
Mapping

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A Robot's Navigation Problems

- Where am I? *Localization*
 - Where have I been? *Map making*
 - Where am I going? *Mission planning*
 - What's the best way there? *Path planning*
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Why Mapping?

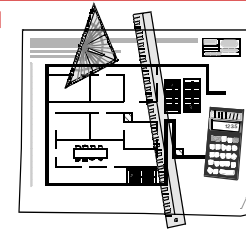
- Learning maps is one of the fundamental problems in mobile robotics
- Maps allow robots to efficiently carry out their tasks, allow localization ...
- Successful robot systems rely on maps for localization, path planning, activity planning etc.

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How to Form a Map

1. By Hand



2. Automatically: **Map Building**

The robot **learns** its environment

Motivation:

- by hand: hard and costly
- dynamically changing environment
- different look due to different perception

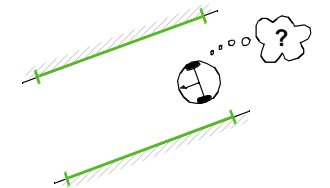
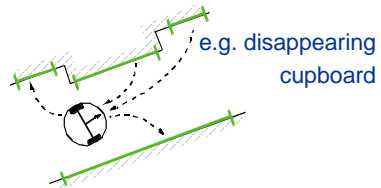
3. Basic Requirements of a Map:

- a way to incorporate **newly sensed** information into the existing world model
- information and procedures for **estimating the robot's position**
- information to do **path planning** and other **navigation tasks** (e.g. obstacle avoidance)
- Measure of Quality of a map
 - topological correctness
 - metrical correctness
- But: Most environments are a mixture of **predictable** and **unpredictable** features
→ hybrid approach
model-based vs. behaviour-based

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The Problems

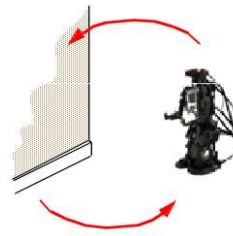
1. Map Maintaining: Keeping track of changes in the environment



- e.g. measure of belief of each environment feature

2. Representation and Reduction of Uncertainty

position of robot \rightarrow position of wall



position of wall \rightarrow position of robot

- probability densities for feature positions
- additional exploration strategies

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Mapping

■ Things to consider:

- Map precision must match application
- Precision of features on map must match precision of the robot's sensory data
- Map complexity directly affects computational complexity, and reasoning about localization and navigation

■ Two basic approaches

- Continuous
- Decomposition (discretization)

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Representation of the Environment

■ Environment Representation

- Continuous Metric \rightarrow coordinates and heading (x, y, θ)
- Discrete Metric \rightarrow metric grid
- Discrete Topological \rightarrow topological grid

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Environment Modeling

■ Raw sensor data, e.g. laser range data, images

- large volume of data, low distinctiveness on the level of individual values
- makes use of all acquired information

■ Low level features, e.g. lines and other geometric features

- medium volume of data, average distinctiveness
- filters out the useful information, still ambiguities

■ High level features, e.g. doors, a car, a monument

- low volume of data, high distinctiveness
- filters out the useful information, few/no ambiguities, not enough information

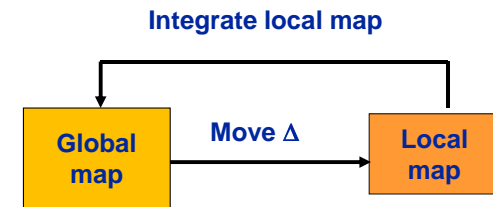
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Representing the robot

- How to represent the robot itself on a map?
- Point-robot assumption
 - Represent the robot as a point
 - Assume it is capable of omnidirectional motion
- Robot in reality is of nonzero size
 - Dilation of obstacles by robot's radius
 - Resulting objects are approximations
 - Leads to problems with obstacle avoidance

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Basic Idea



- Sense and create a local map
- Move a little
 - Record change in position, orientation
- Sense and create a local map
 - Fuse/tile together

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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 estimated
 - How can we identify that we are at a **previously visited place**?
 - This problem is the so-called **data association problem**.

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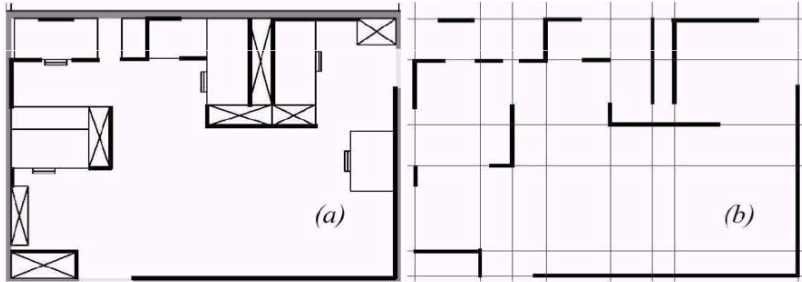
Continuous representation

- Exact decomposition of the environment
- Closed-world assumption
 - Map models all objects
 - Any area of map without objects has no objects in the corresponding environment
 - Map storage proportional to the density of objects in the environment
- Map abstraction and selective capture of features to ease computational complexity

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Continuous representation

- Match map type with a sensing device
 - In case of laser ranger finder, may represent the map as a series of infinite lines
 - Fairly easy to fit laser range data to series of lines



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Continuous representation

- In conjunction with position representation maintain:
 - **Single hypothesis:** extremely high accuracy possible
 - **Multiple hypothesis:**
 - Either, depict as geometric shape
 - Or, as a discrete set of possible positions
- Advantages of continuous representation
 - High accuracy possible
- Disadvantages
 - Can be computationally expensive
 - Typically only 2D

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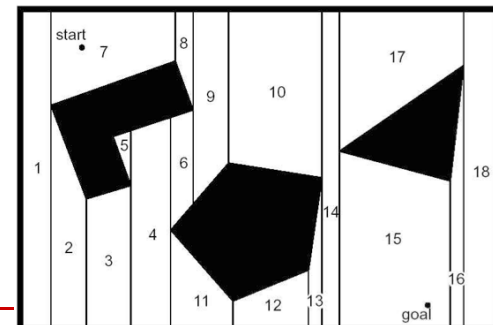
Decomposition Approach

- Capture only the useful features of the environment
- Computationally better for planning, particularly if the map is hierarchical

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Exact cell decomposition

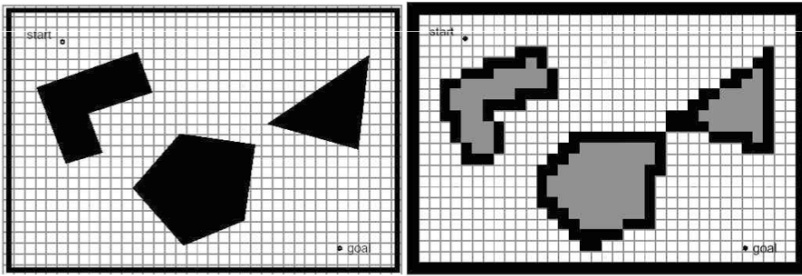
- Model empty areas with geometrical shapes, e.g. trapezoids
- Can be extremely compact (18 nodes in this figure)
- Assumption: the robot position within each area of free space is not important



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Fixed cell decomposition

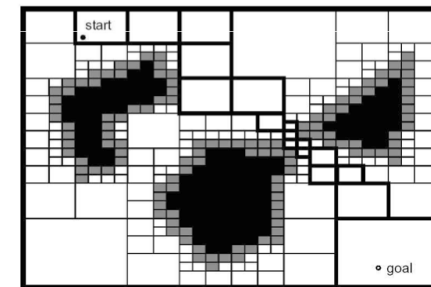
- Tessellate world: use a discrete approximation
- Each cell is either empty or full
- Inexact (note loss of narrow passageway on right)



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Adaptive cell decomposition

- Multiple types of adaptation: quadtree, exact, etc.
- Recursively decompose until a cell is completely free or completely an object
- Very space efficient compared to the fixed cell approach



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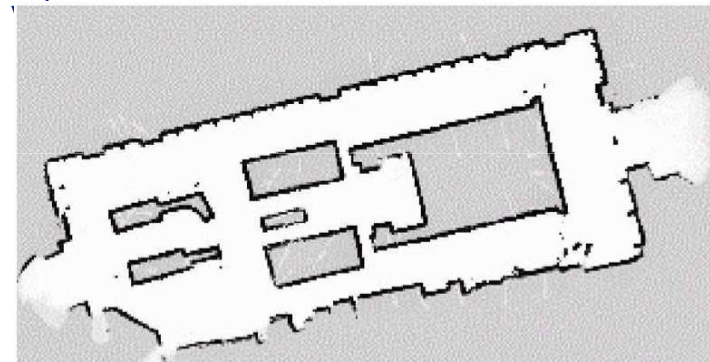
Occupancy Grid Representation

- Typically fixed cell decomposition
- Each cell is either filled or free
 - Counter for cell: zero indicates a free cell, above a certain threshold the cell is considered to be filled with an object
- Particularly useful with range-based sensors
 - If the sensor strikes something in a cell, increase cell counter
 - If sensor goes over cell and strikes something else, decrease cell counter (presuming it is free space)
 - By also discounting cell values over time, can deal with transient obstacles
- Disadvantages
 - Map size a function of the size of environment and size of cell
 - Imposes an a priori geometric grid on the world

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Occupancy Grid Example

- Shade of grey of cell proportional to the cell counter



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Occupancy Grid Map updates

- Occupancy grid representation
 - Each cell indicates probability is free space and probability is occupied
 - Need method to update cell probabilities given sensor readings at time t
- Update methods
 - Sensor model
 - Bayesian
 - Dempster-Shafer

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Topological Decomposition Approach

- Use environment features most useful to robots
- A graph specifying nodes and the connectivity between them
 - Nodes are not of fixed size
 - Free space not explicitly represented
 - A node is an area the robot can recognize its entry to and exit from

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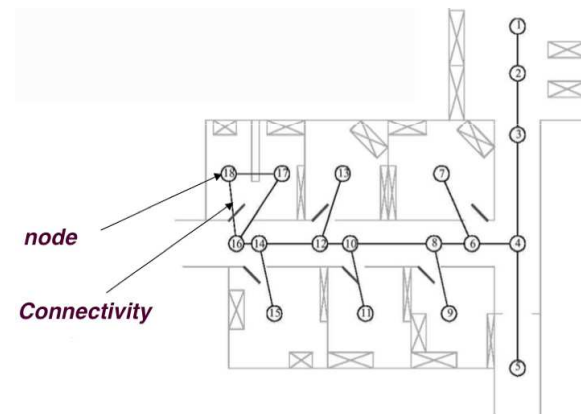
Topological Decomposition

- To robustly navigate with a topological map a robot
 - Must be able to localize relative to the nodes
 - Must be able to travel between the nodes
- These constraints require the robot's sensors to be tuned to the particular topological decomposition
- Major advantage is the ability to model non-geometric features (like artificial landmarks) that help localization

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Topological Decomposition Example

- Here the robot must be able to detect intersections between halls, and halls and rooms.



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Current challenges

- Real world is dynamic
- Perception still very error prone
 - Hard to extract useful information
 - Occlusion of objects
- Traversal of open space
- How to build up topology
- Sensor fusion

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The Simultaneous Localization and Mapping (SLAM) Problem

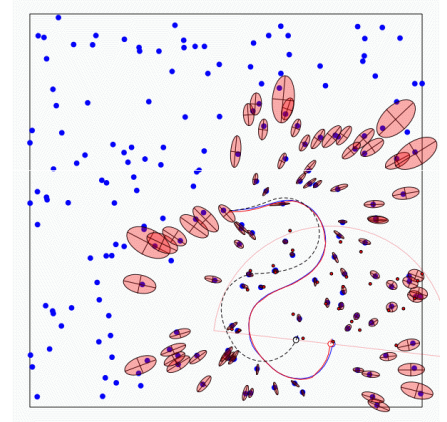
Consider a robot exploring an unknown, static environment

Given:

- The robot's controls
- Observations of nearby features

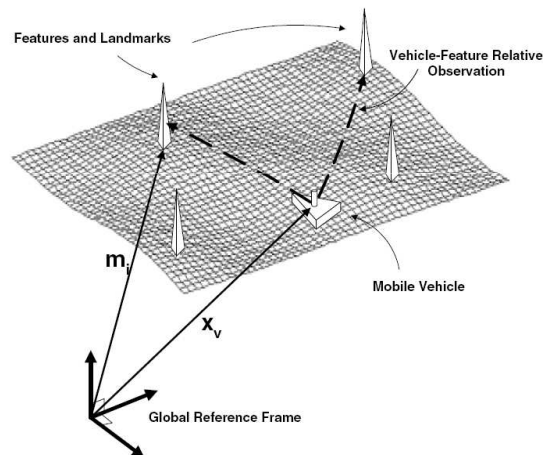
Estimate:

- Map of features
- Path of the robot



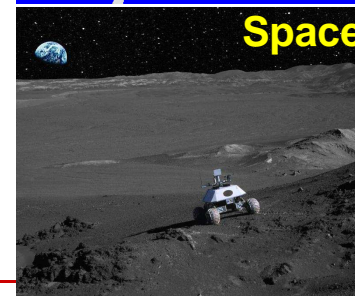
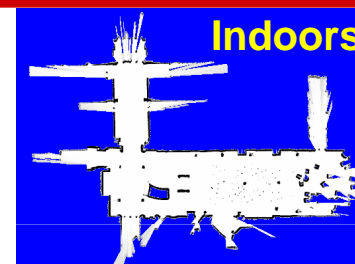
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Structure of the Landmark-based SLAM-Problem



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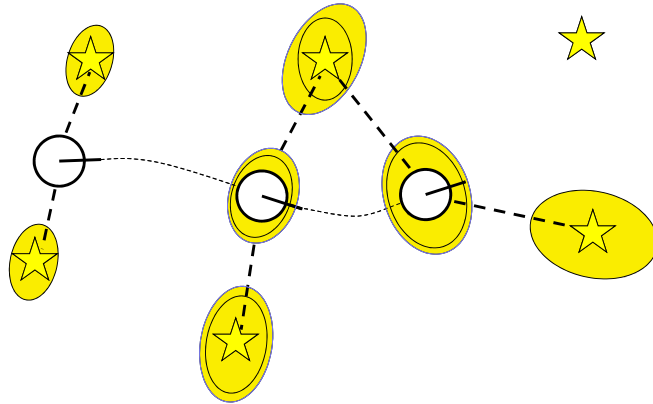
SLAM Applications



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Why is SLAM a hard problem?

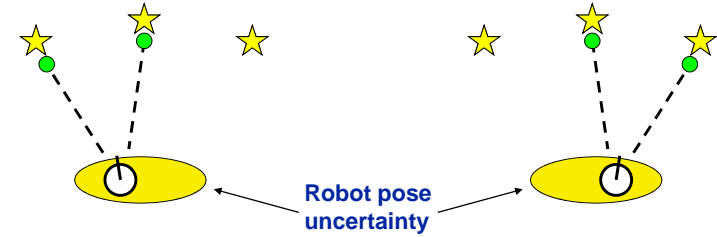
SLAM: robot path and map are both **unknown**



Robot path error correlates with the errors in the map

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Why is SLAM a hard problem?



- In the real world, the mapping between observations and landmarks is unknown
- Picking wrong data associations can have catastrophic consequences
- Pose error correlates data associations

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