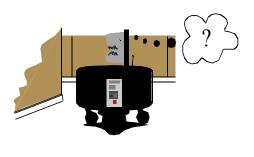
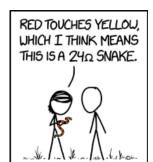
Localization where am I? (again?)



Bookkeeping



- ◆ Assignment 3
 - ◆ Comments?
- ♦ Next Reading: none
 - ◆ Unless you are behind; catch up
- ◆ Today
 - ◆ Knowledge Representation
 - Maps
 - Belief states
- Upcoming
 - ◆ Grades



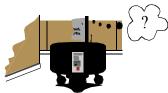
The last band of color indicates the snake's tolerance for being held before biting.

Localization Review (1)



3

- ◆ What is localization?
 - Figuring out location wrt. a model of the world
- ◆ What are the two purely proprioceptive approaches?
 - Odometry: belief about motion only
 - Wheel encoders, mostly
 - ♦ Dead reckoning: belief about motion + heading sensors





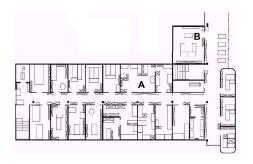


Localization Review (2)



- What is sensor aliasing?
 - ♦ Different locations giving the same sensor readings
- ◆ What is behavior-based navigation?
 - Navigating without localizing





Belief Representations



5

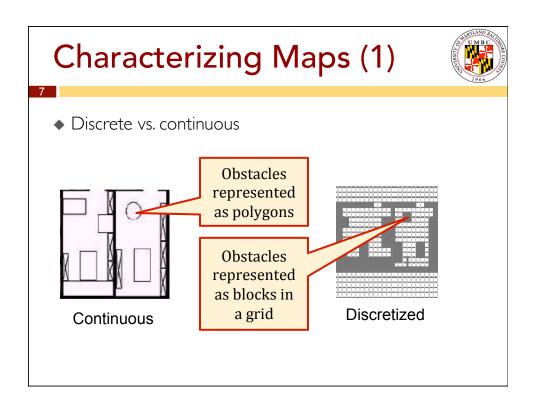
- ◆ (Model of) the map or environment
 - Discrete vs. continuous
 - Probabilistic vs. labeled
 - Geometric vs. topographical vs. semantic
- ◆ Beliefs about the robot's state or location
 - Discrete vs. continuous
 - Probabilistic vs. bounded vs. point
 - Single vs. multiple hypotheses
- Paths
 - Consecutive vs. kidnapped

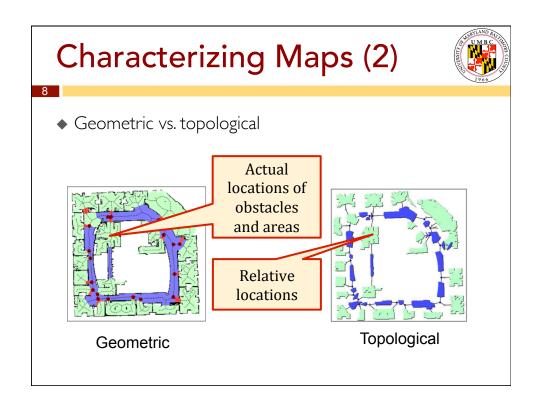
Design decisions: based on storage efficiency, reasoning speed, sensor capability, intended task, ...

Map Representations



- ◆ How precise does it have to be?
 - ◆ To accomplish what?
- ♦ What types of features are represented?
 - Depends on robot's sensors
 - ◆ If the robot can't see it, no point storing it
 - ♦ How much processing power do we have?
- What characteristics does it have?



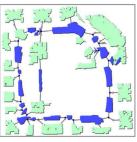


Characterizing Maps (3)

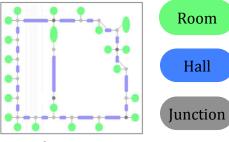


9

- ◆ Semantically labeled
 - ◆ Example: semantically labeled topological map





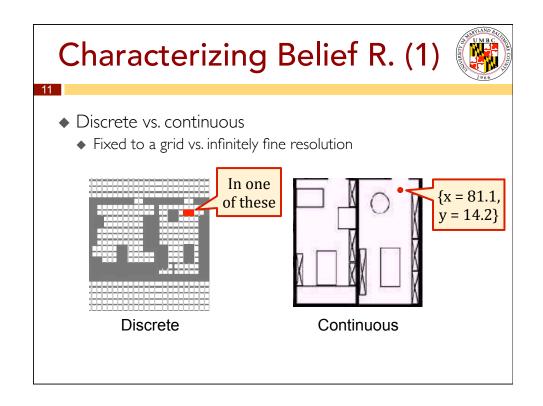


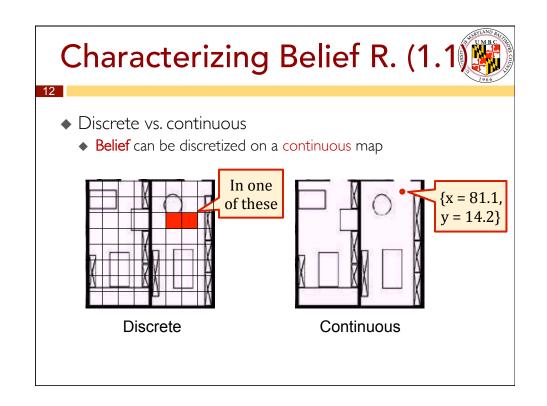
Semantic

Location (Belief) Representation



- ◆ What characteristics does it have?
- Discrete vs. continuous
 - Fixed to a grid, or anywhere?
- ◆ Single vs. multiple hypotheses
 - At any given time, how many possible locations are being considered?
- ◆ Probabilistic vs. bounded vs. point
 - ◆ The first two are multiple-hypothesis

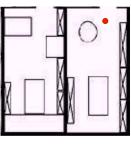




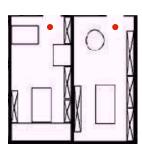
Characterizing Belief R. (2)



◆ Single hypothesis vs. multiple hypothesis



Single



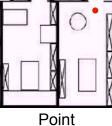
Multiple

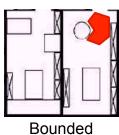
Characterizing Belief R. (3)



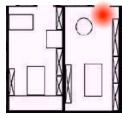
14

◆ Probabilistic vs. bounded vs. point





Polygon



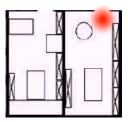
Probabilistic

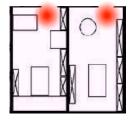
- ◆ You are here
- Somewhere in here (undifferentiated)
- Spread of likelihood

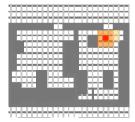
Probability & Combinations



◆ Single or multiple, discrete or continuous







Single Hypothesis

Multiple Hypothesis

Discrete

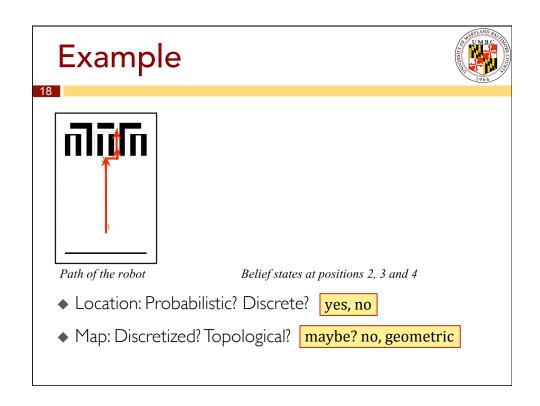
◆ Point: these are orthogonal choices

Belief Representation



- ◆ a) Continuous map with single hypothesis
- ♦ b) Continuous map with multiple hypothesis
- ♦ d) Discretized map with probability distribution
- d) Discretized topological map with probability distribution

Belief Repr position x • a) Continuous map probability P with single hypothesis ♦ b) Continuous map with multiple hypothesis probability P ♦ d) Discretized map with probability distribution ◆ d) Discretized probability P topological map with probability distribution



The Environment



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- ◆ Can contain:
 - ◆ Static or dynamic obstacles
 - Features (e.g., doors, floor tiles)
- ◆ Can be semantically labeled
- ◆ Environment Representation
 - ♦ Continuous Metric $\rightarrow \{x,y,\theta\}$
 - ◆ Discrete Metric
 → metric grid (eg, sq. D76)
 - ◆ Discrete Topological
 → topological grid

The Environment: Features



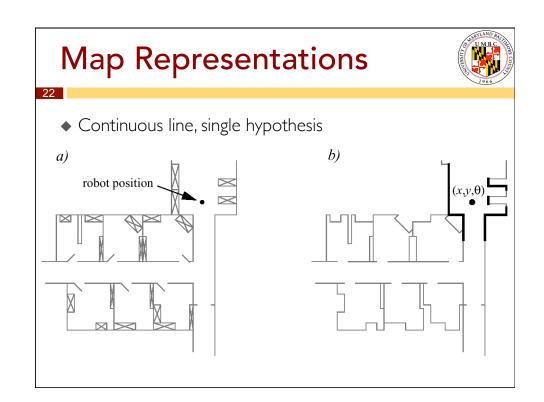
to get

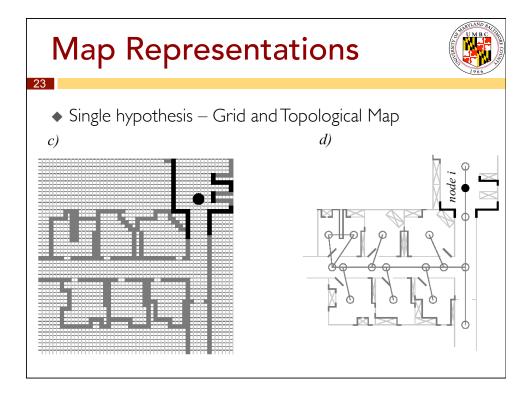
- ◆ Raw sensor data (ex.: laser range, grayscale images)
 - ◆ Lots of data, low distinctiveness (per reading)
 - ◆ Uses all acquired information
- ◆ Low level features (ex.: line extraction)
 - Some data, average distinctiveness
 - Filters out some useful information, still ambiguities
- ◆ High level features (ex.: doors, a car, the Eiffel tower)
 - ◆ Little data, high distinctiveness
 - ◆ Filters out the useful information, few/no ambiguities, insufficient environmental information

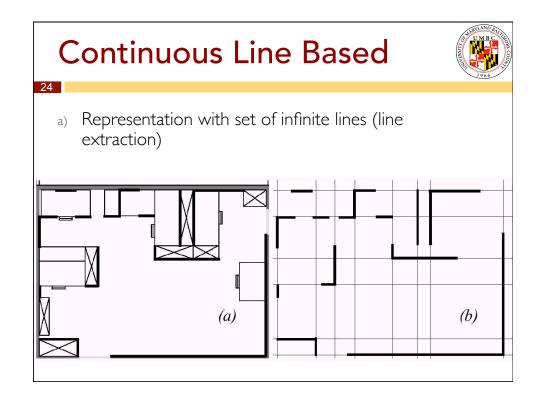


About Map Representations

- 1. Map precision vs. application
 - ◆ How precise does it need to be?
- 2. Features precision vs. map precision
 - ◆ 20cm. map precision ≠ 20cm. obstacle avoidance
- 3. Precision vs. computational complexity
 - More capability = more computational complexity
- ◆ Continuous Representation
- Decomposition (Discretization)





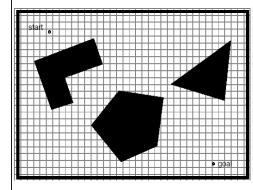


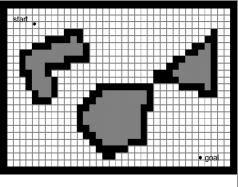
Map Decomposition (1)



27

- ◆ Fixed cell decomposition
 - Narrow passages disappear



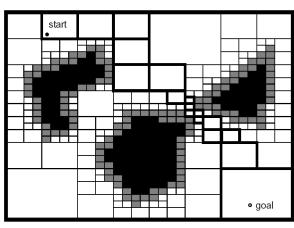


Map Decomposition (2)



28

◆ Adaptive cell decomposition

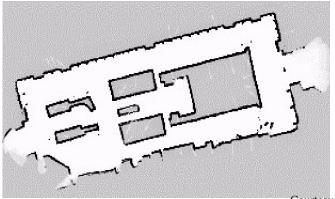


Map Decomposition (3)



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◆ Fixed cell decomposition — Example with very small cells



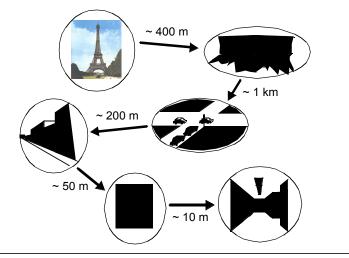
Courtesy of S. Thrun

Map Decomposition (4)



32

◆ Topological Decomposition

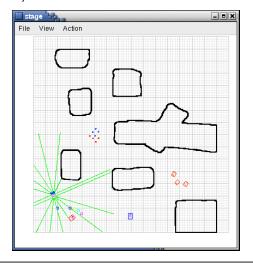


Map Decomposition (5)



33

♦ Occupancy Grid



Probabilistic Map-Based Localization

- ◆ Consider a mobile robot moving in a known environment
- ◆ As it starts to move from a precisely known location, it might keep track of its location using odometry.
- ◆ However, after a certain movement the robot will get very uncertain about its position.
- → update using an observation of its environment.
- observation lead also to an estimate of the robots position which can than be fused with the odometric estimation to get the best possible update of the robots actual position.

