CSE-571 Probabilistic Robotics

Mapping

Types of SLAM-Problems









[Lu & Milios, 97; Gutmann, 98: Thrun 98; Burgard, 99; Konolige & Gutmann, 00; Thrun, 00; Arras, 99; Haehnel, 01;...]





[Leonard et al., 98; Castelanos et al., 99: Dissanayake et al., 2001; Montemerlo et al., 2002;...

Problems in Mapping

- Sensor interpretation
 - How do we extract relevant information from raw sensor data?
 - How do we represent and integrate this information over time?
- Robot locations have to be known
 - How can we estimate them during mapping?

Occupancy Grid Maps

- Introduced by Moravec and Elfes in 1985
- Represent environment by a grid.
- Estimate the probability that a location is occupied by an obstacle.
- Key assumptions
 - Occupancy of individual cells is independent

$$Bel(m_t) = P(m_t | u_1, z_2 ..., u_{t-1}, z_t)$$
$$= \prod_{x, y} Bel(m_t^{[xy]})$$

Robot positions are known!

Updating Occupancy Grid Maps

Idea: Update each individual cell using a binary Bayes filter.

$$Bel(m_t^{[xy]}) = \eta \ p(z_t \mid m_t^{[xy]}) \sum_{m_{t-1}^{[xy]}} p(m_t^{[xy]} \mid m_{t-1}^{[xy]}, u_{t-1}) Bel(m_{t-1}^{[xy]})$$

Additional assumption: Map is static.

$$Bel(m_t^{[xy]}) = \eta \ p(z_t \mid m_t^{[xy]}) Bel(m_{t-1}^{[xy]})$$

Inverse Sensor Model for Occupancy Grid Maps

Combination of linear function and Gaussian:



 $\overline{B}(m_t^{[xy]}) = \log odds(m_t^{[xy]} | z_t, x_t) - \log odds(m_t^{[xy]}) + \overline{B}(m_{t-1}^{[xy]})$

Incremental Updating of Occupancy Grids (Example)



Alternative: Simple Counting

- For every cell count
 - hits(x,y): number of cases where a beam ended at <x,y>
 - misses(x,y): number of cases where a beam passed through <x,y>

$$Bel(m^{[xy]}) = \frac{hits(x, y)}{hits(x, y) + misses(x, y)}$$

• Assumption: P(occupied(x,y)) = P(reflects(x,y))

Resulting Map Obtained with Ultrasound Sensors





Occupancy Grids: From scans to maps







Tech Museum, San Jose





occupancy grid map

CAD map