# CMPE 185 Autonomous Mobile Robots

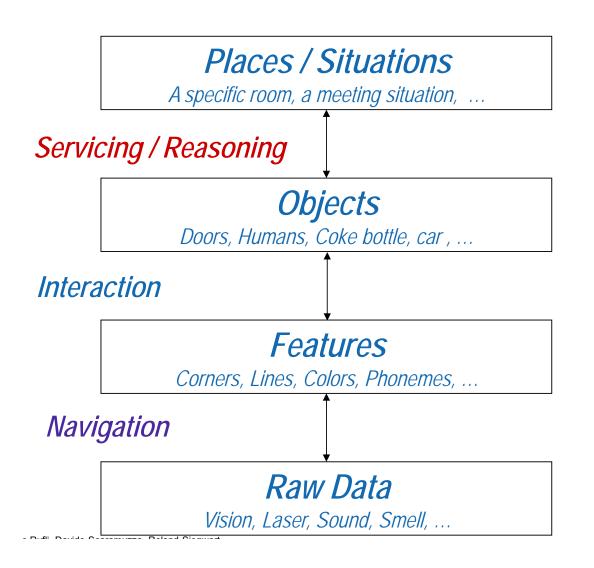
Perception, Sensor Characteristics and Uncertainty

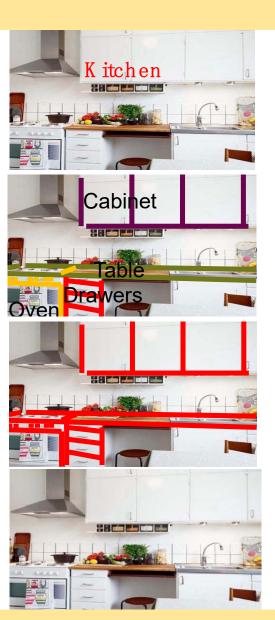
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# Group Project

 Reminder: check the group project assignments on Canvas

# What is Perception of Mobile Robots?





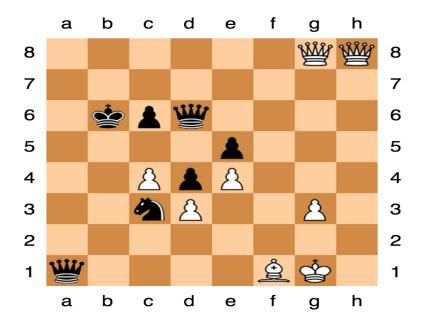
# Perception is Hard!

- Understanding = raw data + (probabilistic) models + context
  - Intelligent systems interpret raw data according to probabilistic models and using contextual information that gives meaning to the data.
- Dealing with real-world situations
- Reasoning about a situation
- Cognitive systems have to interpret situations based on uncertain and only partially available information
- They need ways to learn functional and contextual information (semantics / understanding)

# Perception is Hard!

- "In robotics, the easy problems are hard and the hard problems are easy"
  - S. Pinker. The Language Instinct. New York: Harper Perennial Modern Classics, 1994

beating the world's chess master: EASY

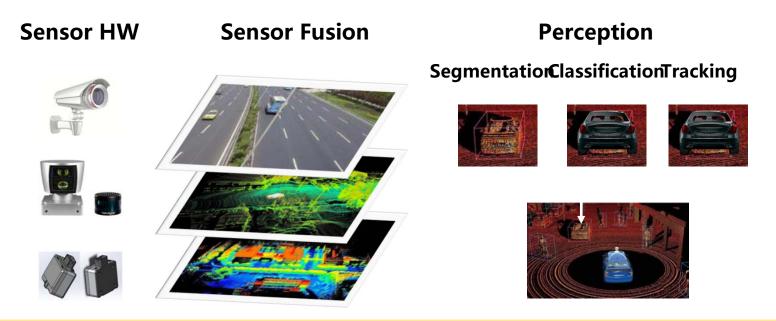


create a machine with some "common sense": very HARD



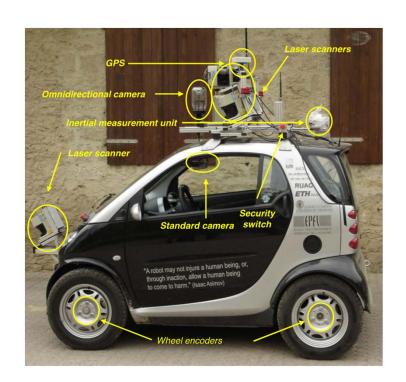
# Example: Perception for a Self-driving Car

- Any information a self-driving car collects about itself or its environment requires sensing
- The self-driving cars that want to learn, map and/or navigate need to collect information about their surroundings
- All sensors have some degree of uncertainty
- Uncertainty can be reduced by multiple measurements.



#### Common Sensors for Mobile Robots

- Wheel encoders
  - Local motion estimation (odometry)
- GPS
  - Global localization and navigation
- Inertial Measurement Unit (IMU)
  - Orientation and acceleration of the robot
- Laser scanners, Radar, Ultrasonic sensors
  - Obstacle avoidance, motion estimation, scene interpretation (road detection, pedestrians)
- Cameras
  - Texture information, motion estimation, scene interpretation
- Bumper
- ...



#### Classification of Sensors

- Robot = sensors + actuators
- Sensors are the key components for perceiving the environment

#### What:

- Proprioceptive sensors
  - measure values internally to the system (robot)
  - e.g. motor speed, wheel load, heading of the robot, battery status
- Exteroceptive sensors
  - information from the robots environment
  - distances to objects, intensity of the ambient light, unique features.

#### How

- Passive sensors
  - Measure energy coming from the environment; very much influenced by the environment
- Active sensors
  - emit their proper energy and measure the reaction
  - better performance, but some influence on environment

# Classification of Sensors

Sensor type	Sensor System	Proprioceptive (PC) or Exteroceptive (EC)	Active or Passive
Tacticle sensors	Bumbers	EC	P
Wheel/motor sensors	Brush encoders	PC	P
	Optical encoders	PC	Α
Heading sensors	Compass	EC	P
	Gyroscope	PC	P
	Inclinometer	EC	A/P
Acceleration sensors	Accelerometer	PC	Р
Beacons	GPS	EC	Α
	Radio, ultrasonic, reflective beacons	EC	Α
Motion/speed sensors	Doppler: radar or sound	EC	Α
Range sensors	Ultrasound, laser rangefinder, structured light, time of flight	EC	Α
Vision sensors	CCD/CMOS cameras	EC	P

## Sensors

Sensor Characteristics

Sensor Uncertainty

• Different Sensors

- Range
  - Lower and upper limits
  - E.g., IR range sensor measures distance between 10 and 80cm
- Resolution
  - Minimum difference between two measurements
  - For digital sensors, it is usually the A/D resolution

e.g. 
$$\frac{5V}{255(8 \ bit)} = 0.02V$$

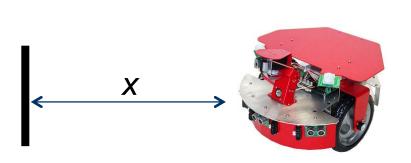
- Dynamic range
  - Used to measure spread between lower and upper limits of sensor inputs
  - Formally, it is the ratio between the maximum and minimum measurable input, usually in decibels (dB)

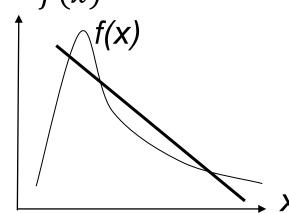
$$Dynamic\ Range = 10 \log \left[ \frac{UpperLimit}{LowerLimit} \right]$$

 E.g., A sonar range sensor measures up to a max distance of 3m, with smallest measurement of 1cm

$$Dynamic Range = 10 \log \left[\frac{3}{0.01}\right]$$
$$= 24.8dB$$

- Linearity
  - A measure of how linear the relationship between the sensor's output signal and input signal
  - Linearity is less important when signal is treated after with a computer
- Linearity example
  - Consider the range measurement from an IR range sensor
  - Let x be the actual measurement in meters, let y be the output from the sensor in volts, and y = f(x)





- Bandwidth or Frequency
  - The speed with when a sensor can provide a stream of readings
  - Usually there is an upper limit depending on the sensor and the sampling rate
    - o e.g., sonar takes a long time to get a return signal
  - Higher frequencies are desired for autonomous control
    - e.g., if a GPS measurement occurs at 1 Hz and the autonomous vehicle uses this to avoid other vehicles that are 1 meter away
- Sensitivity
  - Ratio of output change to input change
    - E.g., range sensor will increase voltage output 0.1 V for every cm distance measured

- Accuracy
  - The difference between the sensor's output and the true value (i.e., error = m v)

$$accuracy = 1 - \frac{|m - v|}{v}$$

- Precision
  - The reproducibility of sensor results

$$precision = \frac{range}{\sigma}$$

 $\sigma$  = standard deviation

#### Sensors: In Situ Characteristics

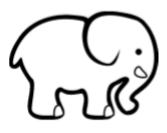
- Random Error
  - Non-deterministic
  - Not predictable
  - Usually described probabilistically





- Systematic Error
  - Deterministic
  - Caused by factors that can be modeled, e.g., optical distortion in camera





#### Sensors: In Situ Characteristics

- Measurements in the real-world are dynamically changing and error-prone
  - Changing illuminations
  - Light or sound absorbing surfaces
- Systematic versus random errors are not well-defined for mobile robots
  - There is a cross-sensitivity of robot sensor to robot pose and environment dynamics
  - Difficult to model, appear to be random

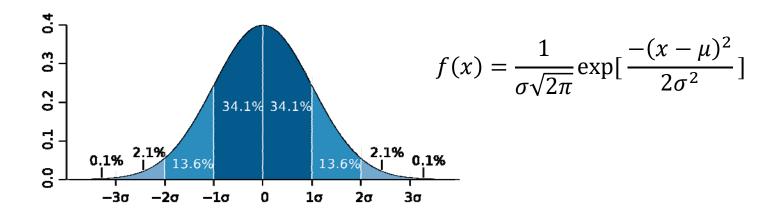
## Sensors

Sensor Characteristics

Sensor Uncertainty

• Different Sensors

- How can it be represented?
  - With probability distributions
- Representation
  - Describe measurement as a random variable X
  - Given a set of *n* measurements with values  $\rho_1$
  - Characterize statistical properties of X with a probability density function f(x)



• Expected value of X is the mean  $\mu$ 

$$\mu = E[X] = \int_{-\infty}^{\infty} x f(x) \, dx$$

• The variance of X is  $\sigma^2$ 

Expected value of X is the mean 
$$\mu$$

$$\mu = E[X] = \int_{-\infty}^{\infty} xf(x) dx$$

$$\mu = E[X] = \frac{\sum_{n=1}^{\infty} x}{n}$$

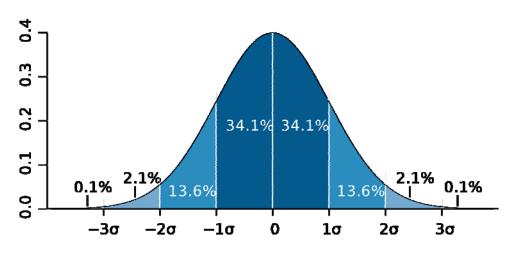
$$\mu = E[X] = \frac{\sum_{n=1}^{\infty} x}{n}$$

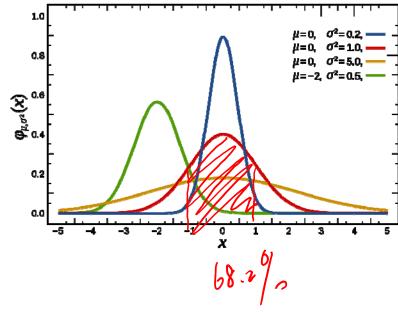
$$\mu = E[X] = \frac{\sum_{n=1}^{\infty} x}{n}$$

$$\sigma^{2} = Var(X) = \int_{-\infty}^{\infty} (x - \mu)^{2} f(x) dx \qquad \sigma^{2} = Var(X) = \frac{\sum^{n} (x - \mu)^{2}}{n}$$

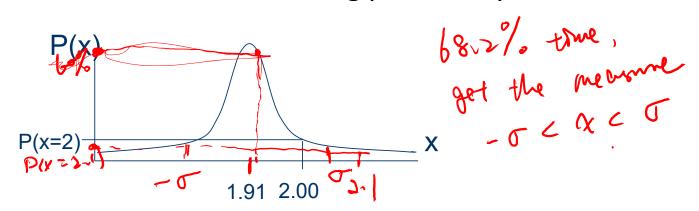
Use a Gaussian Distribution

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[\frac{-(x-\mu)^2}{2\sigma^2}\right]$$





- How do we use the Gaussian?
- Learn the variance of sensor measurements ahead of time
- Assume mean measurement is equal to actual measurement
- Example:
  - If a robot is 1.91 meters from a wall, what is the probability of getting a measurement of 2 meters?
  - Answer: if the sensor error is modeled as a Gaussian, we can assume the sensor has the following probability distribution



■ Then, use the distribution to determine P(x = 2)

• Thank you!