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## Lecture: Sensors

#### 15-491, Fall 2008





# Outline

- Sensor types and overview
- Common sensors in detail
- Sensor modeling and calibration
- Perception processing preview
- Summary

## **Open Loop Control**

• No sensing input



## Why Sense?

- To acquire information about the environment and oneself
- Open loop control suffers from
  - Uncertainty, changes in the world
  - Error detection and correction





## The Sensing Loop

• "Feedback" control



### **Issues to Address**

- What sensors to use?
- How to model the sensor?
- How to calibrate intrinsic/extrinsic models?
- What low-level processing?
- What high-level processing (perception)?

## **Comparison: Human Sensors**

- Sense:
  - Vision
  - Audition
  - Gustation
  - Olfaction
  - Tactition

- **Sensor:** 
  - Eyes
  - Ears





- Tongue

– Nose

– Skin



## **Robot Sensors**

Sense:

- Equilibrioception
- Proprioception
- Magnetoception
- Electroception
- Echolocation
- Pressure gradient

#### Sensor:

- Accelerometer
- Encoders





- Magnetometer
- Voltage sensor
- Sonar



Array of pressure sensors

## LiDar Sensing









#### LiDar Variations

Tartan Racing Team



Boss vehicle

## Sensor Examples

- (CMU) Tartan Racing Urban Challenge vehicle
- Groundhog, subterranean mapping (CMU)
   Carnegie Mellon Mine Mapping Project
- Ocean explorer www.oceanexplorer.noaa.gov



## **Popular Sensors in Robotics**

- LiDar
- Infrared
- Radar
- Sonar
- Cameras
- GPS
- Accelerometers
- Gyros, encoders
- Contact switch















## Auditory



#### **Other Robot Sensors**





## Sensors We Will Look At Today

- Exterioceptive
  - Sonar, LiDar, IR
  - Vision comes later
- Proprioceptive
  - Encoders
  - Accelerometers
  - Gyroscopes
  - GPS (hard to categorize)
  - Micro-switch

# SoNaR: Sound Navigation and Ranging

- Often called sonar, ultrasound, Sodar
- Emit a directional sound wave, and listen for echo(s), time the response











- Key assumption: sound travels at constant speed
- v=344 m/s (dry air, 21C, sea-level)
- So we have



## Power of Returned Signal

- Signal power dissipates as wave travels
  - Depends upon the shape of the wavefront
  - Driven by shape of transmitter (same for radar)

 $Area = \pi r^2 = \pi \left[ d \tan\left(\frac{\alpha}{2}\right) \right]^2 \propto d^2$ 

• Typically a directional cone

d

α



## **Reflection Strength**

• Function of surface angle and surface properties



## How To Detect the Echo?

- Electronic signal processing
- Detect sufficiently large rapid change



## **Imperfect Sensing**

- What can go wrong?
  - Speed of sound changes with temperature, pressure, humidity

$$v_{ideal} = \sqrt{\frac{\gamma k T}{m}}$$

- Surface reflection properties
- Atmospheric attenuation (finite range)
- Multiple echoees (multi-path)
- Quantization in timing
- Inaccuracies in detecting response signal onset
- Cross-talk (echoes from other sensors)

#### Sensor Noise

 Fixed object, sensor returns different values over time => random process



# **Bigger Picture: Perception**

- Given sensor readings, how does robot determine the structure and content of the world?
- Usual way is to *model* the problem



## Sensor Model

- Model the device physics to obtain the expected device properties and parameters
  - Intrinsic model: Device itself
  - Extrinsic model: Where the device is on the robot
- Collect data and *fit* model parameters
  - This is *calibration*
- Level of complexity is a trade off
  - Computation, accuracy, reliability, domain knowledge
- Often need to reason explicitly about *uncertainty*

# **Modeling Sonar**

- What should we model?
- Usually:
  - Mapping from time to range (first return only)
    - We have a physics model with parameters. Calibrate to get parameter values.
  - Model sensor uncertainty
    - How do we do this? What distribution should we use?
- Other possibilities:
  - Signal strength to surface orientation?
  - Using secondary peaks?
  - Profile of response?

### Calibration

- We have a model
  - Derived from the physics (best approach)
  - Look at data and guess a low dimensional model
- Estimate the parameters from a known setup
  - Measure signal response at different distances
    - Optionally different angles, surfaces, humidity, altitude...
  - Fit parameters to the data (e.g. regression) Outlier



## Sensor Noise Modeling

- Sensors are **never** perfect
  - Unmodeled effects
  - True randomness in the environment, robot, and sensing process
- Systematic errors (bias)
- Drift, jumps

### Sensor Bias

- Return may vary as a function of physical setup
  - Surface material/color, orientation, range, atmosphere



### Sensor Noise Model

- Enter the world of statistics
  - Usually choose a parametric model and estimate parameters e.g. Gaussian



## Sensor Filtering

- Usually apply some level of filtering to raw sensor data before feeding into rest of system
- Examples
  - Thresholding you've already seen this
  - Smoothing simple filters
  - Kalman filtering more complex filter exploiting additional domain knowledge
- Resulting estimate used to build perception models
  - Occupancy grids, trackers, etc.

## LiDar

- Light Detection and Ranging
- Different variants, we'll focus on time to return
  - Most common to robotics
  - Same model as Sonar



Narrow	pules of	F
laser lig	jht	





### LiDar

- Timed "echo" from reflection
- Speed of light >> speed of sound



### SICK LiDar

- Very common unit
- Spinning mirror assembly gives line scan
  - Ranges vary (90, 180 degree, 50+m)
  - Scanning rates vary (e.g. 20Hz, 75Hz)
  - Resolutions (e.g. 0.25 degree, 10mm)
  - Accuracy ~30mm stdev in range

Spinning mirror

### SICK LiDar Internals

• From http://web.mit.edu/kvogt/www/lidar.html











#### LiDar Variations

NREC Crusher Vehicle



Crusher with sensors

#### **Colorized LiDar**

#### Used a lot on NREC robots

http://www.aerotecusa.com/





## InfraRed

- Emitter/detector pair
- Output type
  - Digital (strength of return threshold)
  - Analog range using triangulation
- Usually short-range (<1m)
- Can be sensitive to IR sources e.g. sun

## Sharp IR Sensor





### **Proprioceptive Sensors**

## **Optical Encoders**

- Disc to measure *rotational* motion
- Out of phase IR emitter/detector pair



## **Optical Encoders**

• Direction and amount of rotation from edge transitions



### In Practice

- Electronic hardware (MCU or ASIC) provides counting, de-bouncing
- Estimate speed by sampling encoder counts
   Model to provide wheel speed from encoder counts
- How to get vehicle speeds from wheel speeds?
  This is kinematics! (Later in the course)

## Gyroscopes

- Proprioceptive sensor
- Maintaining estimate of orientation
  - Mechanical devices
  - Fiber optic gyroscope
  - Vibrating gyroscope (e.g. MEMS)

#### What is Gyroscope?

#### **Rotational Angular Velocity Sensor**

Operation principle : An angular velocity sensor that works by using the phenomenon generated by Coriolis force when angular velocity is applied to a moving object in relation to velocity and orthogonal directions.



#### Accelerometers

- Measure acceleration in a direction of travel
  - Typically MEMS device
- Also measures gravity
  - Good old relativity...
  - Can use with gyroscopes to remove gravity component
- Typically very noisy
- Need to double integrate to get position

#### Accelerometers

Sensor acceleration, mid-term accel



Lecture: Sensors

## Issues With Accelerometers/Gyros

- Noise
  - Output readings may have approximately additive Gaussian noise
- Drift
  - Signal drifts from true value over time Gyro heading
  - Usually need to integrate accelerometers

## GPS/Glonas/Galileo

- Orbiting satellites
  - Known trajectories
  - Highly precise timers
- Transmit data in Ghz band
  - Ephemeris information
  - Develop pseudo range to satellite
- Solve for receiver position
- Can also solve for velocity





# **GPS** Properties

- Many causes of error
  - Ionospheric effects, line of site clearance
  - Delays in satellite positional updates, multi-path
- Is it Gaussian?
  - Over hours, approximately Gaussian errors
  - Over short time, small error but strong bias
- Improvements
  - DGPS, WAAS (~3m accuracy at 3 sigma)
  - Use an INS (Accelerometers/gyros)

## **GPS/INS**

- Commercial solutions exist (expensive!)
- Fuse integrated INS estimates with GPS
   A big custom Kalman filter (more later)



## Summary

- Know about
  - A whole class of sensors
  - Typical problems with sensors, and sensor uncertainty
  - Basic approach to modeling a sensor
  - Basic filtering techniques