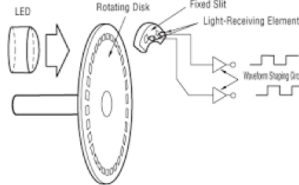


Intro to Robotics

1. Senses = Sensors

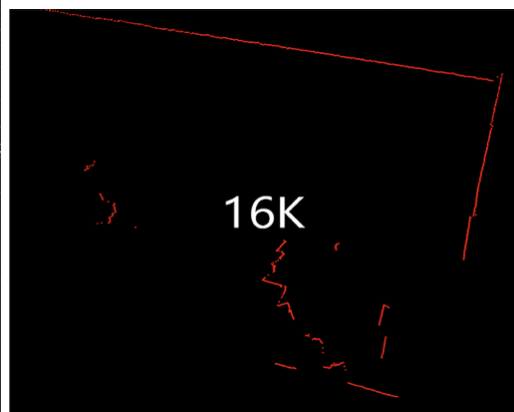
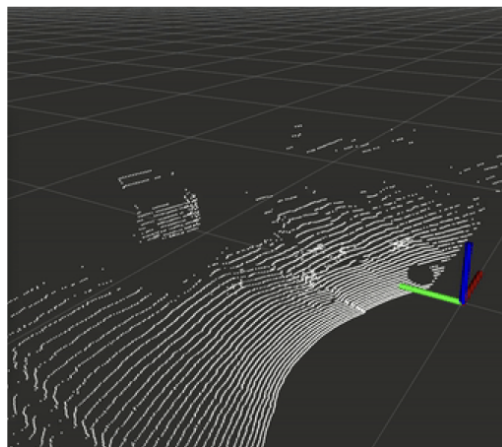
- a. Lots of different kinds of sensors
- b. LIDAR (Laser Imaging Detection & Ranging)
- c. GPS
- d. Motor Encoders
- e. Camera
- f. ...



g.

2. Sensor Problems

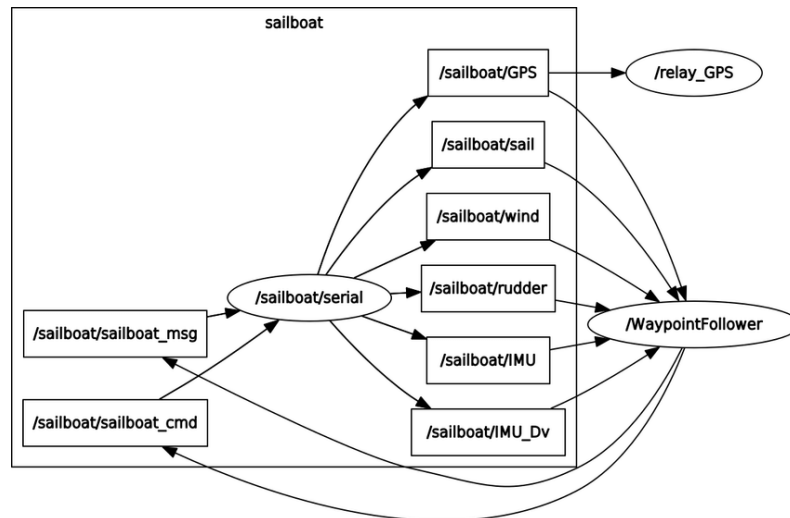
- a. Noise
  - i. Measurements have precision
  - ii. That precision has error!
    - 1. "sensor accurate to  $\pm 0.1$  m"
- b. Drift
  - i. Measurements move over time
- c. Most sensors have both!



d.

### 3. Communication

- a. How to get (stream) of data from sensors?
- b. ROS (Robot Operating System)
  - i. Not actually an operating system
  - ii. middleware
  - iii. ROS is a messaging system
    1. Each program is a node
    2. Nodes can publish messages (at some frequency)
      - a. Each message has a topic
    3. Nodes can subscribe to message topics
    4. ROS routes published messages to nodes that subscribe to them!
- c. The beauty:
  - i. Compartmentalized design
  - ii. All processing becomes nodes!



d.

### 4. Low-Level

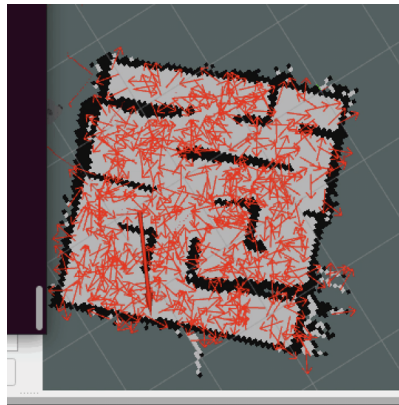
- a. Need to know:
  - i. Where am I?
  - ii. What direction am I pointing?
  - iii. How fast am I moving?
  - iv. ...
- b. These are called state variables
- c. Need to process sensor data
  - i. But sensors are unreliable!
    1. Can't trust sensor data at face value! (noise)
    2. Can't trust sensor data depending on how long sensor has been running (drift)
- d. Can't tell robot what to do without this info!

## 5. Localization

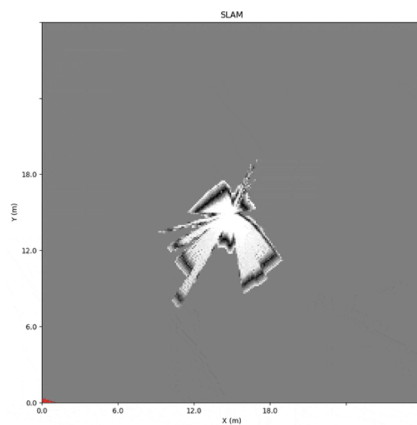
- a. Localization = Figuring out state variables from (unreliable) sensors
- b. Open area of research
- c. Two general styles:
  - i. Map-based
    - 1. If we knew the map we were in.....have a reference frame to compare to!
  - ii. Map-free
    - 1. Kalman filters
- d. Often these processes are used together

## 6. Map-Based Localization

- a. If we have the complete map
  - i. For instance a lidar scan of the terrain
  - ii. Find poses (Position + Orientation) in map that align with lidar scans
  - iii. Monte-carlo Localization
    - 1. Find pose (amongst a bunch) that maximizes prob of sensor data
- b. Problem: No advanced knowledge?
  - i. SLAM: Simultaneous Localization and Mapping
  - ii. Tradeoff:
    - 1. Map size vs accuracy of map



c.



d.

## 7. Map-Free

a. Control theory gives us a set of equations to implement

i. Whenever we issue a control:

$$\begin{aligned}\hat{\vec{x}}_{t|t-1} &= \mathbf{F}_t \hat{\vec{x}}_{t-1|t-1} + \mathbf{B}_t \vec{u}_t \\ \hat{\mathbf{P}}_{t|t-1} &= \mathbf{F}_t \mathbf{P}_{t-1|t-1} \mathbf{F}_t^T + \mathbf{Q}_t\end{aligned}$$

ii. Whenever we get sensor measurement:

$$\begin{aligned}\tilde{\vec{y}}_t &= \vec{z}_t - \mathbf{H}_t \hat{\vec{x}}_{t|t-1} \\ \mathbf{S}_t &= \mathbf{H}_t \hat{\mathbf{P}}_{t|t-1} \mathbf{H}_t^T + \mathbf{R}_t \\ \mathbf{K}_t &= \hat{\mathbf{P}}_{t|t-1} \mathbf{H}_t^T \mathbf{S}_t^{-1} \\ \hat{\vec{x}}_{t|t} &= \hat{\vec{x}}_{t|t-1} + \mathbf{K}_t \tilde{\vec{y}}_t \\ \mathbf{P}_{t|t} &= (\mathbf{I} - \mathbf{K}_t \mathbf{H}_t) \hat{\mathbf{P}}_{t|t-1} \\ \hat{\vec{x}}_{t|t} &= (\mathbf{I} - \mathbf{K}_t \mathbf{H}_t) \hat{\vec{x}}_{t|t-1} + \mathbf{K}_t \vec{z}_t\end{aligned}$$