



Mobile Robot Programming Laboratory

Lab 3

Thursday Week 3

<http://www.andrew.cmu.edu/course/16-362-862>



Agenda Today

CarnegieMellon
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- Administrative Issues
- Estimation vs Prediction
- WMR Modeling
- Stuff Worth Knowing
- Concurrent Programming
- Overview of Lab 3



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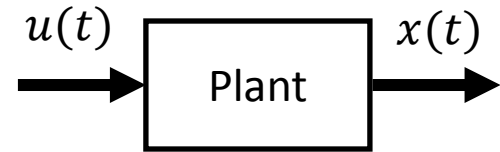
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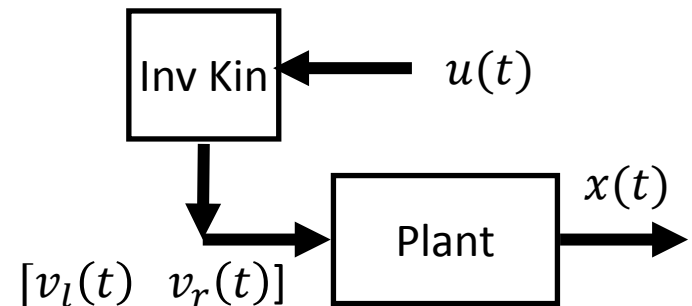


Prediction

- System model relates
state: $x = [x \ y \ \theta]$ to...
inputs: $u = [V \ \omega]$
That you are **about to send**
to the robot.
- Relationship is a
differential equation.
- Sometimes you also need
inverse kinematics
 - what wheel velocities
should I **command** to get
 $u = [V \ \omega]$ used in first
bullet?



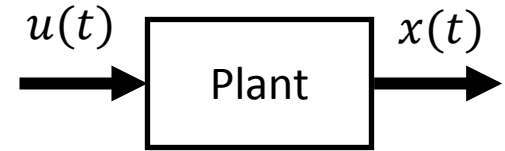
$$\dot{x} = f(x, u)$$





Prediction

- In prediction, knowing \dot{x} is not enough.
- You want to know x - where the robot **will go** if you command some $u(t)$.
- In a computer, integrate in discrete time.



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

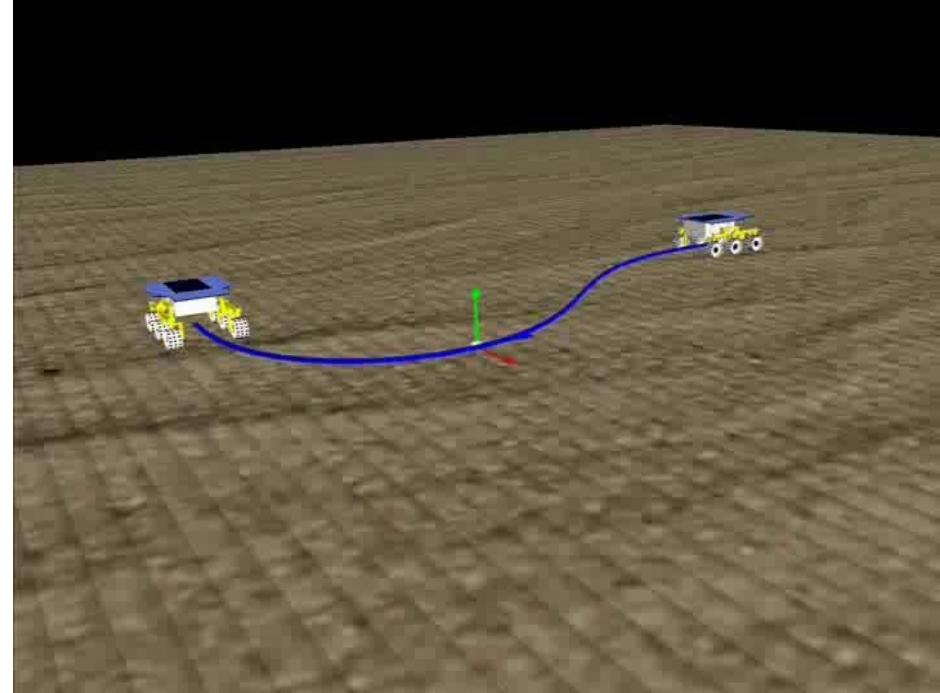
$$x = \int f(x, u) dt$$

$$x = \sum f(x, u) \Delta t$$



Terrain Following in 3D

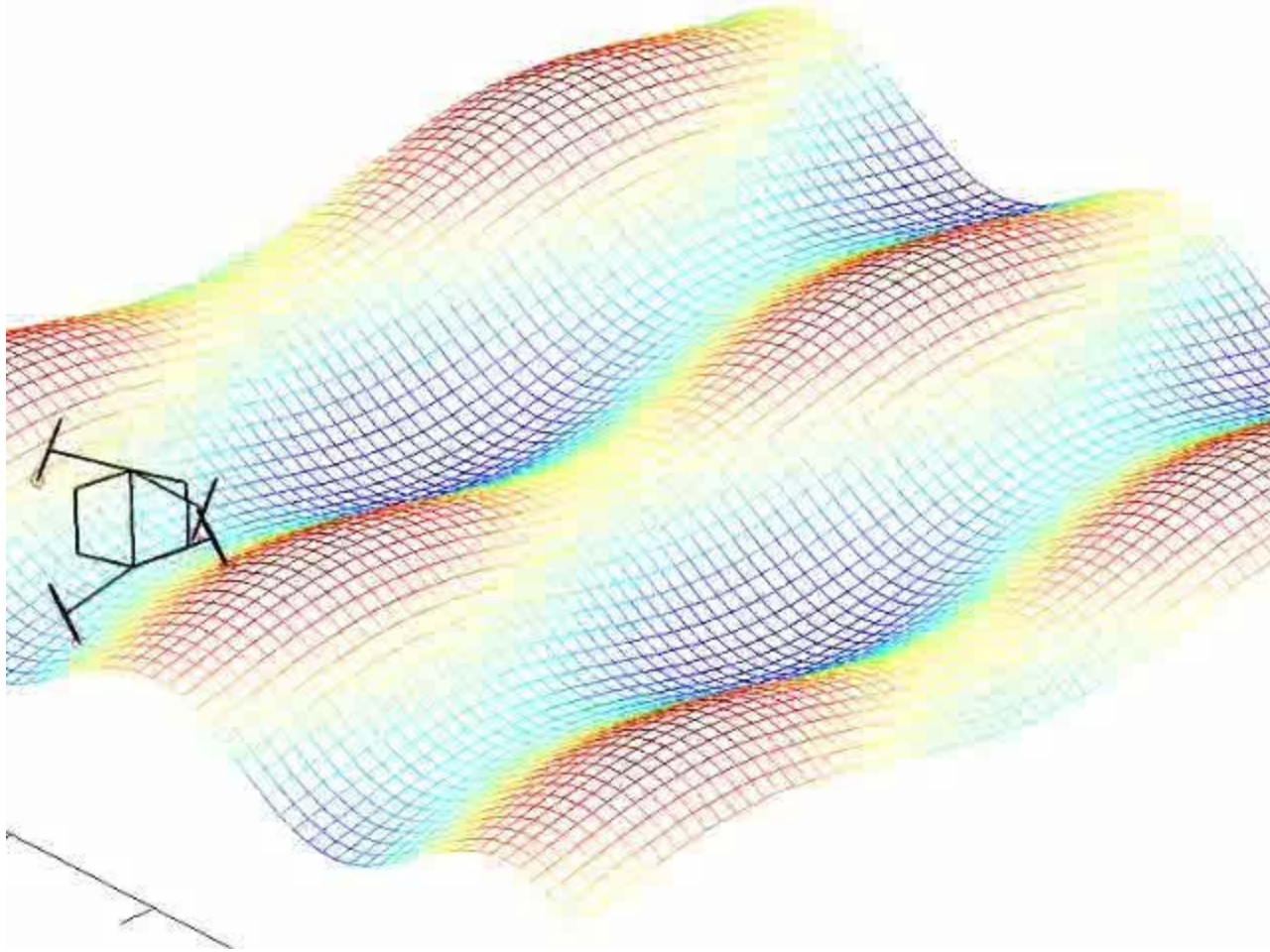
- Prediction in 3D
requires modeling of:
 - Suspension articulation
 - Terrain Contact
 - Wheel Slip





Zoë Experiment

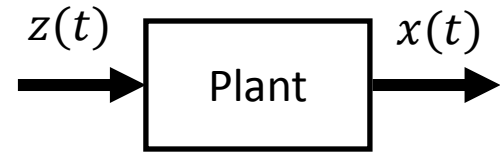
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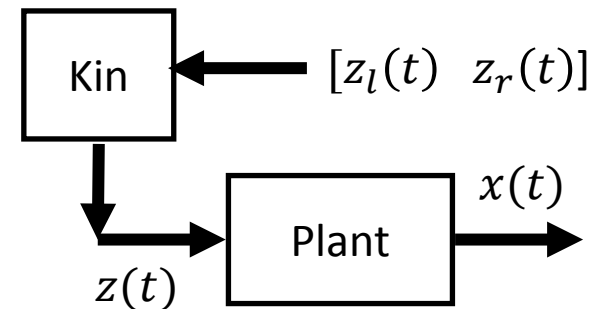


Estimation

- System model relates
state: $x = [x \ y \ \theta]$ to...
measurements: $z = [V \ \omega]$
That **you just received** from
the robot.
- Relationship is a
differential equation.
- Sometimes you also need
forward kinematics
 - What speeds $z = [V \ \omega]$
are consistent with the
wheel speeds I am
measuring now.



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





Estimation

- In estimation, knowing \dot{x} is not enough.
- You want to know x - where the robot **is now** based on the measurement history $z(t)$.
- In a computer, integrate in discrete time.



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

$$x = \int f(x, z) dt$$

$$x = \sum f(x, z) \Delta t$$



Comparison

- 1) The math is **identical**.
- 2) You decide what the symbols mean.
- 3) If written generically, you need to write the code only once.

- Prediction

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

$$x = \sum f(x, u) \Delta t$$

- Estimation

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

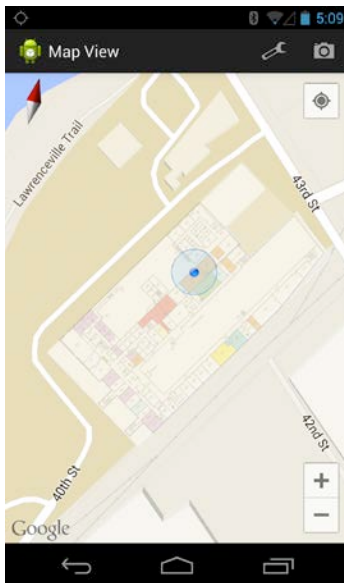
$$x = \sum f(x, z) \Delta t$$



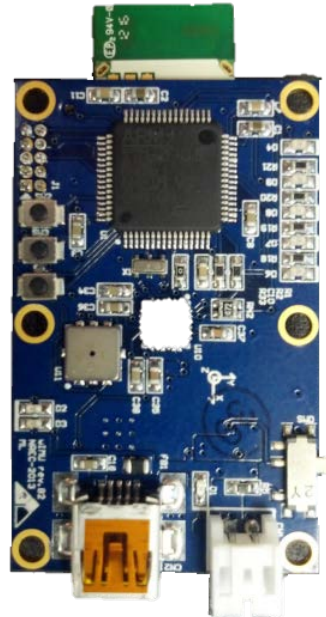


Personal Inertial Navigation

- Indoor “GPS” for airports, malls.
- Entire Inertial Navigation System runs on a microcontroller.



Cell Phone
User Interface



Board Layout



Laced Into Shoe

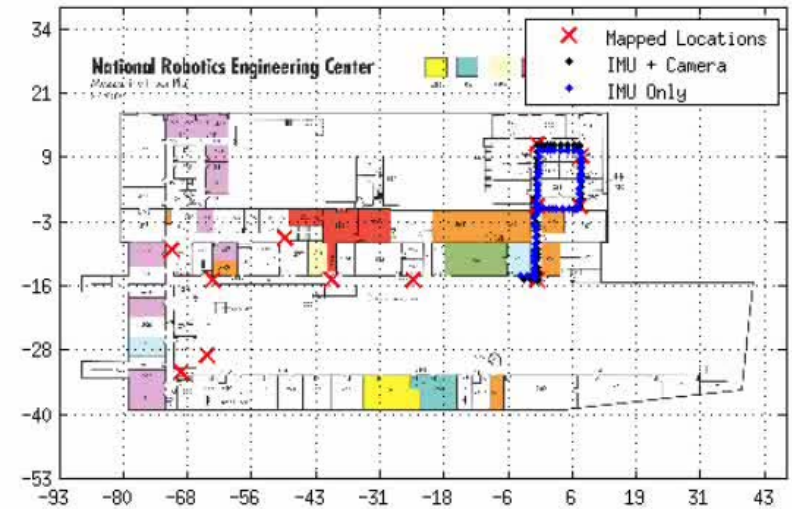
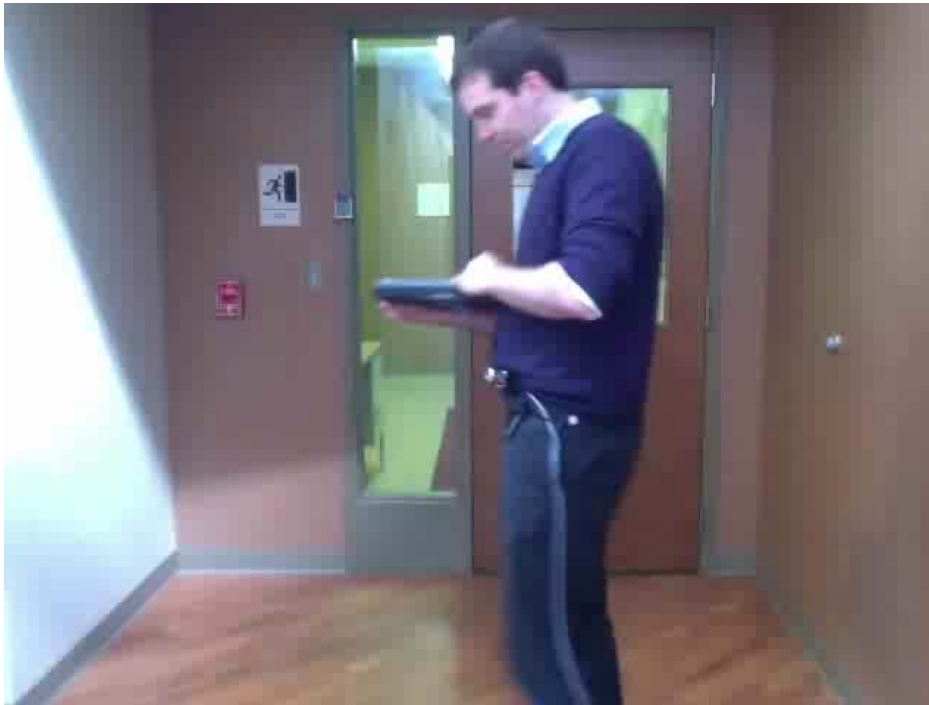


User
Interface

Wireless
IMU



Personal Navigation





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Implementing “Odometry”

- Step 1: Differentiate encoders

$$v_l = \frac{ds_l}{dt}$$

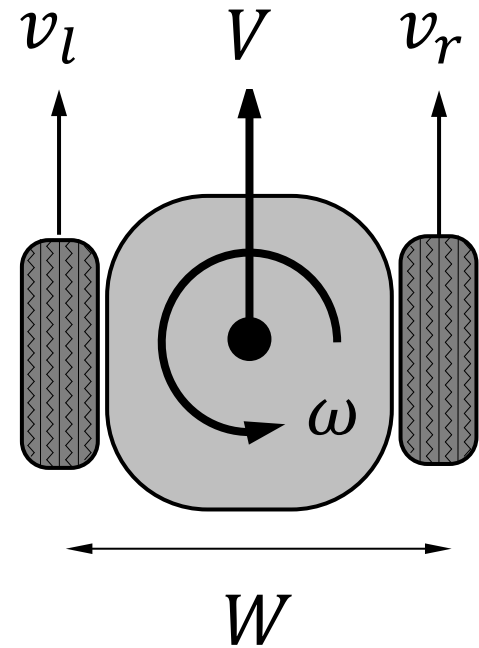
$$v_r = \frac{ds_r}{dt}$$

- Step 2: Find linear and angular velocity of rigid body (called “robot”).

– Inverse Kinematics

$$V = \frac{(v_r + v_l)}{2}$$

$$\omega = \frac{(v_r - v_l)}{W}$$





Recall Path definitions

- Linear velocity

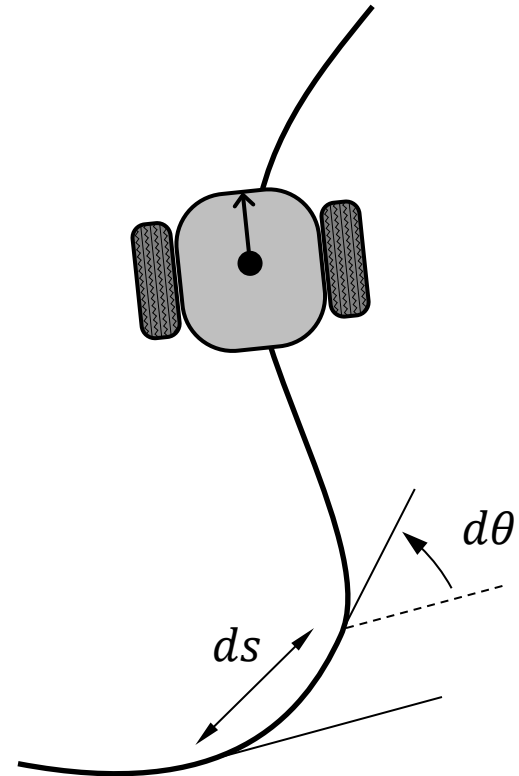
$$V = \frac{ds}{dt} \rightarrow ds = V dt$$

- Angular velocity

$$\omega = \frac{d\theta}{dt} \rightarrow d\theta = \omega dt$$

- Curvature

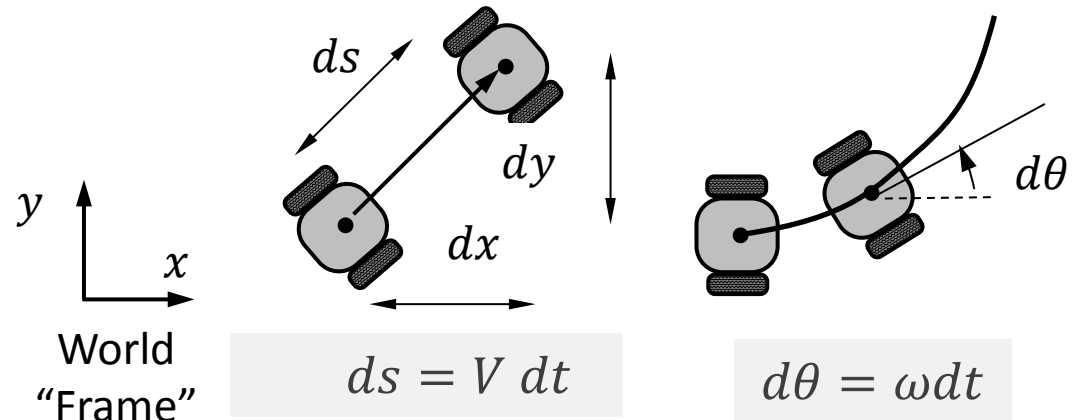
$$\kappa = \frac{d\theta}{ds} \rightarrow d\theta = \kappa ds$$





Implementing “Odometry”

- Step 3: Compute incremental rotation and translation (see opposite)
- Step 4: Project translation onto the world frame:



$$\begin{aligned} dx &= \cos(\theta) ds \\ dy &= \sin(\theta) ds \end{aligned}$$



Implementing “Odometry”

- Step 5: Add it all up.
- It matters (sometimes alot) if you update theta first or last.
- “Midpoint” algorithm (sort of..)
 - $\theta = \theta + \omega dt/2$
 - Do x, y
 - $\theta = \theta + \omega dt/2$

```
x=x(t0)
y=y(t0)
θ=θ (t0)
while( t < tf)
    θ = θ + ωdt
    x = x + Vcos(θ)dt
    y = y + Vsin(θ)dt
end
```

This is most of what you need for this lab



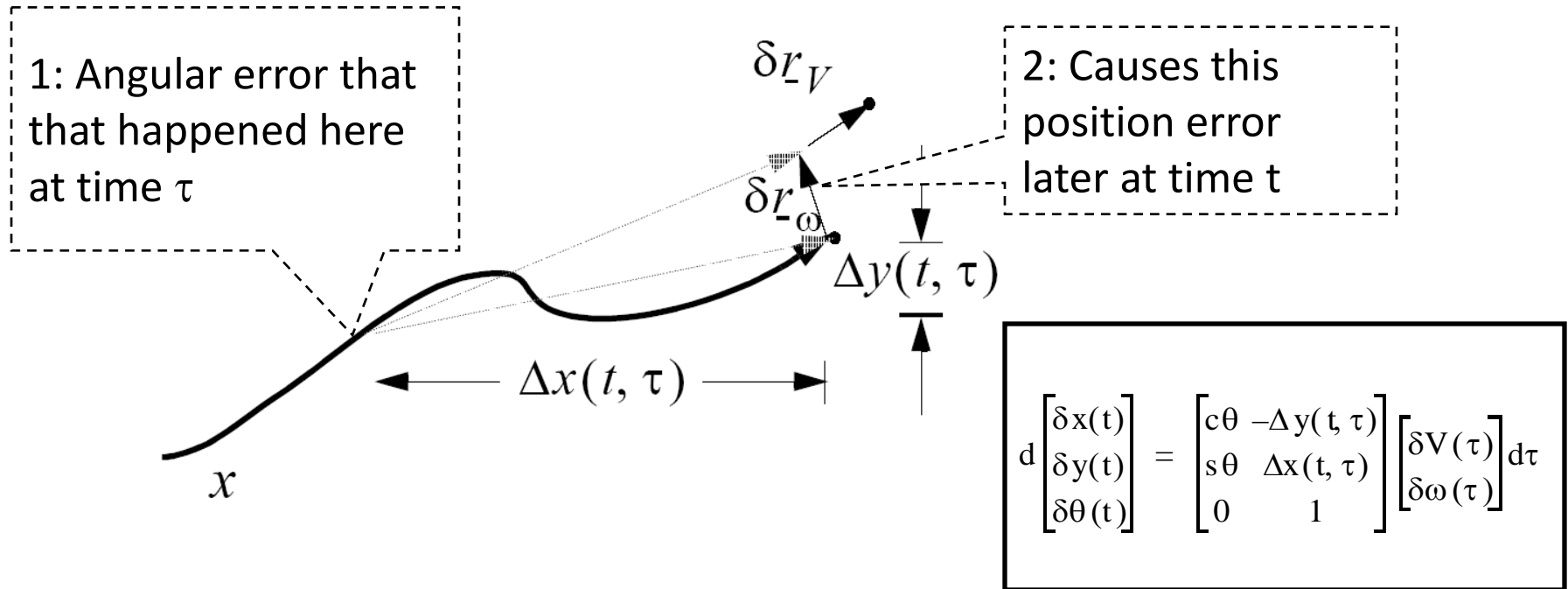
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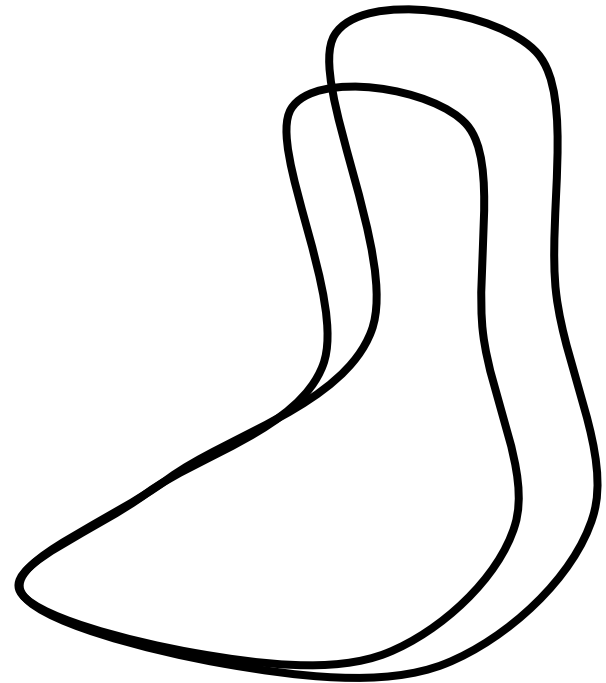
Eg: Error Propagation



- Error propagation process simply adds up the impact of every historical error to produce the present error.
- **Dead reckoning never forgets an error.**



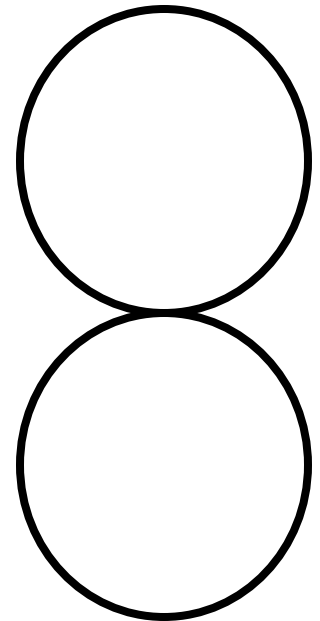
- Some errors (e.g. velocity scale errors) vanish on closed trajectories.
- So closed trajectories are not the way to test your encoder scale factor.





Insights: Symmetry

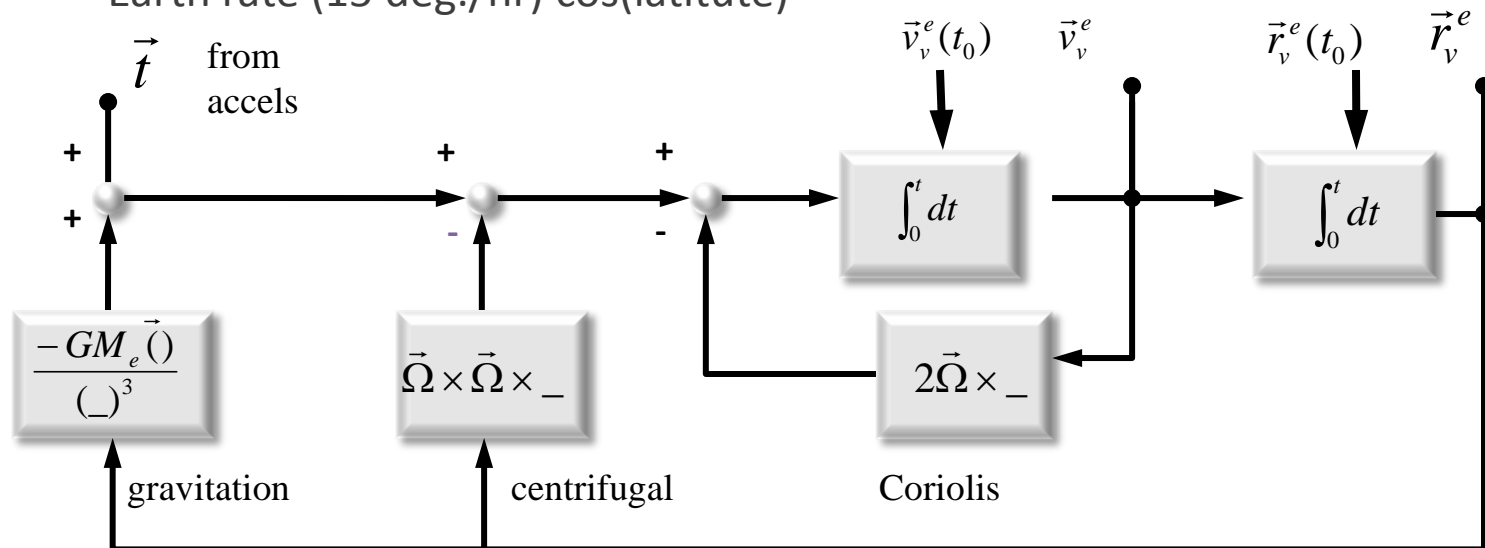
- Some errors (e.g. gyro bias) vanish at the centroid of the trajectory.
- The wrong way to assess gyro bias.





INS: Pragmatic Issues

- Need good initial conditions for:
 - Position, attitude, heading
 - These are almost never easy to get, heading is especially hard.
- Must remove from accelerometers:
 - Effect of gravity
 - Centrifugal and Coriolis forces
- Must remove from gyros:
 - Earth rate (15 deg./hr) $\cos(\text{latitude})$





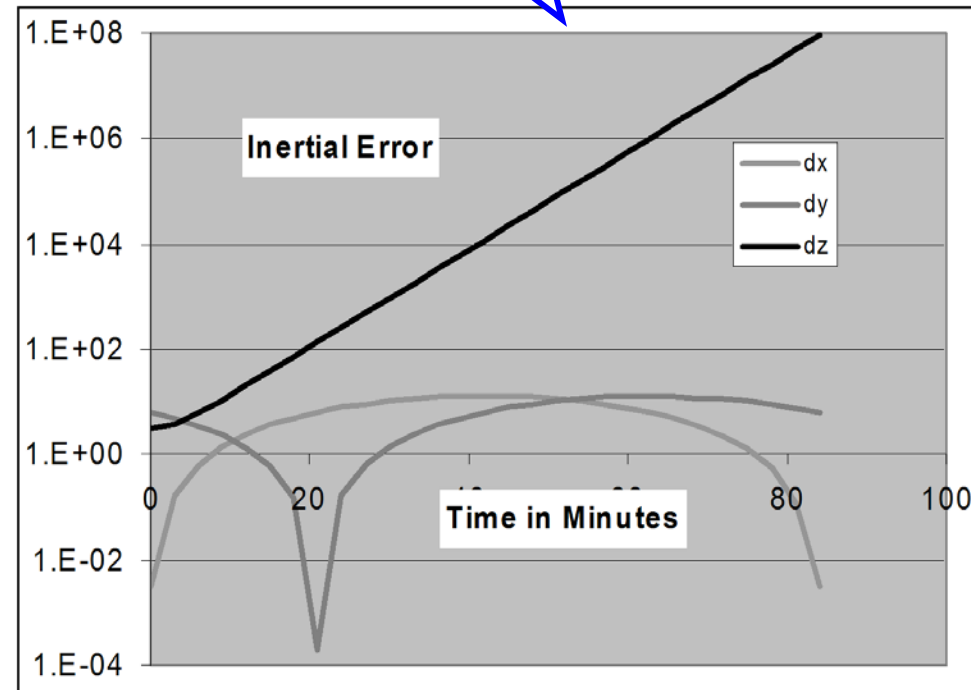
- If the accelerometer biases are constant, the solutions are:

$$\delta x = \frac{\delta t_x}{g_0/R_0} \left[1 - \cos \left(\sqrt{\frac{g_0}{R_0}} t \right) \right]$$

$$\delta y = \frac{\delta t_y}{g_0/R_0} \left[1 - \sin \left(\sqrt{\frac{g_0}{R_0}} t \right) \right]$$

$$\delta z = \frac{\delta t_z}{2g_0/R_0} \left[\cosh \left(\sqrt{\frac{2g_0}{R_0}} t \right) \right]$$

Error magnitudes
for 1 micro-g
biases.



- Gravity field is a mixed blessing !!



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Consistency

- Most control algorithms perform better on, and many require, consistent data.
 - You **would not** change a variable by a small random amount in the middle of an algorithm.
 - Almost any algorithm will break or perform erratically. Robots will do random, even dangerous stuff
- Suppose there is no latency in sensor data, encoder data comes in at 50 Hz, robot is moving, and your code to pick up an object looks like so:
 - `myPose = estimatePose(lastMessage)` % **first** time
 - `objPose = locateObject(myPose)`
 - `myPose = estimatePose(lastMessage)` % **second** time
 - `myTraj = computeTrajectory(objPose,myPose)`



Consistency

- This algorithm will always underestimate the distance to the object because the robot moved between **looking** and **doing**.
 - Any robot always does move. Its up to you to deal with it in a principled manner.
 - You can at least use state estimates that are consistent in time (occur at the same time or roughly the same time).
- If the variables involved are logical rather than signals, inconsistency is a major major safety issue.
- Inconsistency is bad and the badness is proportional to:
 - The delay between updates
 - The speed of the robot
 - The sensitivity of the algorithm to inconsistency.



Concurrency

- **YOU** must control when changes to signals are allowed to enter your algorithm.
- Yet, any concurrent (multiprocessor, multiprocess, multi-thread) system **does not control, out of the box**, precisely when shared data is accessed.
 - So, it **does change variables by random amounts in the middle of an algorithm**.
- **Even ROS is doing this under the hood** when you access the LatestMessage struct.
- Moral:
 - **DO NOT** read the ROS lastMessage more than once in your “main loop” (the highest level loop of your entire algorithm).
 - Update, using incoming data (encoder, lidar, other) the robot and world state estimate however often you like in a concurrent thread/process/processor **BUT** introduce it **once at the top** of your main loop and use that state estimate everywhere else until your code returns to the top of main loop again.
- Mind these rules and save yourself sleepless nights in October and November chasing bugs that come and go randomly.



Concurrency - Callbacks

- Callbacks are not executed in your main thread.
Hence concurrency issues...
- Doing odometry every time an encoder message arrives at 50 Hz is a **good** idea because it improves accuracy of the state estimate.
- BUT, passing those updates to your main algorithm at random times is a **bad** idea.
- You can use a semaphore or synchronous code but its pointless because your code is way slower than 50Hz anyway.



Concurrency - Callbacks

- Makes more sense to simply process encoder data as fast as you can and **wait** if there is no new data when you need it.
 - Works when your code runs at 400Hz because it waits for new data (and runs at 50Hz)
 - Works when your code runs at 10 Hz (because it uses the latest encoder update)
 - BTW you could queue the updates and process all of them. That is slightly more accurate.
- Optimal solution is to **do odometry at 50Hz but update state in the main loop once per iteration.**
 - However doing odometry as fast as the main loop using the latest encoder data at that time is good enough for this course.



Dealing with Latency

- Everything about concurrency above matters even without sensor data latency.
- Latency is a second, compounding issue.
 - Encoder messages are always late with respect to reality.
 - The latency of the real data cannot be eliminated.
 - So, **your state estimate is always where the robot “was”**.
- It is possible to use prediction to remove the **latency** error at the cost of introducing **prediction** error.
 - Compute DRState as usual every cycle
 - Use PrState where $\text{PrState} = \text{DRState} + (V, w) \times \text{latency}$
- In addition to incoming measurements (encoders, lidar), there are equivalent latency issues for the commands going out to the robot.



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Preparation

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- Did everyone read lab 3 before today?



Lab 3 Preparation

- Click for [Lab3 Writeup](#)