

# Computational Principles of Mobile Robotics

Post maintenance and localization

“You are in a maze of twisty passages all alike”

## Pose maintenance

- It may be useful or necessary to know where a robot is relative to some spatial representation (a map).
- Often expressed computing  $P(q|s)$  (probability that the robot is in state  $q$  given measurements  $s$ ) and then choosing a maximum value of this.

## Pose maintenance and localization

- Can distinguish between maintaining pose (good prior estimate) versus localization (no prior estimate).
- Can distinguish between global localization (where is here) versus local localization (is here different from other locations).

## 9.1 Simple landmark measurement

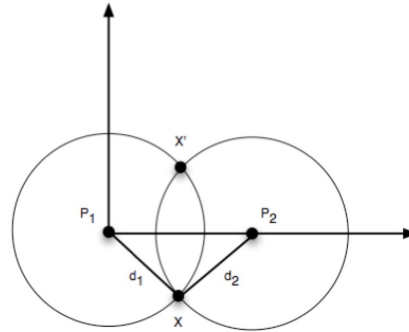
- Problem often involves determining if specific landmarks in the environment map to landmarks in the map and then using geometry to constrain the state of the vehicle.
  - Landmark properties influence the task (how easy are they to find, uniqueness, geometry properties).

## 9.1.1 Landmark classes

- Active versus passive
- Sensing modality, examples and properties
  - Vision -> generally implies bearing only relative information
  - Laser -> relative pose
  - GPS -> absolute position

## 9.1.2 Triangulation and trilateration

- Triangulation – use of bearings (angles) to determine geometry.
- Trilateration – use of distances to determine geometry.
  - Often incorrectly referred to as triangulation.



Robot senses  $P_1$  and  $P_2$  and estimates  $d_1$  and  $d_2$ . Robot is either at  $x$  or  $x'$ .

## Triangulation with uncertainty

- Perfect geometry is complicated in the real world as bearing and distance measures are corrupted with noise.
- Characterize the error in pose estimation as the **geometric dilution of precision (GDOP)**. Given state estimate  $X$  and sensor readings  $S$

$$\text{GDOP} = \frac{\Delta X}{\Delta S}$$

In the limit, this becomes the Jacobian

## 9.2 Fingerprinting

- For a given sensor we can compute  $F(\text{pose}) = \text{measurements}$ .
- We can then seek to find the inverse of this that takes measurements and obtains the pose directly.
- The computational task involves modelling the inverse function.
  - DNN's have found good successes here.
- Good performance for radio-based mechanisms (e.g., localization from base station signal properties).



## 9.3 Servo control

- Given a function  $I(q)$  that describes the sensor reading as a function of pose, then we can perform servo control (homing) by comparing  $I(q)$  to  $I(q_{\text{goal}})$ .
- If  $I(q)$  is smooth, then this can be very effective, especially for  $q$  near  $q_{\text{goal}}$ .

## 9.4 Recursive filtering

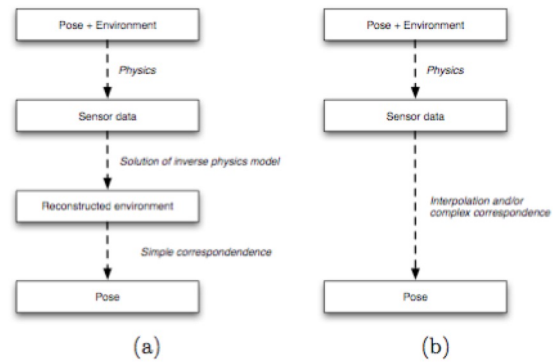
- Standard recursive filtering techniques (e.g., EKF, particle filters) can be applied to the problem of ongoing state estimation.
  - Covered in some detail earlier.

## 9.5 Localization in ROS

- Navigation 2 stack provides implementation of standard tools for mapping and localization.
- Localization is provided via an adaptive Monte Carlo localization process.

## 9.6 Non-geometric methods: perceptual structure

- Assign to each location a **signature** that captures the sensor information.
- Construct a direct mapping between signature and location.



Pose estimation using geometry methods (a) and perceptual structure (b).

## 9.6.1 Eigenlandmarks

- Rather than having specific pre-defined landmarks, we can instead assume that our landmarks have some inherent structure that we can use.
  - E.g., pipe junctions, etc.
- Build an Eigenspace representation of the landmarks and use this lower dimensionality space to represent and recognize landmarks.

## 9.7 Correlation-based localization

- Occupancy-grid maps and local occupancy-grid measurements can be compared using correlation-based approaches.
- Let  $P(z|s)$  be the probability that a location  $z$  is occupied given measurements  $s$ .
  - This can be correlated against the map to identify likely robot states.

## 9.8 Global localization

- If no good prior for the robot's pose exists then all potential states are equally likely (kidnapped robot problem).
- Global localization is very difficult, even in very simple 2D polygon worlds.
  - For general environments the problem is NP-hard.

## 9.9 Biological approaches to localization

- Biological systems exist that do localization very well.
  - Examples include songbirds, ants, others.
- Some evidence of multi-cue integration processes in some species.