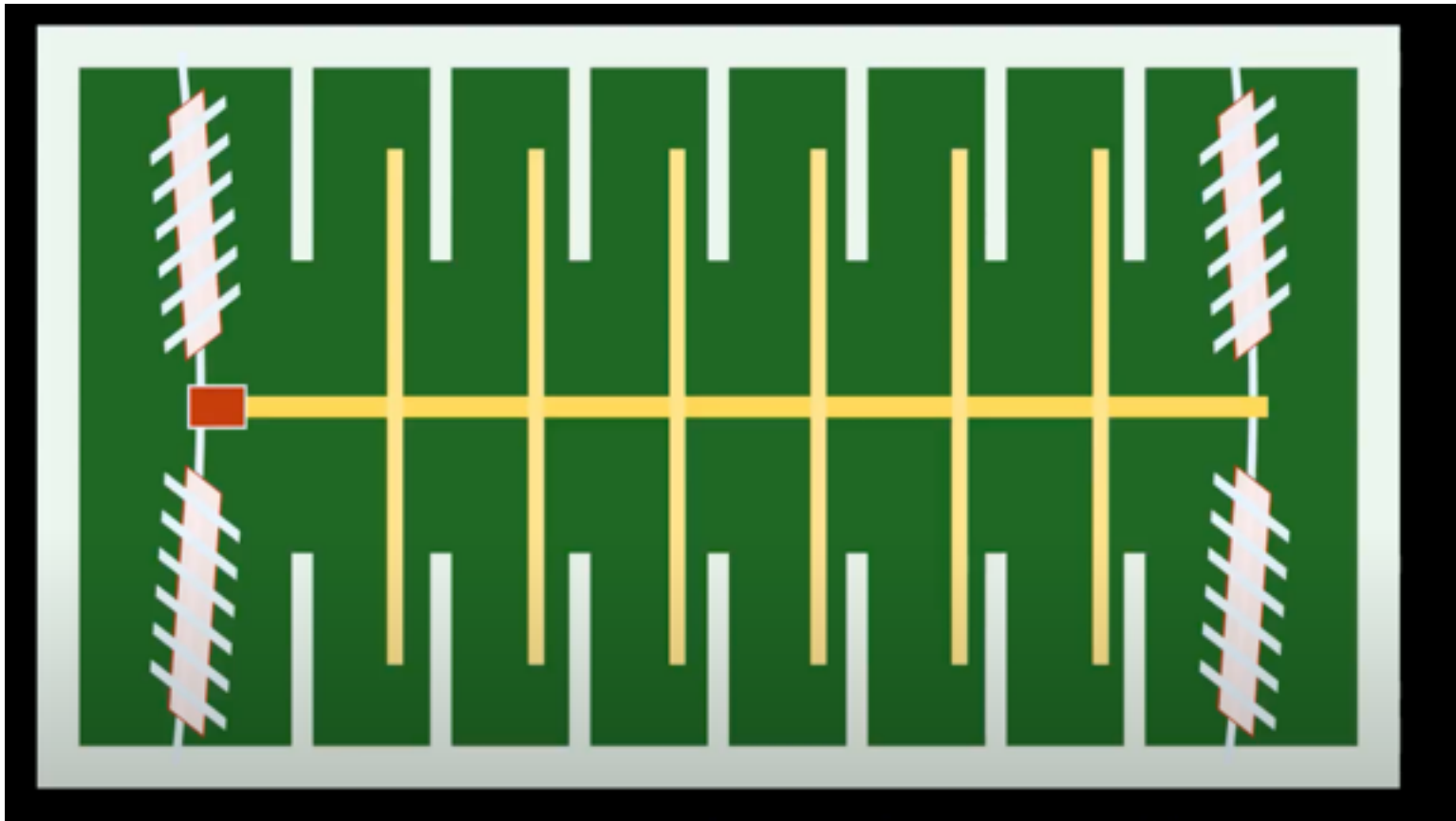
Measuring Displacement

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

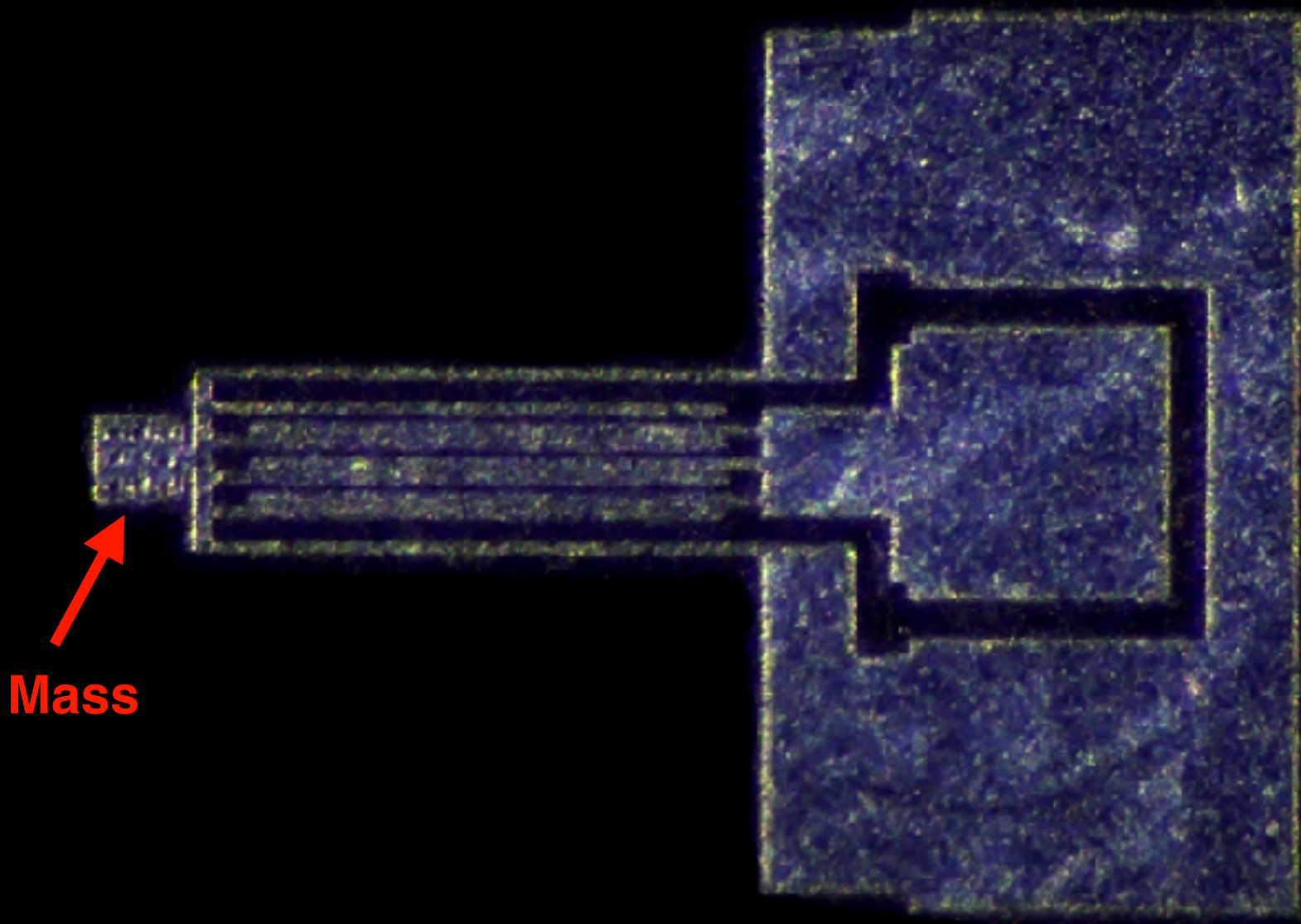


Measuring Displacement

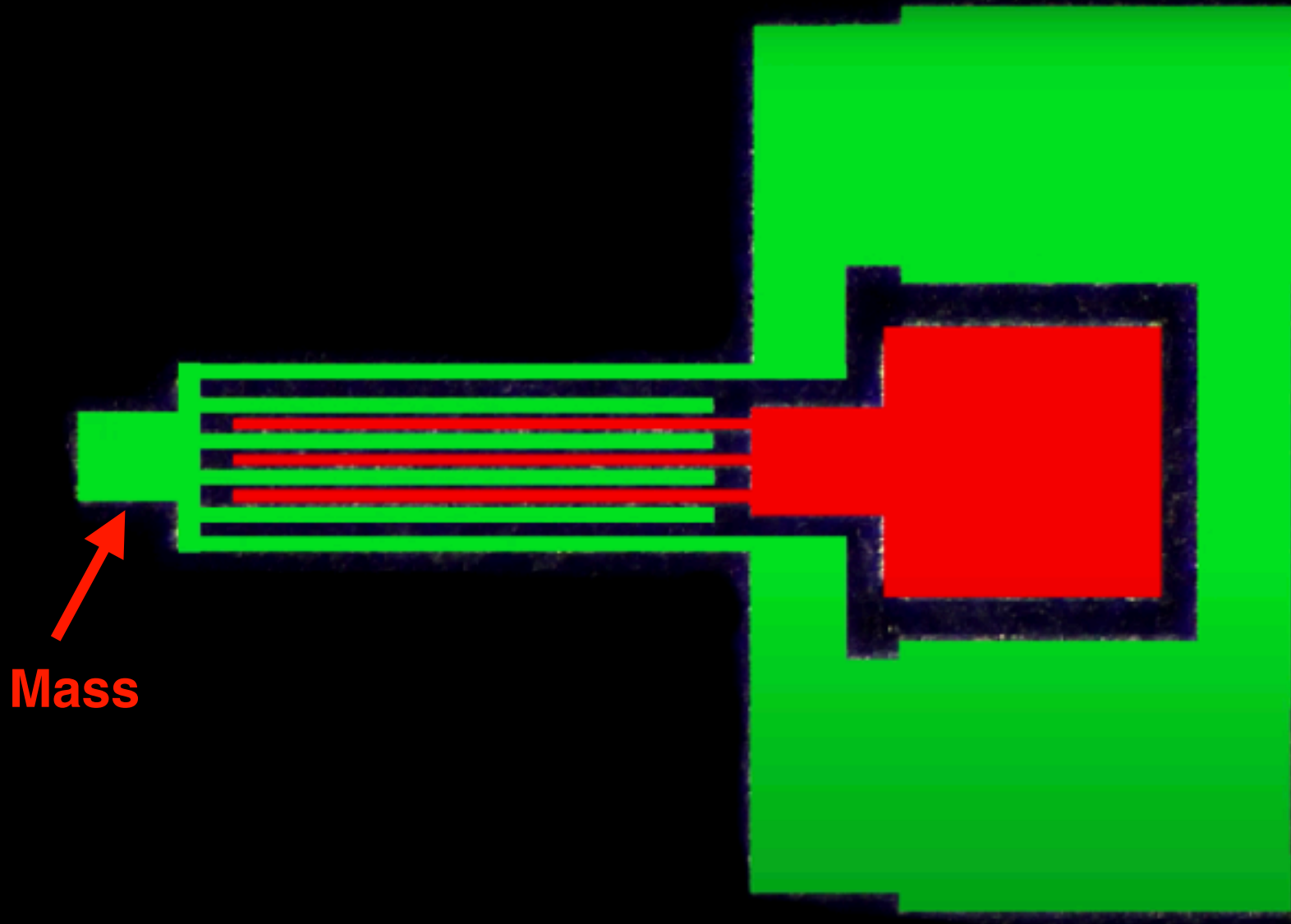
- How do we measure displacement?
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MEMS Accelerometer

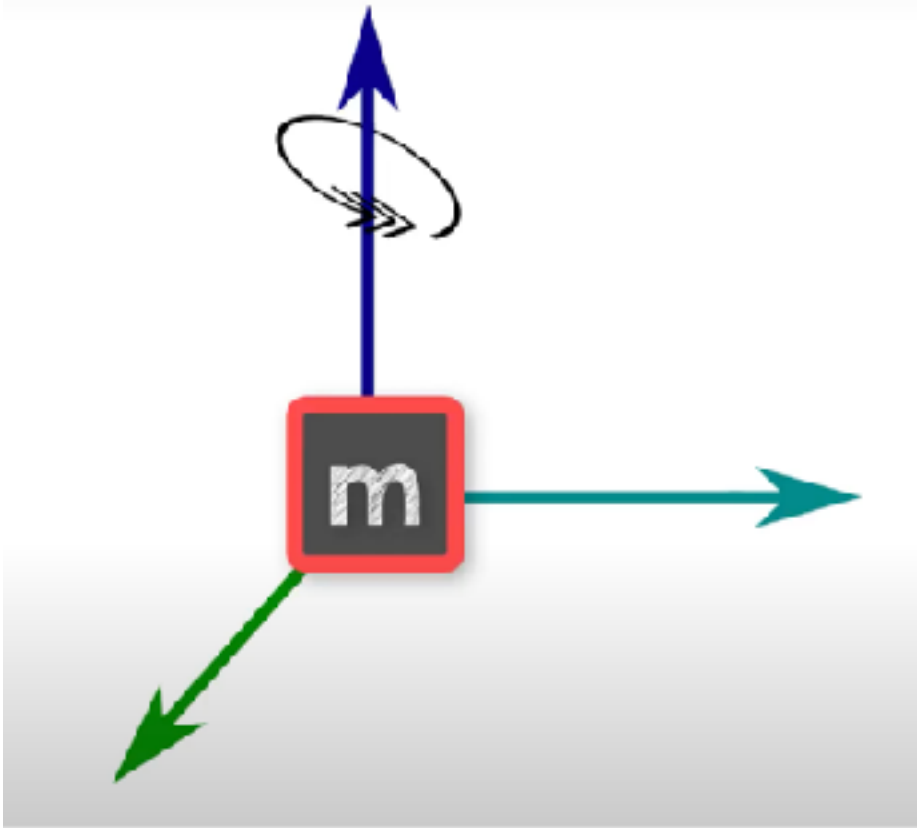


MEMS Accelerometer



How Gyroscopes Work?

The Coriolis Effect

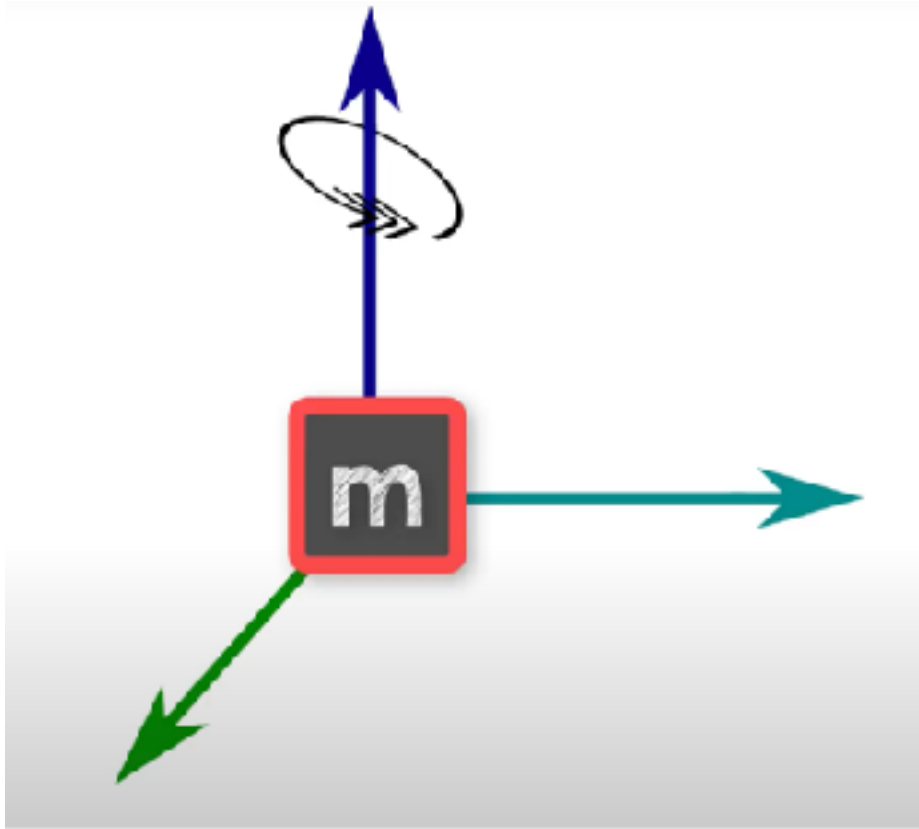


- Assume V_x
- Apply ω
- Experiences a fictitious force $F(\omega, V_x)$ following right hand rule

The Coriolis Effect

How Gyroscopes Work?

The Coriolis Effect

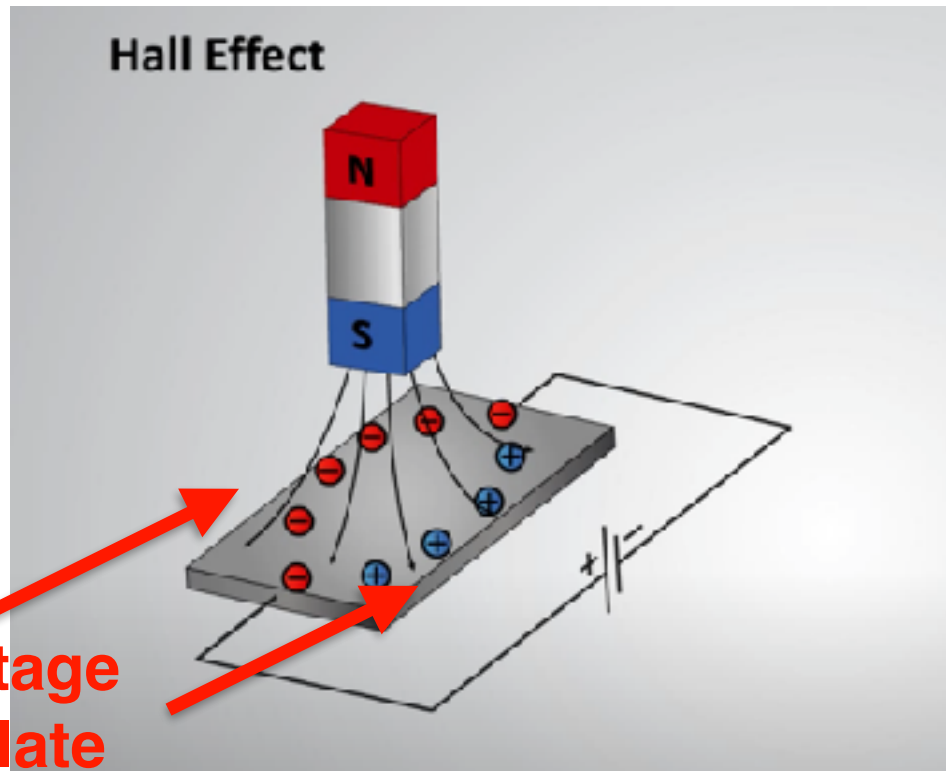


- Assume V_x
- Apply ω
- Experiences a fictitious force $F(\omega, V_x)$ following right hand rule

Can measure F in a similar fashion and use it to recover ω

How Magnetometers Work

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



**Measure voltage
across the plate**

Rest of this Lecture

- Basic principles of operation of different IMU sensors: accelerometer, gyroscope, magnetometer
- **Understanding Sources of Errors**
- Dead reckoning by fusing multiple sensors

Gyroscope

Measured angular velocity:

$$\tilde{\omega} = \omega + b + \eta$$

True angular velocity

Bias

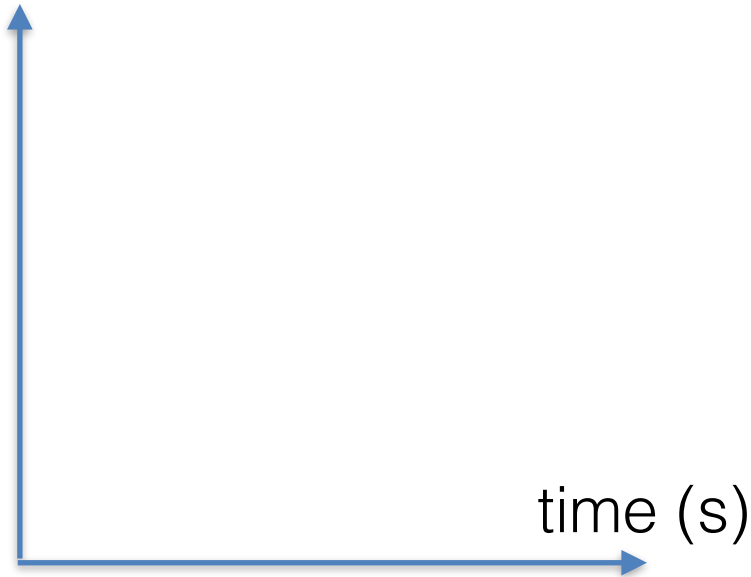
Noise (Gaussian, zero mean)

- How to get from angular velocity to angle?
 - Integrate, knowing initial position
- Linear integration? What are we missing?

Gyroscope- Some Math

Gyro Integration

Angle (degrees)



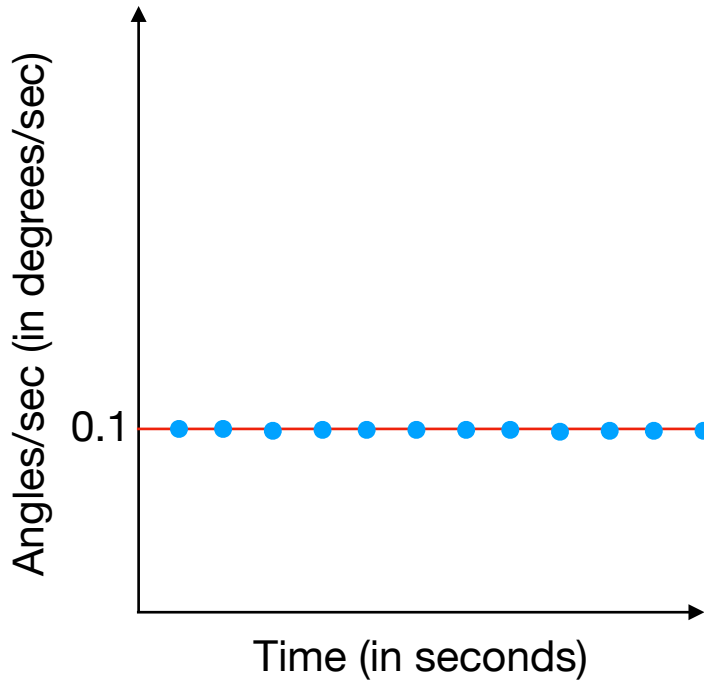
- Let's plot this for gyro measurement and for orientation
- Let's include ground truth and measured data for each

Consider:

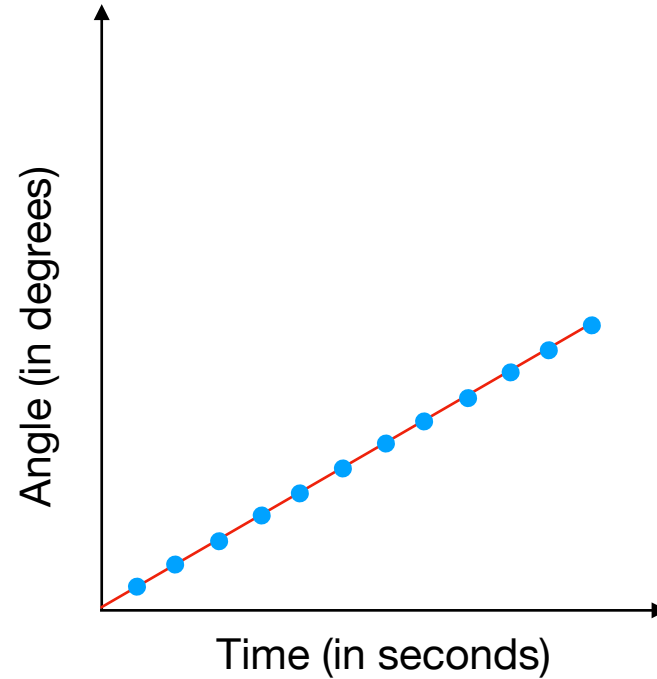
- linear (angular) motion, no noise, no bias
- linear (angular) motion, with noise, no bias
- linear (angular) motion, no noise, bias
- nonlinear motion, no noise, no bias

Gyro integration: linear motion, no noise, no bias

**Gyro Measurement
(angular velocity vs time)**



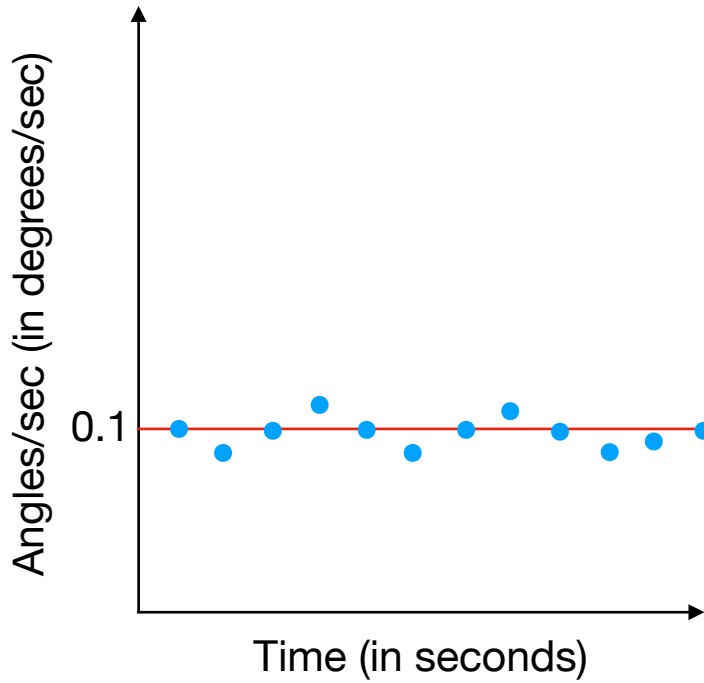
**Integrated Orientation
(angle vs time)**



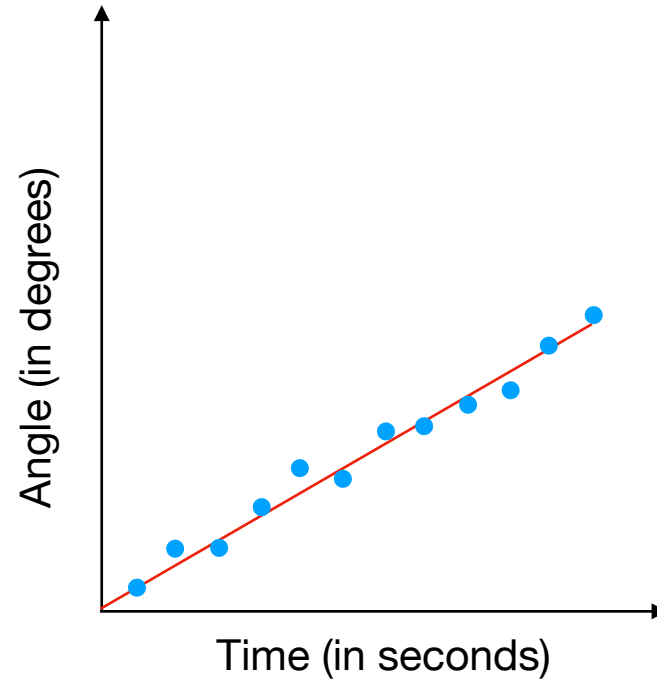
- Ground truth
- Measured/estimated angle

Gyro integration: linear motion, noise, no bias

Gyro Measurement

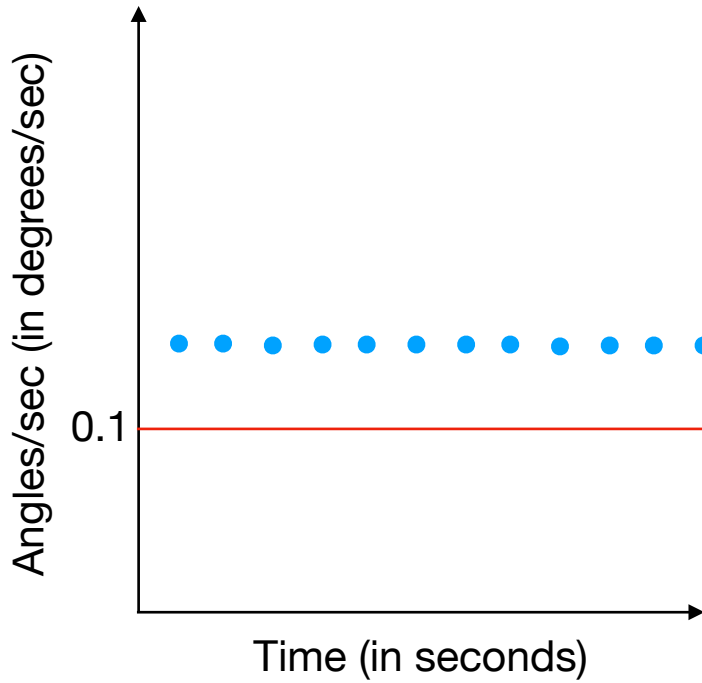


Integrated Orientation

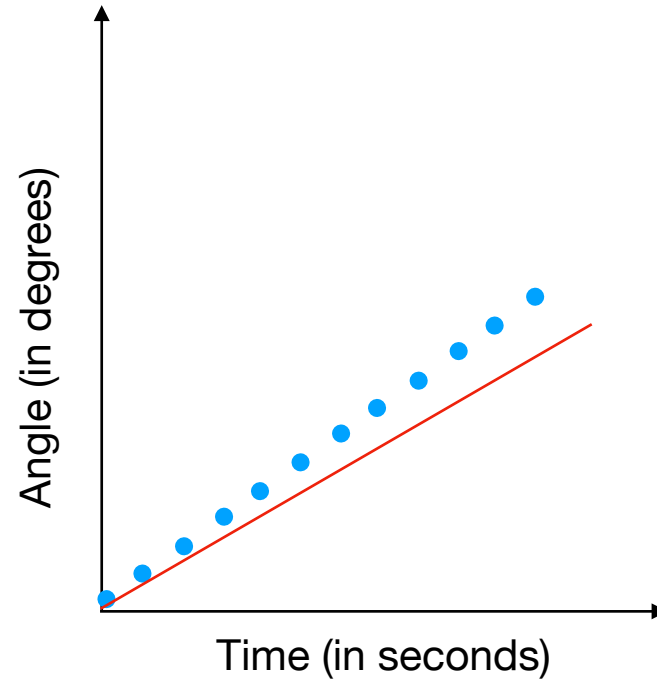


Gyro integration: linear motion, no noise, bias

Gyro Measurement

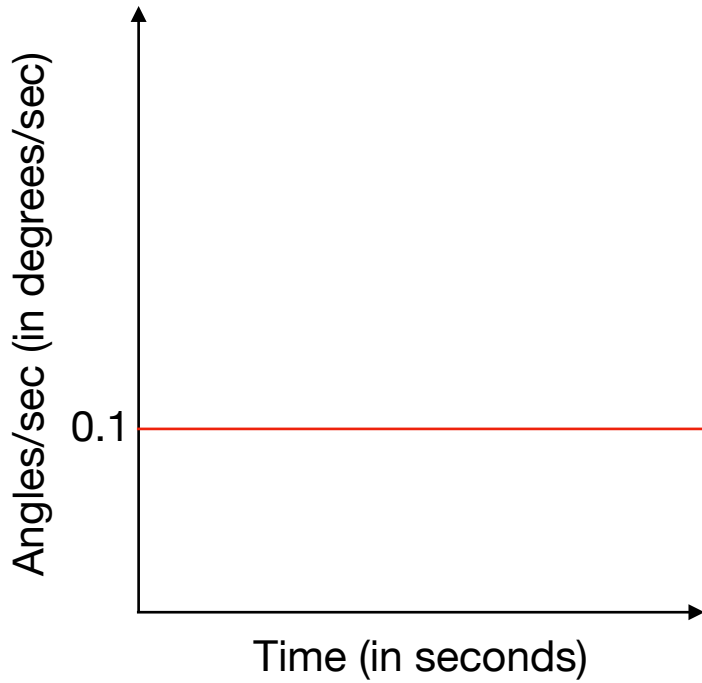


Integrated Orientation

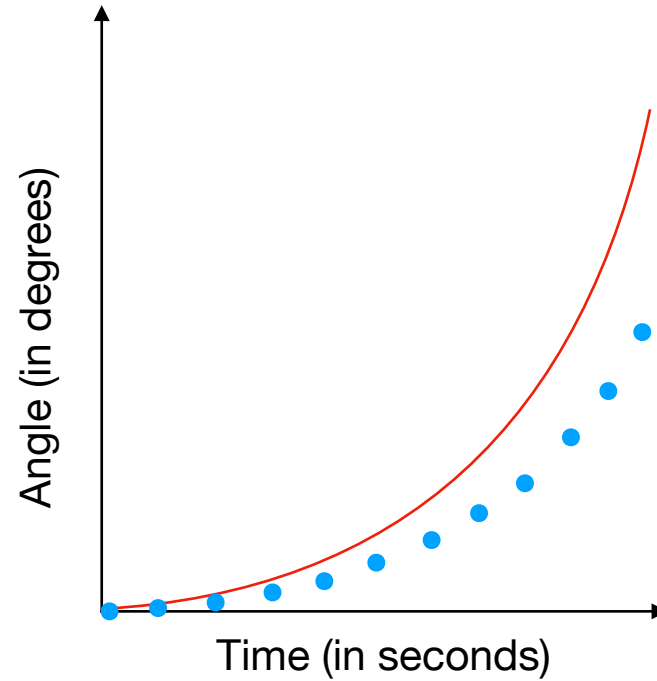


Gyro integration: nonlinear motion, no noise, no bias

Gyro Measurement



Integrated Orientation



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

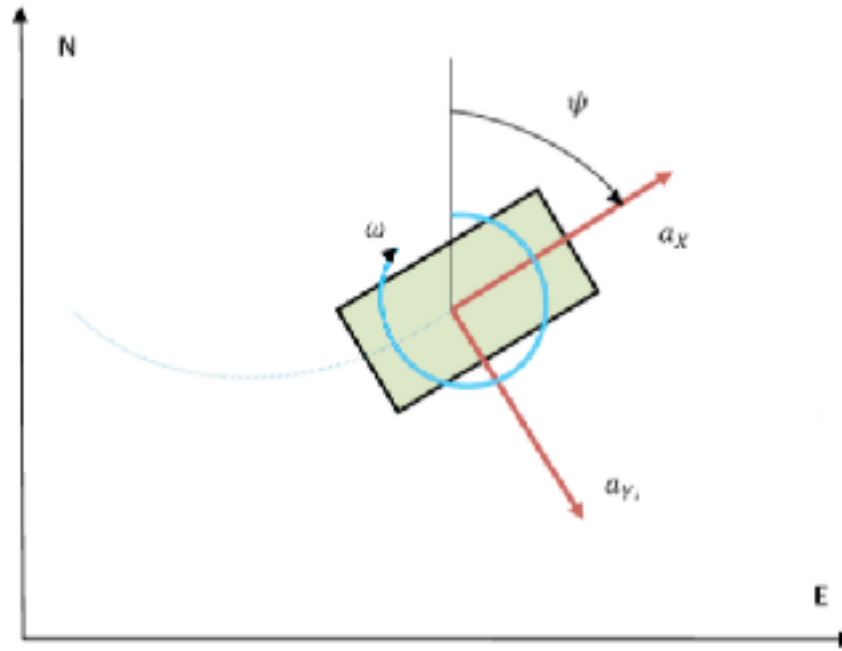
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Dead Reckoning

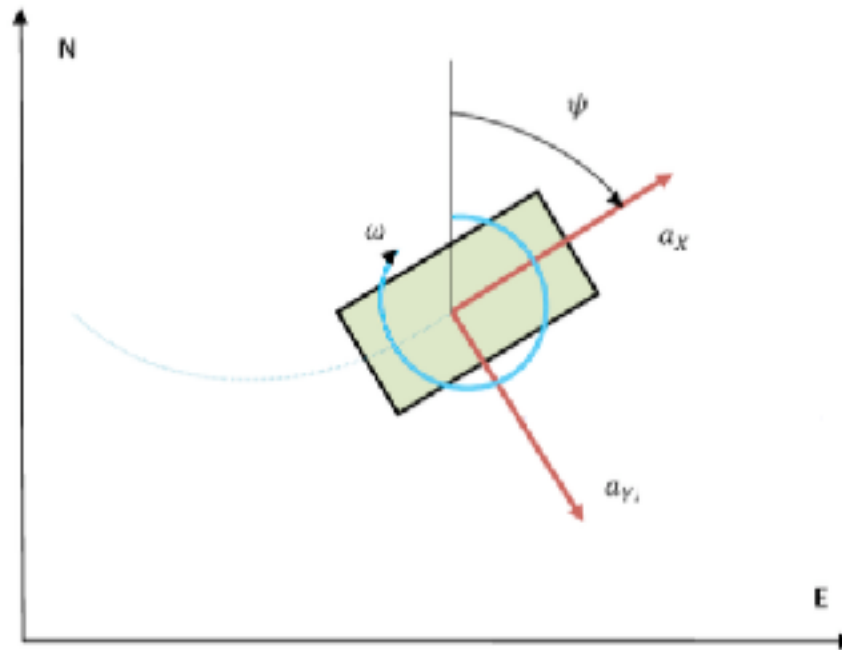
- The process of calculating one's current position by using a previously determined position, and advancing that position based upon known or estimated speeds over elapsed time and course
- Key things to keep in mind:
 - Frames of reference
 - Orientation change

2D Inertial Navigation in Strapdown System



- Have a_x , a_y , ψ ; want E, N

2D Inertial Navigation in Strapdown System



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

2D Inertial Navigation in Strapdown System

$$\begin{bmatrix} a_N \\ a_E \end{bmatrix} = \begin{bmatrix} \cos\psi & -\sin\psi \\ \sin\psi & \cos\psi \end{bmatrix} \begin{bmatrix} a_X \\ a_Y \end{bmatrix}$$

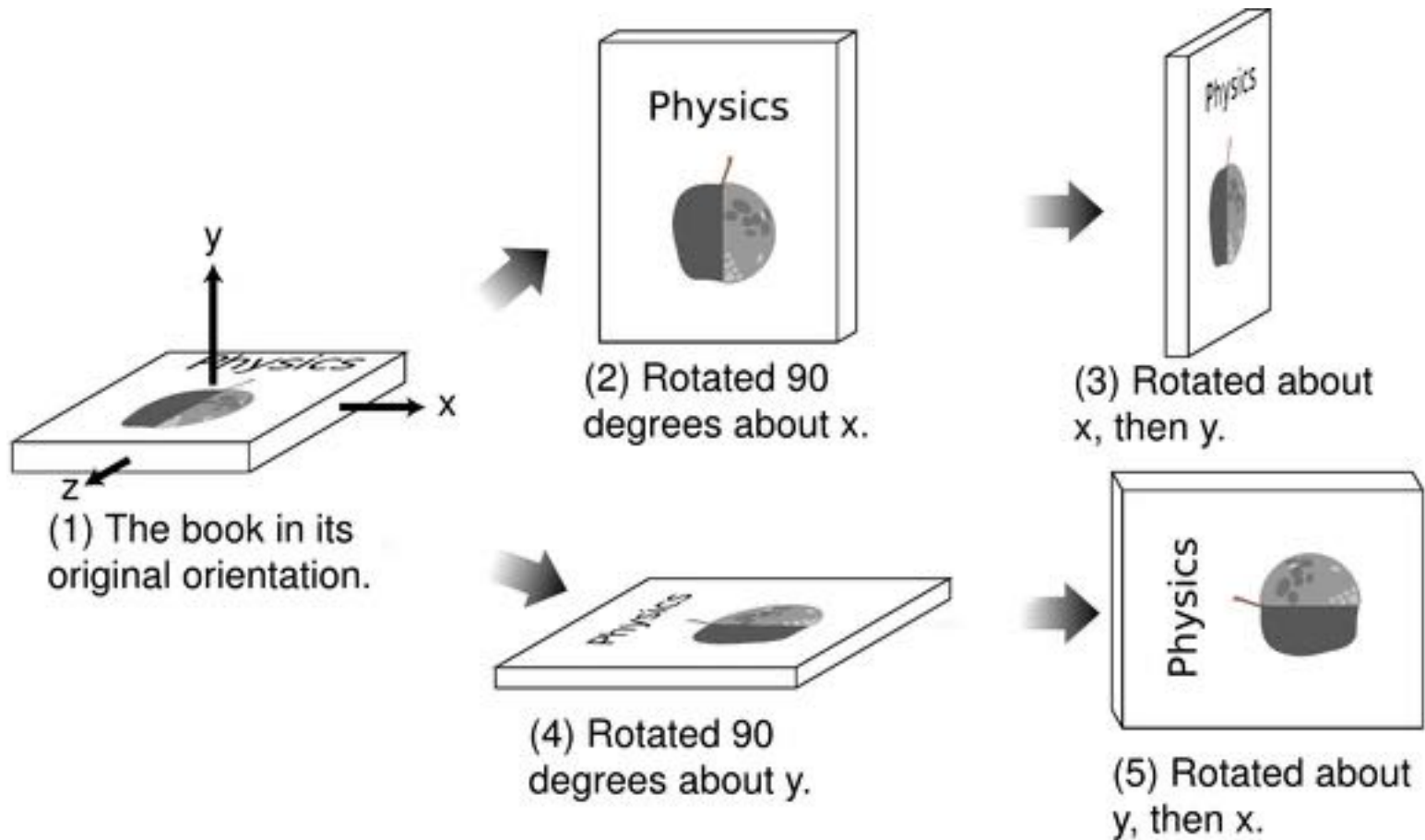
$$V_N(t) = V_N(t_0) + \int_{t_0}^t a_N(t) dt$$

$$X_N(t) = X_N(t_0) + \int_{t_0}^t V_N(t) dt$$

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

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

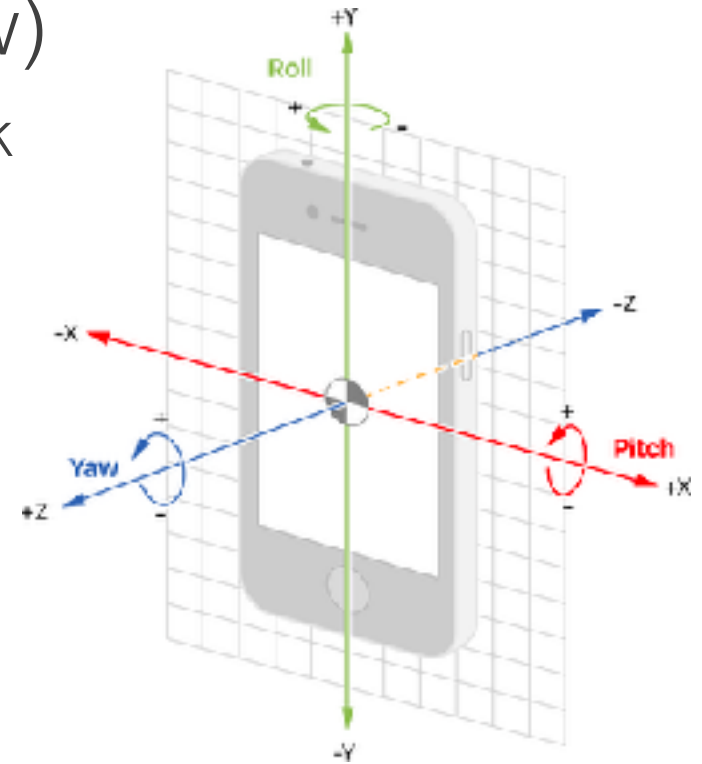
How about 3D Rotations?



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



Lecture Recap

- Importance of IMUs for navigation and sensing
- Coordinate systems and 6DOF
- IMU history and current use cases
- Basic principles of operation of different IMU sensors: accelerometer, gyroscope, magnetometer
- Understanding Sources of Errors
- Dead reckoning by fusing multiple sensors

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3. Case-study based application of inertial sensing:
Pothole patrol
4. Practical approaches to accounting for sensory noise in real-world settings.

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