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Lecture 2: Sensors

15-491, Fall 2007





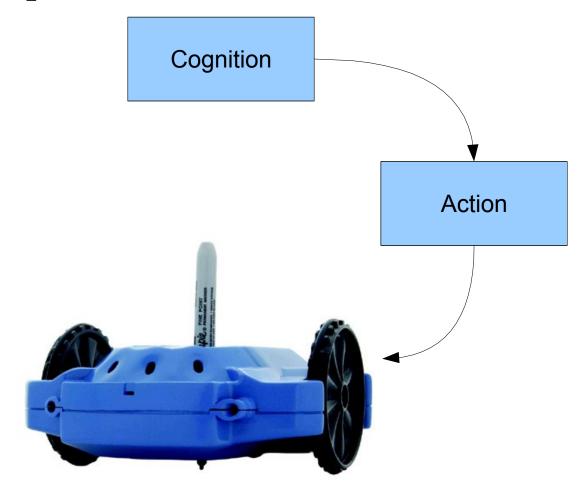
Outline

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

2

Open Loop Control

No sensing input



Why Sense?

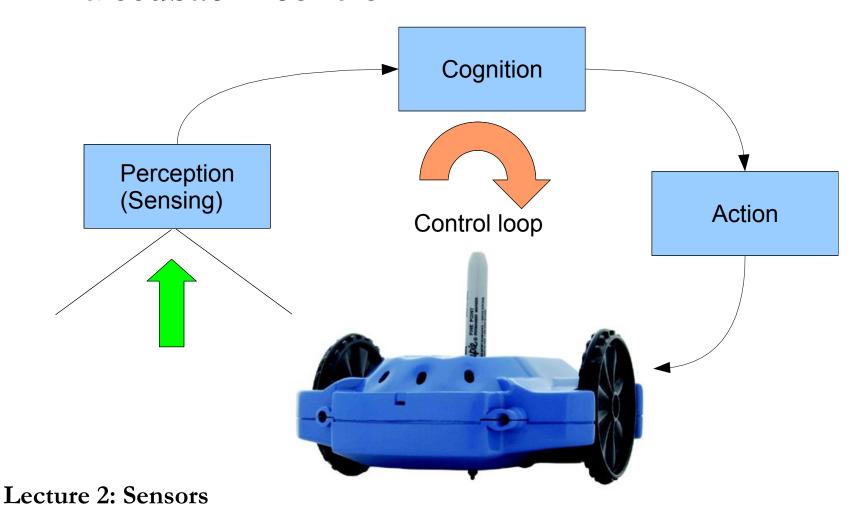
- To acquire information about the environment and oneself
- Open loop control suffers from
 - Uncertainty, changes in the world
 - Error detection and correction





The Sensing Loop

• "Feedback" control



5

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)?

6

Comparison: Human Sensors

Sense:

- Vision
- Audition
- Gustation
- Olfaction
- Tactition

Sensor:



















Robot Sensors

Sense:

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

Sensor:





- Magnetometer
- Voltage sensor
- Sonar

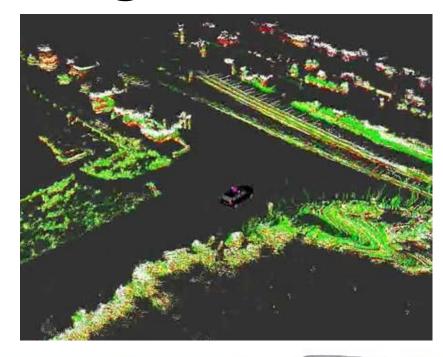
Array of pressure sensors



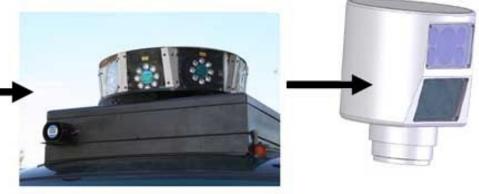


LiDar Sensing









LiDar Variations

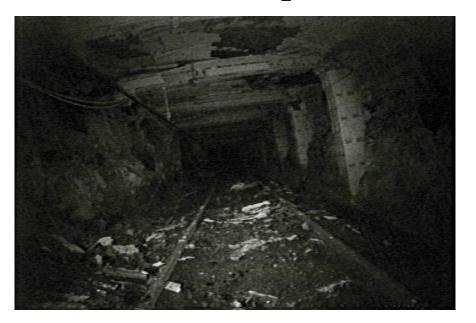
Tartan Racing Team



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

















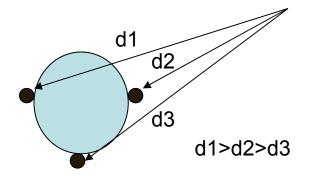


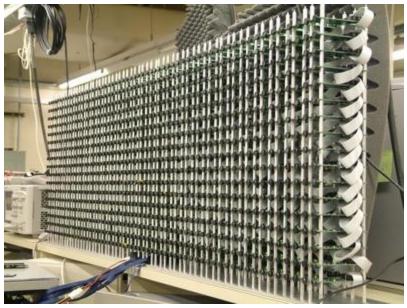




Auditory







Other Robot Sensors











Resistive Bend

Lever Switch



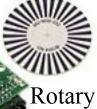
Piezo Bend



Pendulum Resistive Tilt



Pyroelectric Detector



Encoder



Gas



Accelerometer



UV Detector



Radiation





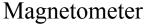
CDS Cell



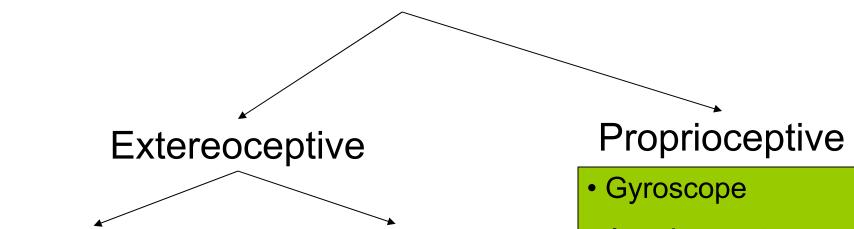
Compass



Magnetic Reed Switch



Sensing Classification



Laser/LiDar

Active

- Sonar
- Radar
- Structured light
- InfraRed

Passive

Vision

- Microphone array
- Chemical sensors
- Tactile sensor

- Accelerometers
- Odometers
- Voltage sensors
- Stress/strain gauge

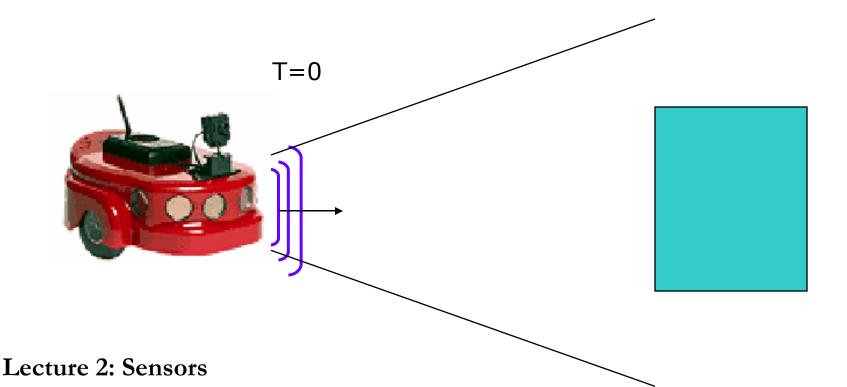
Sensors We Will Look At Today

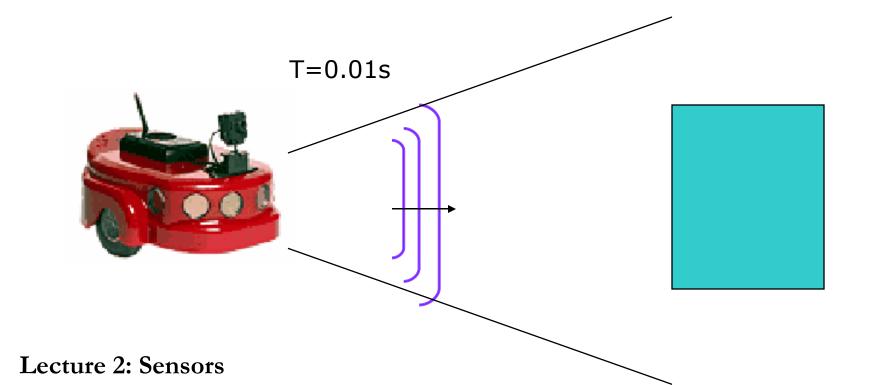
- Exterioceptive
 - Sonar, LiDar, IR
- Proprioceptive
 - Encoders
 - Accelerometers
 - Gyroscopes
 - GPS (hard to categorize)
 - Micro-switch

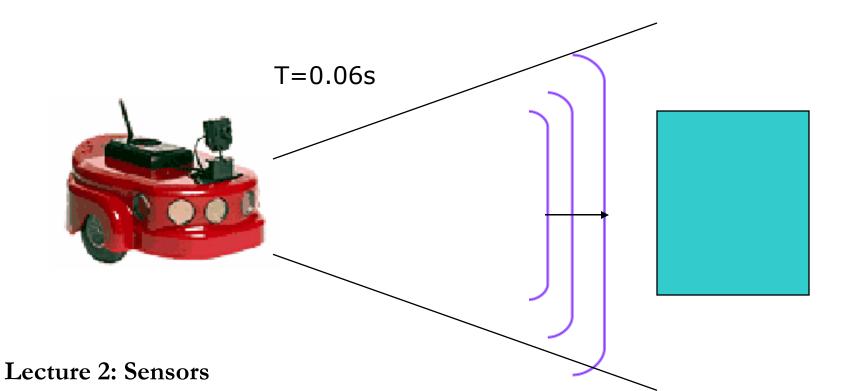
16

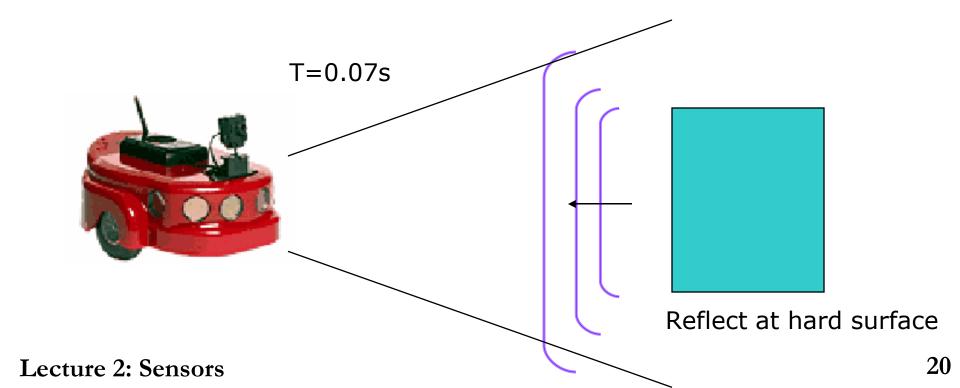
SoNaR: Sound Navigation and Ranging

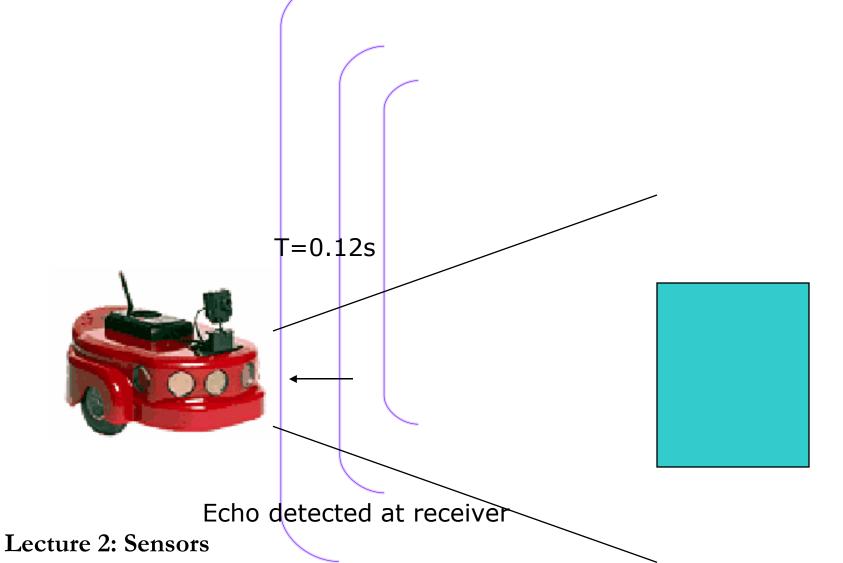
- Often called sonar, ultrasound, Sodar
- Emit a directional sound wave, and listen for echo(s), time the response



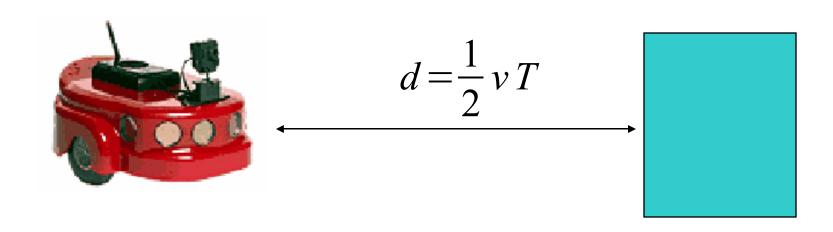






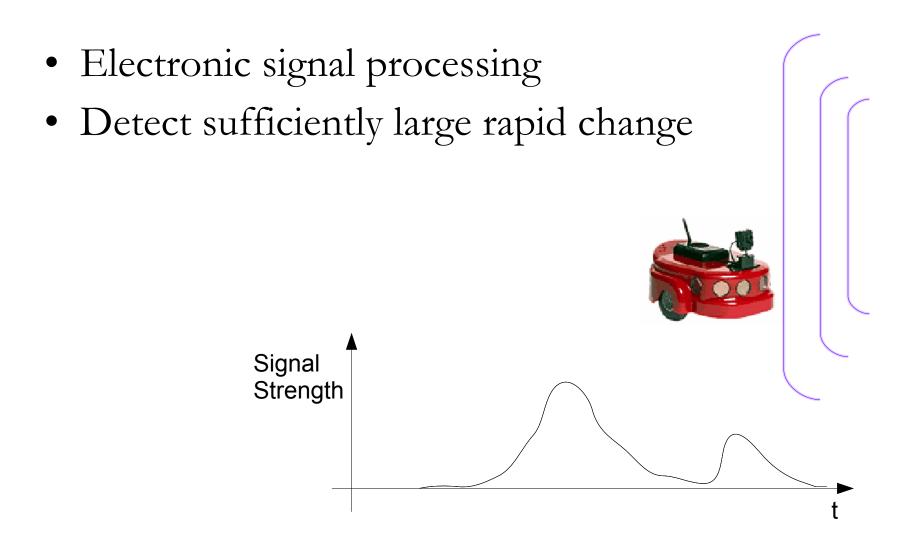


- Key assumption: sound travels at constant speed
- v=344 m/s (dry air, 21C, sea-level)
- So we have



Lecture 2: Sensors

How To Detect the Echo?



Lecture 2: Sensors

Imperfect Sensing

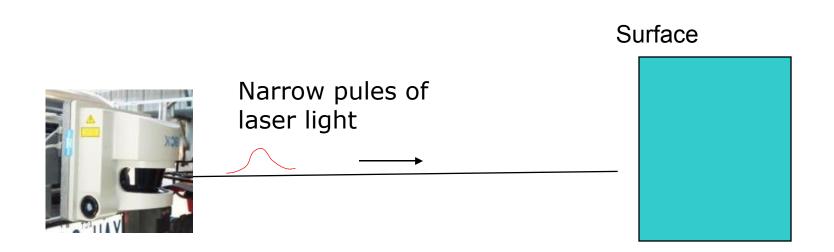
- What can go wrong?
 - Speed of sound changes with temperature, pressure, humidity

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

- Surface reflection properties
- Atmospheric attenuation (finite range)
- Multiple echos (multi-path)
- Quantization in timing
- Inaccuracies in detecting response signal onset

LiDar

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

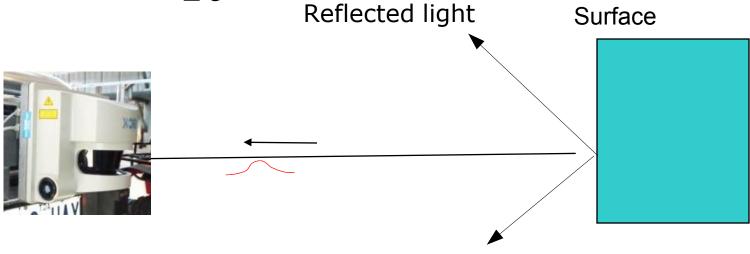


LiDar

• Timed "echo" from reflection

$$c_{vaccum} = \frac{1}{\sqrt{\mu_0 \epsilon_0}} \approx 3 \times 10^8 \, m.s^{-1}$$

$$d = \frac{T}{2c}$$



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 in range

Spinning mirror

SICK LiDar Internals

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









LiDar Variations

NREC Crusher Vehicle

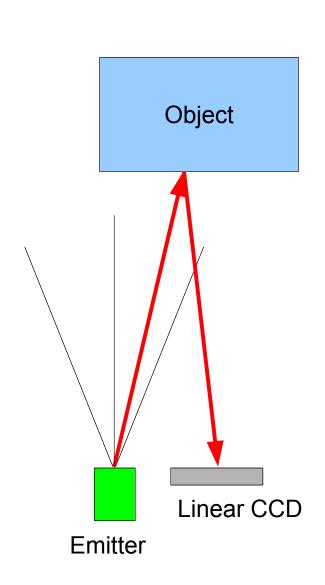


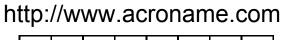
Crusher with sensors

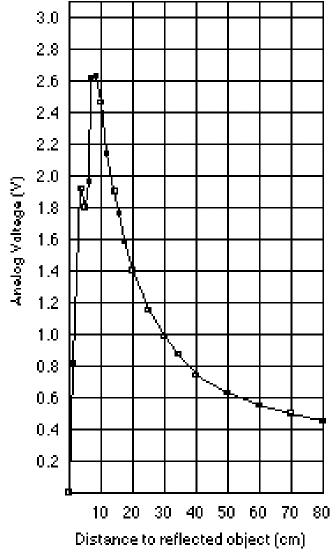
InfraRed

- 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



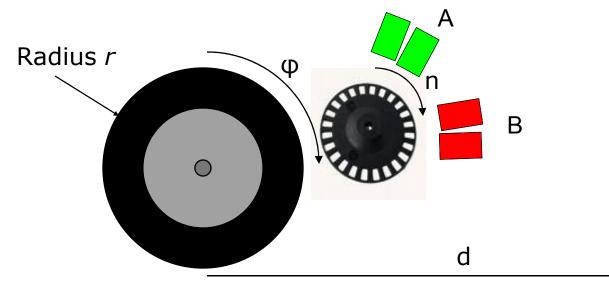




Proprioceptive Sensors

Optical Encoders

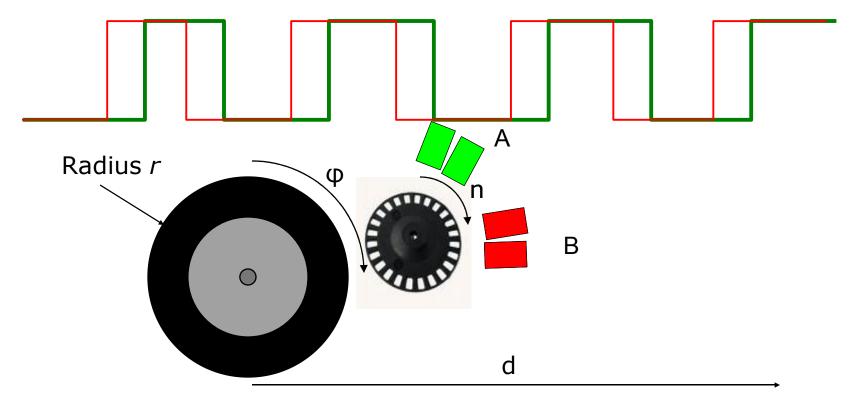
- Disc to measure *rotational* motion
- Out of phase IR emitter/detector pair



33

Optical Encoders

Direction and amount of rotation from edge transitions



Lecture 2: Sensors

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)

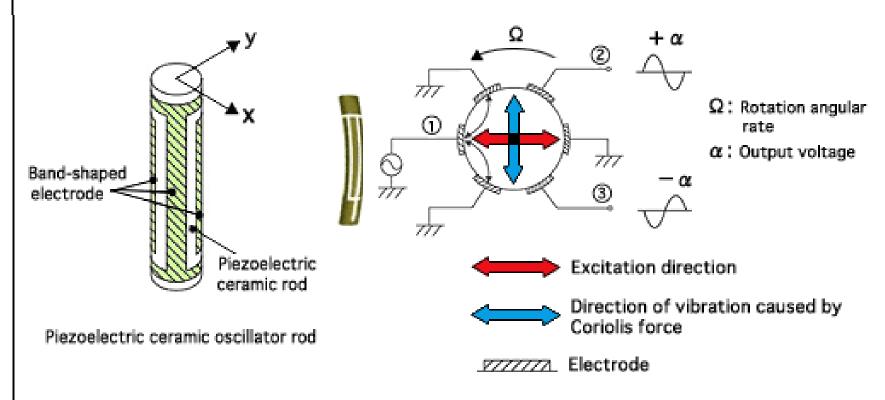
Lecture 2: Sensors

Gyroscopes

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

Rotational Angular Velocity Sensor

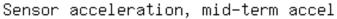
Operation principle: An angular velocity sensor that works by using the phenomenon generated by Coriolis force when angular velocity is applied to a moving object in relation to velocity and orthogonal directions.

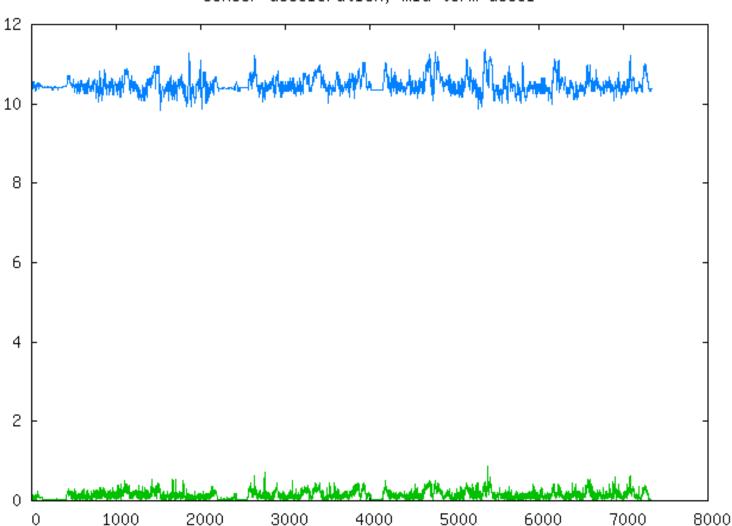


Accelerometers

- 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





Lecture 2: Sensors

Issues With Accelerometers/Gyros

Noise

Output readings may have approximately additive
Gaussian noise

• Drift

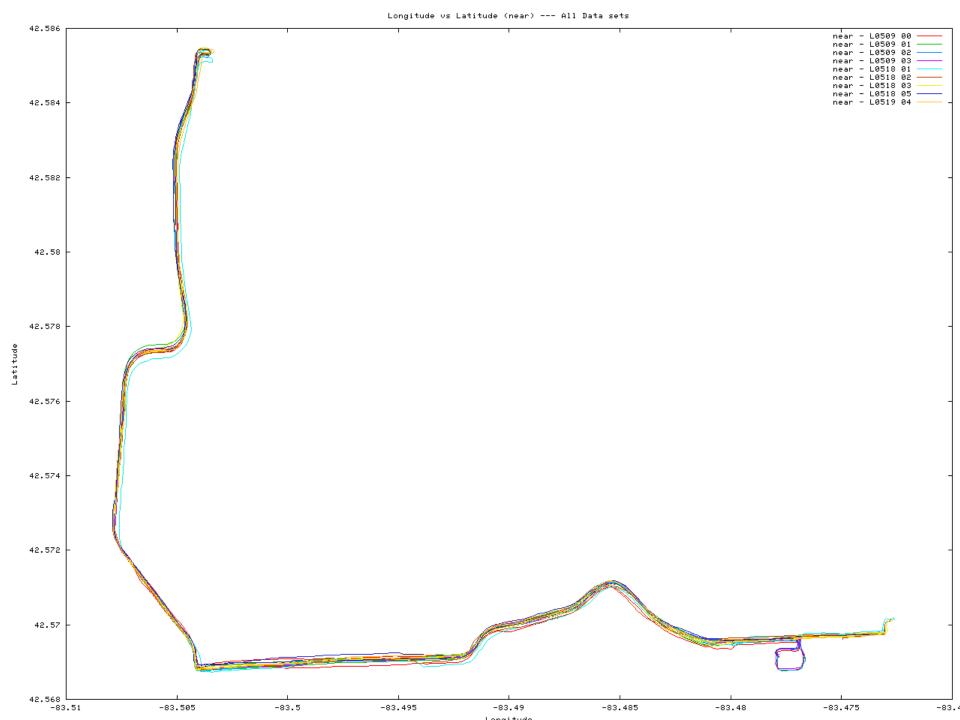
- Signal drifts from true value over time Gyro heading
- Usually need to integrate accelerometers

GPS/Glonas/Galileo

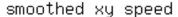
- Orbiting satellites
 - Known trajectories
 - Highly precise timers
- Transmit data in Ghz band
 - Ephemeris information
 - Develop pseudo range to satellite
- Solve for receiver position
- Can also solve for velocity

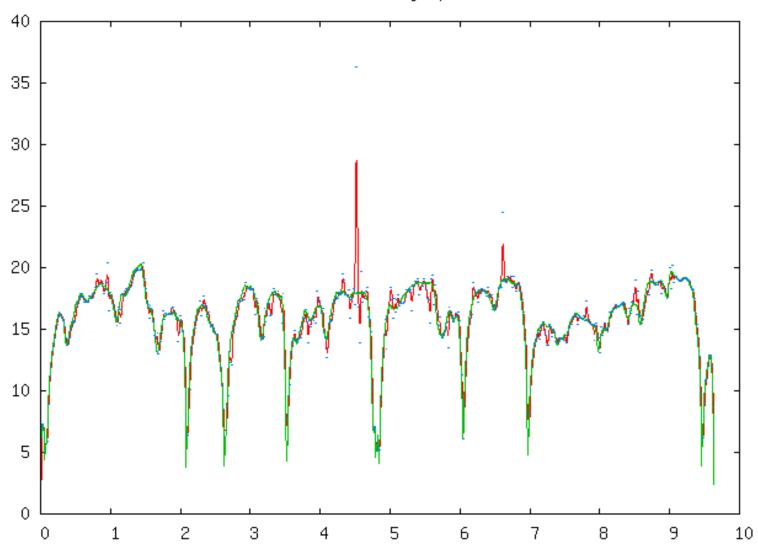






Ground Speed Profile smoothed xy speed





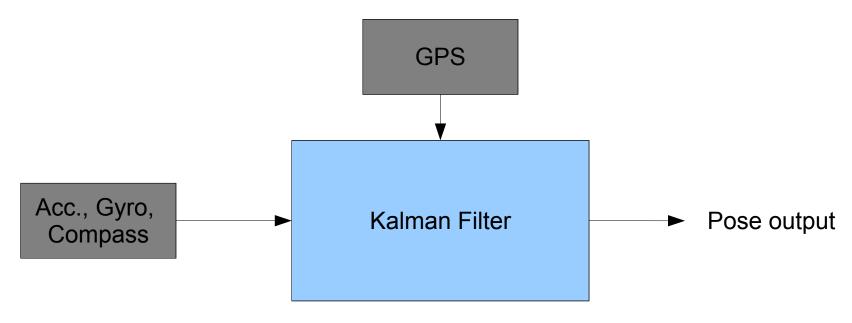
Lecture 2: Sensors

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)

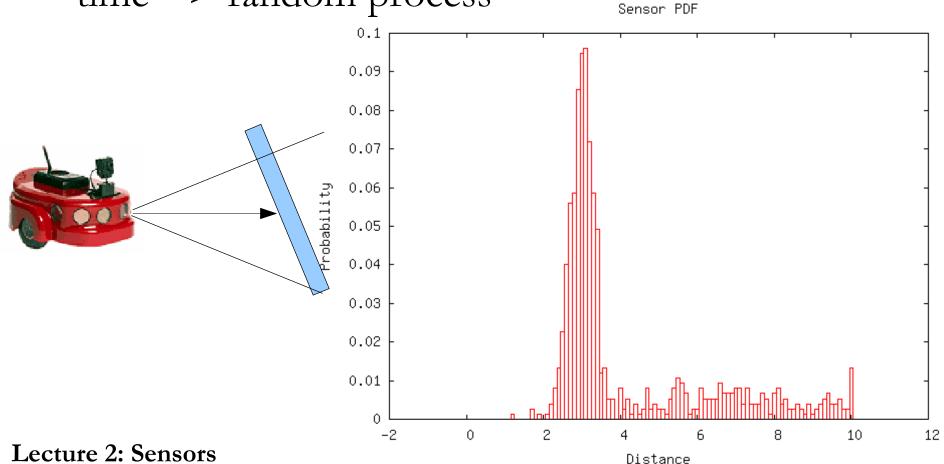


Sensor Processing and Modeling

- Sensors are never perfect
- Noise
- Systematic errors (bias)
- Drift, jumps
- Unmodeled artifacts (looks like bias)

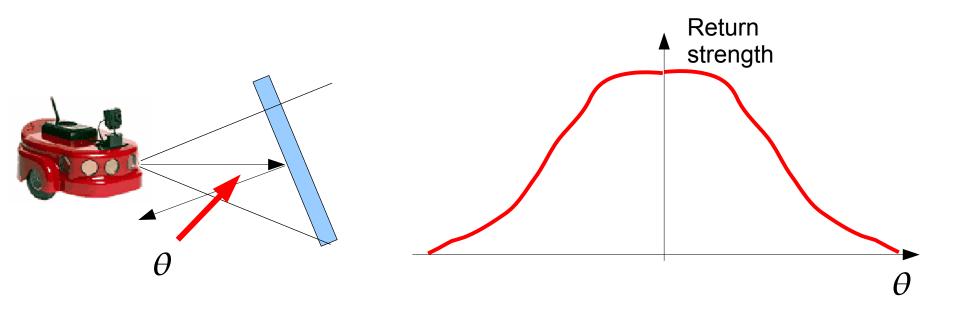
Sensor Noise

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

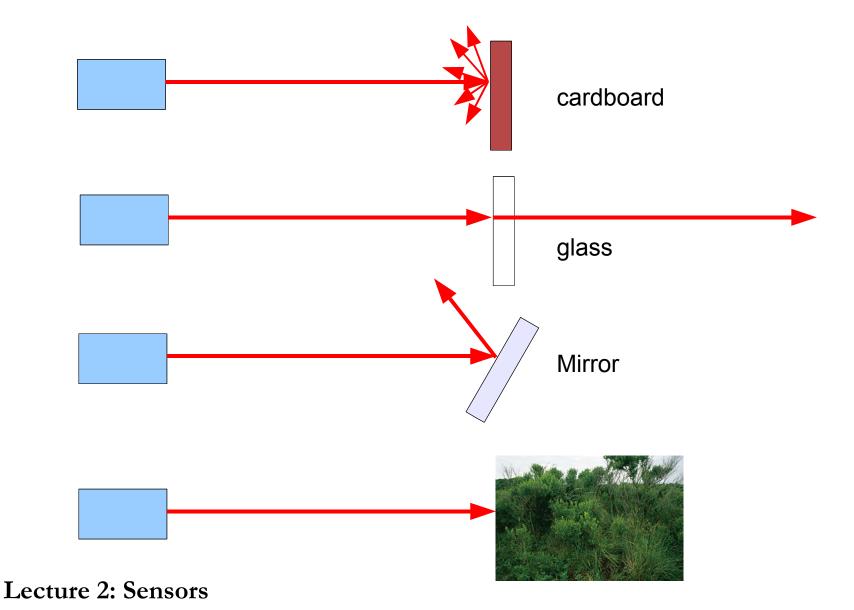


Sensor Bias

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



LiDar Returns and Material



Colorized LiDar

http://www.aerotecusa.com/



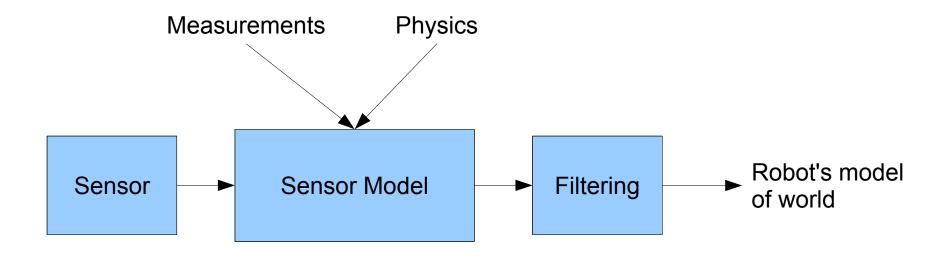
Lecture 2: Sensors

Helicopter Example

http://robots.stanford.edu

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
- 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

Sensor Model Components

- Intrinsic model
 - Model of how sensor operates
- Extrinsic model (most sensors)
 - Putting measurements into robot coordinate frame

An Example

- IR range sensor: triangulated range
- Intrinsic model
 - Given a return, what does it mean for depth?
- Extrinsic model
 - Given depth, where is the obstacle?
- Process
 - Sample data from known "standard" configuration
 - Estimate model/parameters from the data/physics
 - Test the result

Sensor Uncertainty

- Systematic errors (material, orientation etc.)
- Random errors (any other unknowns)
- Combine into uncertainty model
- Uncertainty model
 - Full pdf (e.g. Particles, histograms)
 - Parametric representation e.g. Gaussian
- Need to use probabilistic algorithms!

Filtering Approaches

- Often useful to apply initial filtering to signal
 - Use known constraints: domain knowledge
- Thresholding
 - Hystersis
 - Adaptive thresholds (hard)
- Smoothing
 - Linear filter
 - Kalman filter (later)
- Outlier rejection

Summary

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

Icreate and Scribbler

- ICreate
 - IR sensors
 - Bump sensor
 - Encoders
- Scribbler
 - Encoders
 - IR sensors