# Day 32

Range Sensor Models

#### Range Sensors

- range sensors measure the distance between the robot and the sensed object(s)
  - bearing measurement can be obtained by rotating the sensor or using multiple sensors arranged on a circular arc
- many non-vision based sensors can be modeled by measuring the distance along a beam or cone

#### Infra-red Range Sensor

 uses an IR LED and a linear CCD array to triangulate distances



Sharp GP2Y0A21YK0F data sheet

angle increases as distance increases

#### Infra-red Range Sensor

- has a minimum and maximum distance
  - changes depending on the model but range is on the order of a few centimeters to a few meters
- inexpensive (less than \$20)
- measurements affected by the material properties of the object
  - but less sensitive to the orientation of the reflecting surface
- Iow accuracy

- ultrasound is sound above the upper limit of human hearing (approximately 20,000 Hz)
- inexpensive ultrasonic transducers are essentially ceramic discs that vibrate at ultrasonic frequencies
  - operate via the piezoelectric effect
    - an applied external electric field causes a small change in the physical dimensions
    - conversely, an applied stress will induce a small electric field







- basic idea of operation is simple
  - transducer emits a short pulse of ultrasound
  - receiver listens for echo
  - time elapsed between pulse and echo is proportional to twice the distance to the object



speed of sound in air is given approximately by

 $c = 331 + 0.6T \frac{\text{m}}{\text{s}}$ 

where T is temperature in Celsius

- humidity affects c
- modern electronics can easily measure elapsed time with sufficient precision to obtain good distance resolution

most materials are specular reflectors of ultrasound



Figure 3.-Angle of incidence. (I = angle of incidence, R = angle of reflection.)

 multiple reflections produce erroneous distance measurements



Figure 4.—Multiple reflections causing false range reading.

 unfocussed ultrasound transducers have a wide beam pattern with significant side lobes



Figure 2: Typical propagation pattern for the Polaroid 6500 Series ultrasonic sensor. Courtesy of Polaroid [1].

Cao and Borenstein, "Experimental Characterization of Polaroid Ultrasonic Sensors in Single and Phased Array Configuration"

beamwidth + side lobes leads to spurious measurements



Figure 4: Baseline results for measuring distance to (a) the board and (b) the pole placed at different distances and with the sonar paned to different angles between 0 and 90°.

Cao and Borenstein, "Experimental Characterization of Polaroid Ultrasonic Sensors in Single and Phased Array Configuration"

## LIDAR

- light detection and ranging
- <u>commercial range finder</u>
   <u>disassembled</u>



#### LIDAR

for time of flight sensor d = ½ c∆t
 speed of light c = 3×10<sup>8</sup> m/s

1 ns = 150 mm 100 ps = 15 mm10 ps = 1.5 mm

- typical angular resolution below 1 degree
- typical errors on the order of 10 cm or less
- cost > \$1,000

 given the map and the robot's location, find the probability density that the range finder detects an object at a distance z<sup>k</sup><sub>t</sub> along a beam

 $p(z_t^k | x_t, m)$ 

the k is here because range sensors typically return many measurements at one time; e.g., the sensor might return a full 360 degree scan made up of K measurements at once

$$z_{t} = \left\{ z_{t}^{1}, z_{t}^{2}, ..., z_{t}^{k}, ..., z_{t}^{K} \right\}$$



- most range finders have a minimum and maximum range
- we seek a model that can represent
  - 1. the correct range measurement with noise
  - 2. unexpected obstacles
  - 3. failures
  - 4. random measurements



**Figure 6.4** "Pseudo-density" of a typical mixture distribution  $p(z_t^k \mid x_t, m)$ .

- I. Correct Measurement with Noise
- suppose that the beam intersects an obstacle at a range of  $z_t^{k^*}$ 
  - model this measurement as a narrow Gaussian with mean  $z_t^{k^*}$

$$p_{\text{hit}}(z_t^k | x_t, m) = \begin{cases} \eta N(z_t^k; z_t^{k^*}, \sigma_{\text{hit}}^2) & \text{if } 0 \le z_t^k \le z_{\text{max}} \\ 0 & \text{otherwise} \end{cases}$$

(a) Gaussian distribution  $p_{\rm hit}$ 



- 2. Unexpected Obstacles
- in a dynamic environment there will be obstacles not found on a static map; these obstacles will cause range measurements shorter than expected
- if such obstacles appear continuously and independently at a constant average rate then they can be modeled as a Poisson process
- the time between events in a Poisson process has an exponential distribution with probability density function

$$f(x;\lambda) = \begin{cases} \lambda e^{-\lambda x}, & x \ge 0, \\ 0, & x < 0. \end{cases}$$

$$p_{\text{short}}(z_t^k | x_t, m) = \begin{cases} \eta \,\lambda_{\text{short}} e^{-\lambda_{\text{short}} z_t^k} & \text{if } 0 \le z_t^k \le z_t^{k^*} \\ 0 & \text{otherwise} \end{cases}$$

(b) Exponential distribution  $p_{\rm short}$ 



- 3. Failures
- > range finders can fail to sense an obstacle in which case most sensors return  $z_{\rm max}$
- modeled as a point mass distribution centered at  $z_{\text{max}}$

$$p_{\max}(z_t^k | x_t, m) = \begin{cases} 1 & \text{if } z_t^k = z_{\max} \\ 0 & \text{otherwise} \end{cases}$$

(c) Uniform distribution  $p_{\rm max}$ 

$$\begin{array}{c|c} p(z_t^k \mid x_t, m) \\ \hline \\ \hline \\ \hline \\ z_t^{k*} & z_{\max} \end{array}$$

- 4. Random Measurements
- unexplainable measurements are modeled as a uniform distribution over the range  $[0, z_{max}]$

$$p_{\text{rand}}(z_t^k | x_t, m) = \begin{cases} \frac{1}{z_{\text{max}}} & \text{if } 0 \le z_t^k \le z_{\text{max}} \\ 0 & \text{otherwise} \end{cases} \quad \begin{array}{c} \text{(d) Uniform distribution } p_{\text{rand}} \\ p(z_t^k | x_t, m) \end{cases}$$

 $z_{\rm max}$ 

 $z_{t}^{k*}$ 

the complete model is a weighted sum of the previous four densities with weights

$$w_{\text{hit}} + w_{\text{short}} + w_{\text{max}} + w_{\text{rand}} = 1$$

$$p(z_t^k | x_t, m) = \begin{bmatrix} w_{\text{hit}} \\ w_{\text{short}} \\ w_{\text{short}} \\ w_{\text{max}} \\ w_{\text{rand}} \end{bmatrix}^T \begin{bmatrix} p_{\text{hit}}(z_t^k | x_t, m) \\ p_{\text{short}}(z_t^k | x_t, m) \\ p_{\text{max}}(z_t^k | x_t, m) \\ p_{\text{rand}}(z_t^k | x_t, m) \end{bmatrix}$$