

CS-417 INTRODUCTION TO ROBOTICS AND INTELLIGENT SYSTEMS

Exploration

Three Main Challenges in Robotics

1. Where am I? (Localization)

- Sense
- relate sensor readings to a world model
- compute location relative to model
- assumes a perfect world model

2. What the world looks like? (Mapping)

- sense from various positions
- integrate measurements to produce map
- assumes perfect knowledge of position
- Together 1 and 2 form the problem of Simultaneous Localization and Mapping (SLAM)
- 3. How do I go from A to B? (Path Planning)
 - More general: Which action should I pick next?

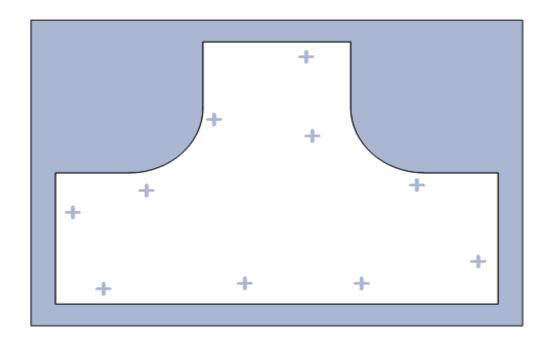
Mapping

- What the world looks like
- Improve the accuracy of the map
- Ensure that all the important parts of the environment are mapped – Exploration!

Environment Representation (Map)

- Grid Based Maps
- Feature Based Maps
- Topological Maps
- Hybrid Maps

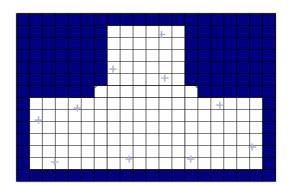
Consider this Environment:



Three Basic Map Types

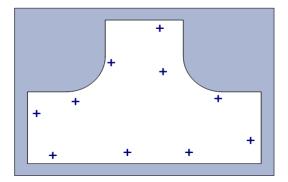
Grid-Based:

Collection of discretized obstacle/free-space pixels



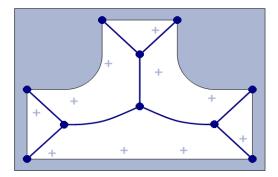
Feature-Based:

Collection of landmark locations and correlated uncertainty

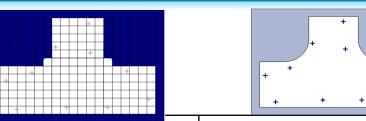


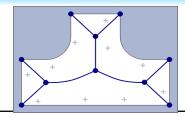
Topological:

Collection of nodes and their interconnections



Three Basic Map Types

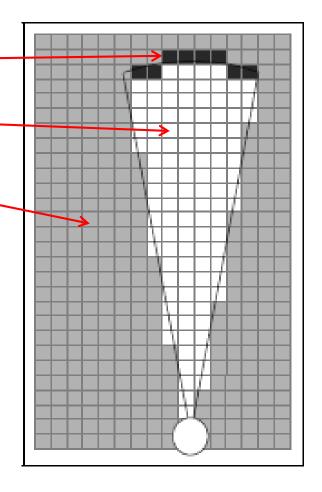




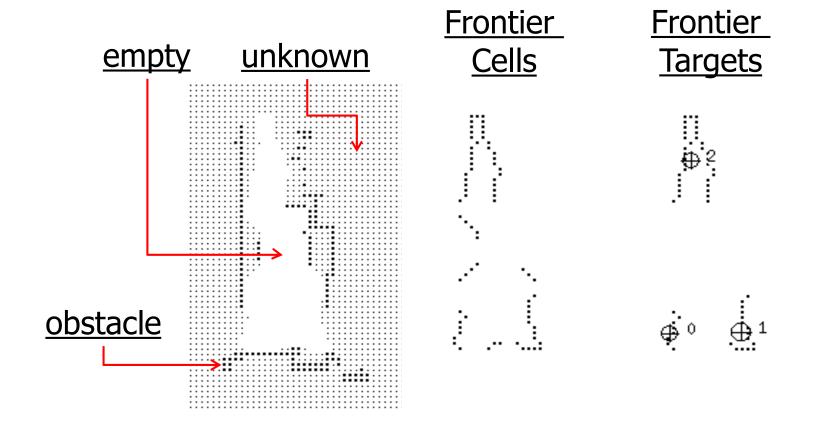
	Grid-Based	Feature-Based	Topological
Construction	Occupancy grids	Kalman Filter	Navigation control laws
Complexity	Grid size and resolution	Landmark covariance (N³)	Minimal complexity
Obstacles	Discretized obstacles	Only structured obstacles	GVG defined by the safest path
Localization	Discrete localization	Arbitrary localization	Localize to nodes
Exploration	Frontier-based exploration	No inherent exploration	Graph exploration

Grid Based Maps

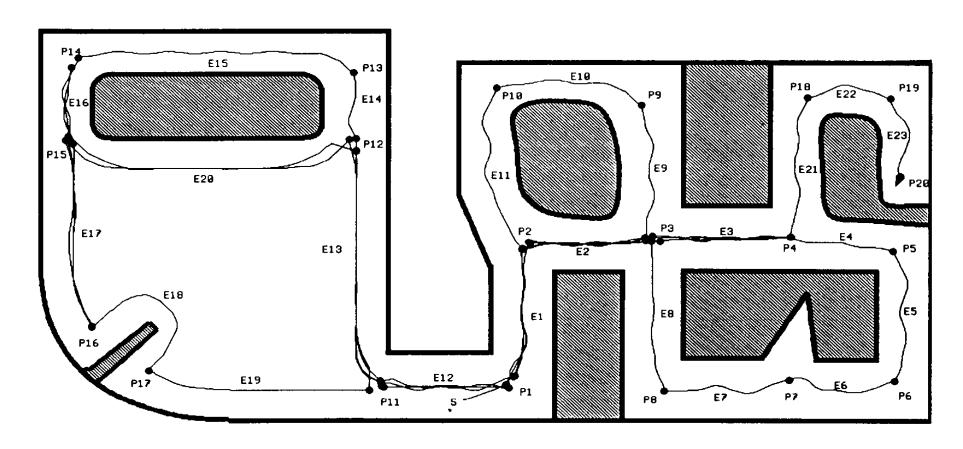
- Occupied cells
- Free cells
- Unknown cells



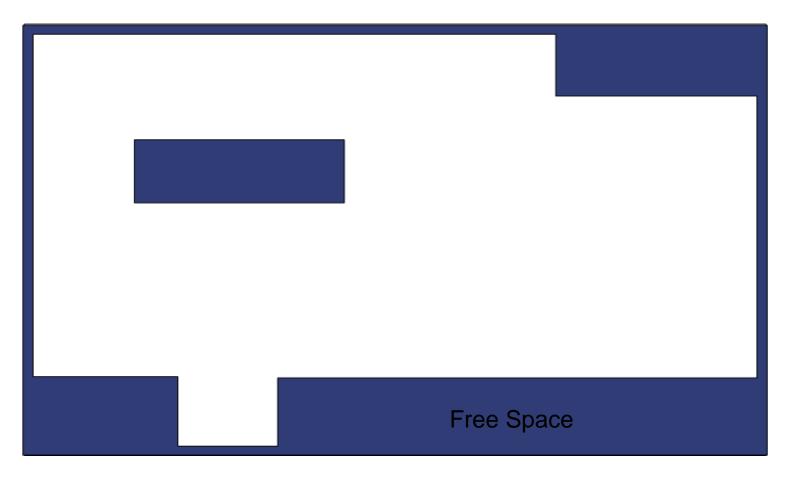
Frontier based Exploration (Grid Maps)



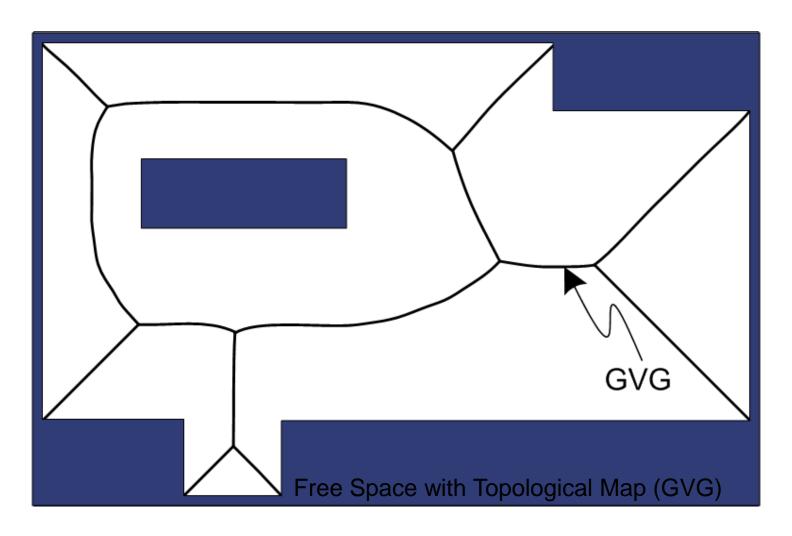
Topological Representations



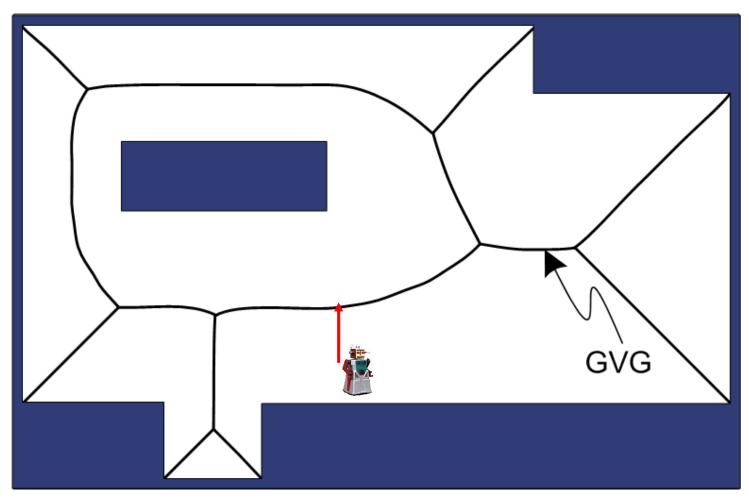
• B. J. Kuipers and Y.-T. Byun. "A robot exploration and mapping strategy based on a semantic hierarchy of spatial representations". In *Journal of Robotics and Autonomous Systems*, 8: 47-63, 1991.



H. Choset, J. Burdick, "Sensor based planning, part ii: Incremental construction of the generalized voronoi graph". In IEEE Conference on Robotics and Automation, pp. 1643 – 1648, 1995.

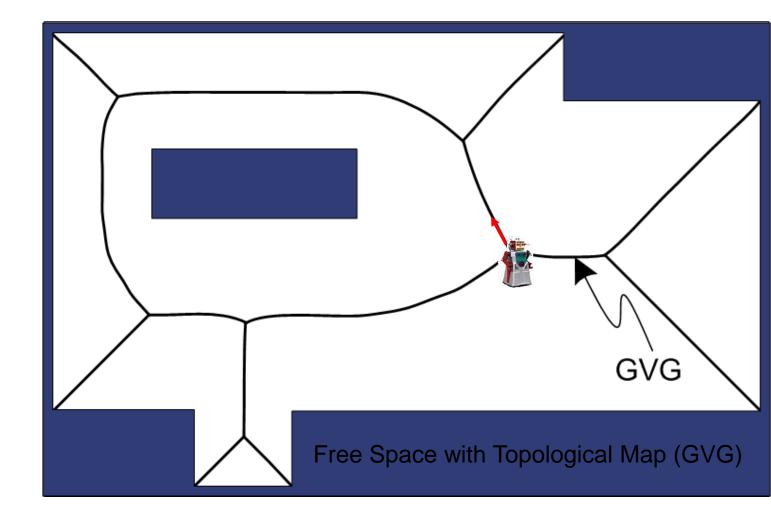


Access GVG



Free Space with Topological Map (GVG)

- Access GVG
- Follow Edge
- Home to the MeetPoint
- Select Edge

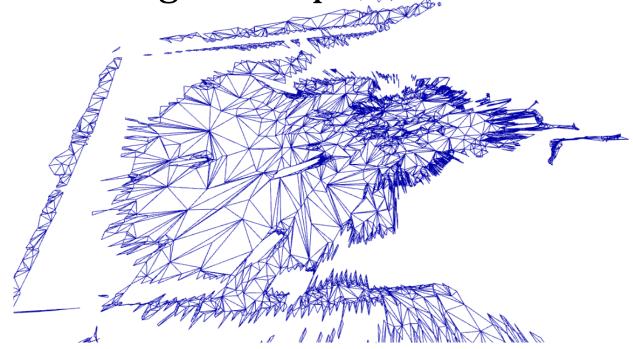


Exploration via Graph Search

- Exhaustive Depth First Search
- Breadth-First Search
- Heuristics

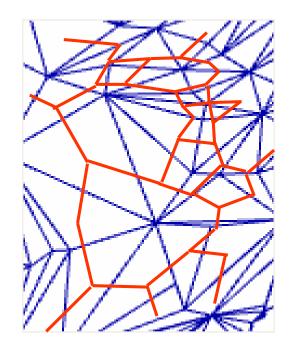
Irregular Triangular Mesh (ITM)

- Terrain Representation
- Underlying Topological Structure
- Path Planning and Exploration



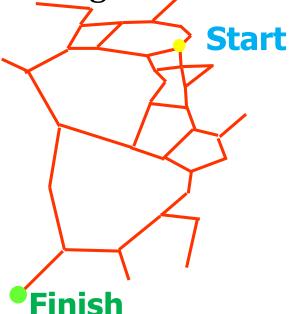
From 2.5D Representation to Topological

Convert ITM into Connected Graph

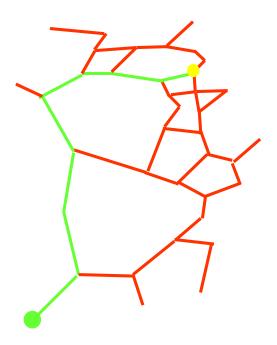


- Convert ITM into Connected Graph
- Planning using Graph Search Algorithms:

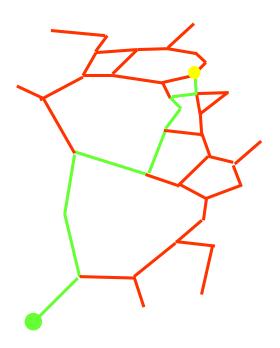
Dijkstra, A* search algorithms



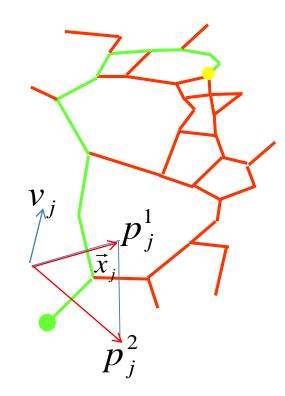
- Convert ITM into Connected Graph
- Path Planning using Graph Search Algorithms:
 - Dijkstra, A* search algorithms
- Different Cost Functions Q
 - Number of triangles Q = 1



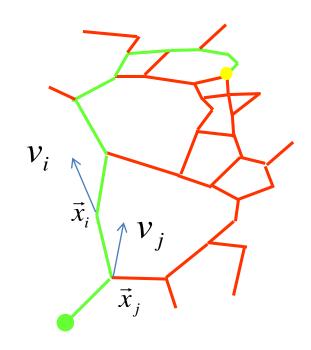
- Convert ITM into Connected Graph
- Path Planning using Graph Search Algorithms:
 - Dijkstra, A*
- Different Cost Functions Q
 - Number of triangles
 - Euclidian distance $Q = \|\vec{x}_i \vec{x}_j\|$



- Convert ITM into Connected Graph
- Path Planning using Graph Search Algorithms:
 - Dijkstra, A*
- Different Cost Functions Q
 - Number of triangles
 - Euclidian distance
 - Slope of each triangle $v_j = \frac{p_j^1 \times p_i^2}{\|p_j^1\| \|p_j^2\|}$

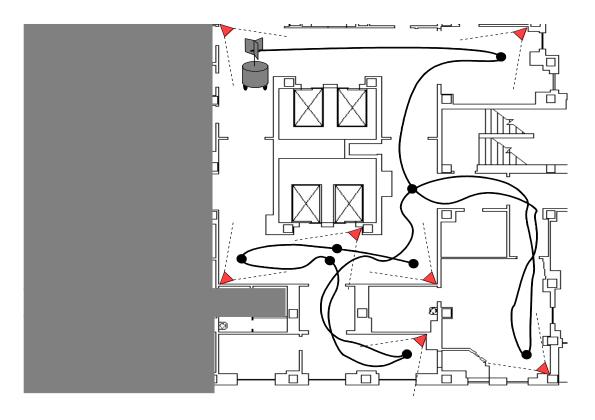


- Convert ITM into Connected Graph
- Path Planning using Graph Search Algorithms:
 - Dijkstra, A*
- Different Cost Functions Q
 - Number of triangles
 - Euclidian distance
 - Slope of each triangle
 - Cross triangle slope



Exploration Planning Problem

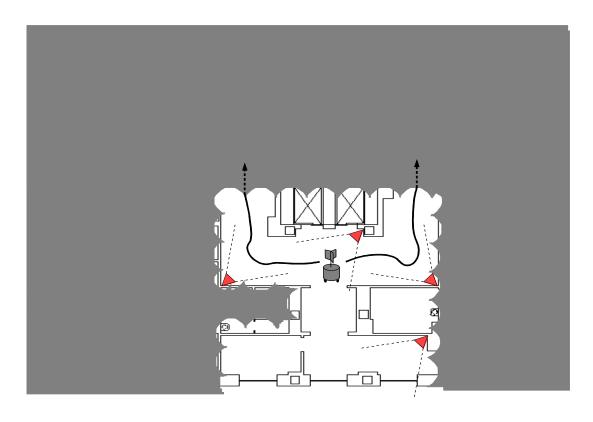
Two fundamental problems for path planning during exploration and mapping:



Exploration Planning Problem

Two fundamental problems for path planning during exploration and mapping:

- Planning for relocalization
- Planning the exploration of new territory



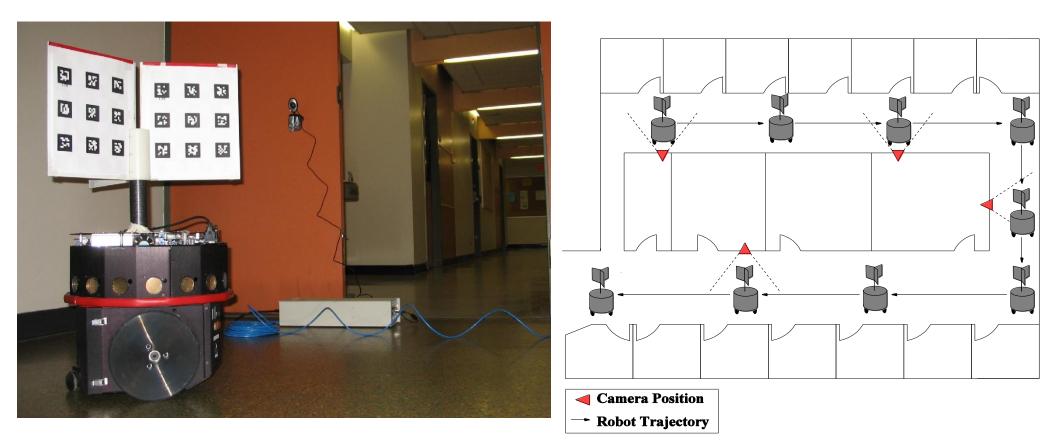
Previous Localization Planning

- Reduce measure of map or position entropy
- Variety of graph search planning algorithms (breadth first, A*-search, RRT)
- Evaluate paths with simulation, or Cramer-Rao bounds for expected uncertainty
- e.g. [Fox et al RAS 1998], [Sim and Roy ICRA 2005], [He et al ICRA 2008], [Censi et al ICRA 2008]

Previous Exploration Planning

- Includes motion into unexplored regions
- Typically requires prior knowledge of environment properties or rough layout
- Computation of exploration effects is a challenge
- e.g. [Bourque and Dudek IROS 1999], [Bourgault et al IROS 2002], [Kollar and Roy IJRR 2008]

Exploring a Camera Sensor Network



D. Meger, I. Rekleitis, and G. Dudek. "Heuristic Search Planning to Reduce Exploration Uncertainty", IROS 2008.

Heuristic Search Planning Method

- Solution to exploration planning for camera sensor networks
 - Composed of two alternated steps: exploration and re-localization
 - Combined distance and uncertainty cost function
 - Heuristic search for good paths

- Decision (exploration vs exploitation)
- Target Node
- Path Planning through the known graph
- Exploration Strategies

- Decision (exploration vs. exploitation)
 - Epsilon-Greedy
 - Epsilon-First
 - Adaptive
 - Bounded Uncertainty
- Target Node
- Path Planning through the known graph
- Exploration Strategies

- Decision (exploration vs. exploitation)
- Target Node (Exploration)
 - Random
 - Shortest distance
 - Maximum Uncertainty
 - Minimum Uncertainty
- Path Planning through the known graph
- Exploration Strategies

- Decision (exploration vs. exploitation)
- Target Node (Relocalization)
 - Maximum Uncertainty
- Path Planning through the known graph
- Exploration Strategies

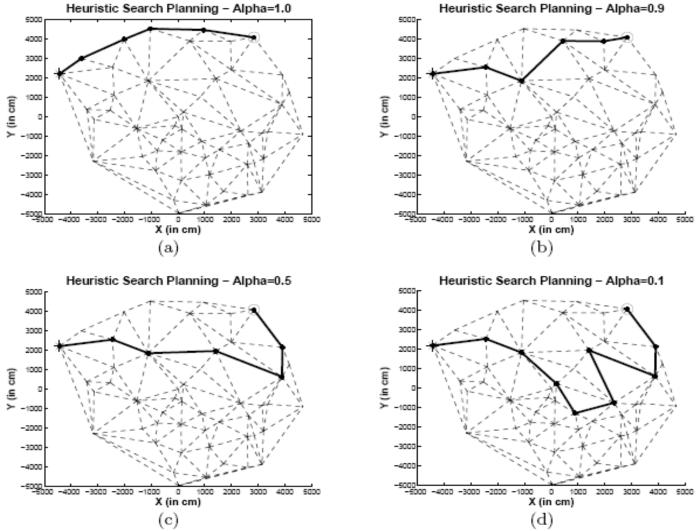
- Decision (exploration vs. exploitation)
- Target Node
- Path Planning through the known graph
 - Work with D. Meger and G. Dudek [IROS 2008]
 - A* based strategy
 - Cost: $C(p) = \omega_d length(p) + \omega_u trace(P(p))$
 - Distance-based "cost-to-go" heuristic function h used to compute estimated cost

$$C(n) = f(n) + h(n)$$

C(n) = f(n) + h(n) Estimated cost through n Cost so far Estimated cost to go

Exploration Strategies

Effect of α Parameter for Relocalization



Heuristic Search

 Graph search to optimize cost function $C(p) = \omega_d length(p) + \omega_u trace(\Sigma(p))$

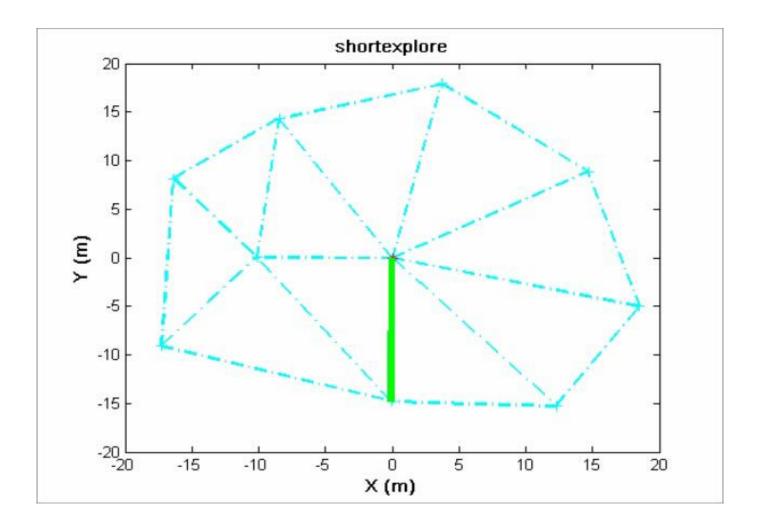
- Heuristic search allows considering only a fraction of the paths, ordered by expected cost
- Distance-based "cost-to-go" heuristic function h used to compute estimated cost

$$C(n) = f(n) + h(n)$$

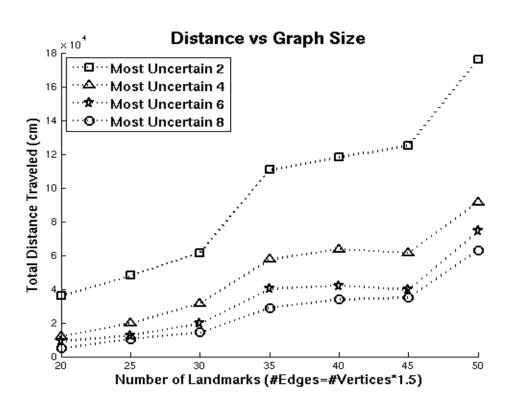
Estimated cost through n Cost so far Estimated cost to go

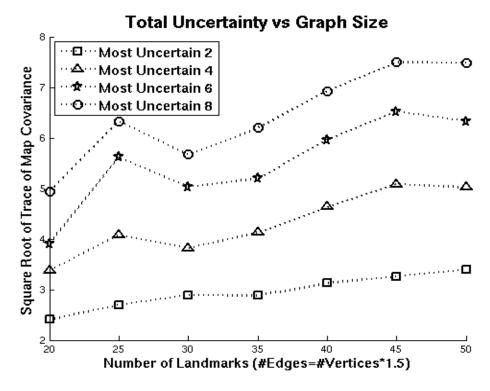
- Decision (exploration vs. exploitation)
- Target Node
- Path Planning through the known graph
- Exploration Strategies
 - One Step Exploration
 - Ear based exploration (submitted to IROS 2012)

Shortest Node P(exploit)=0.3

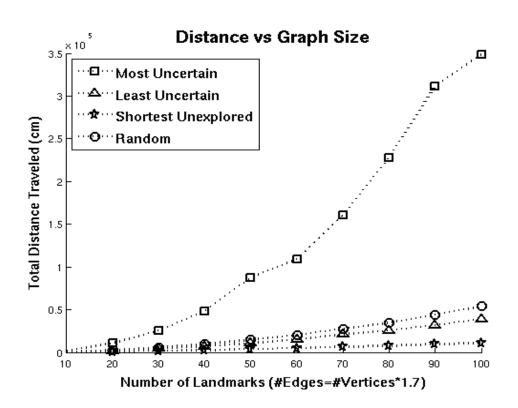


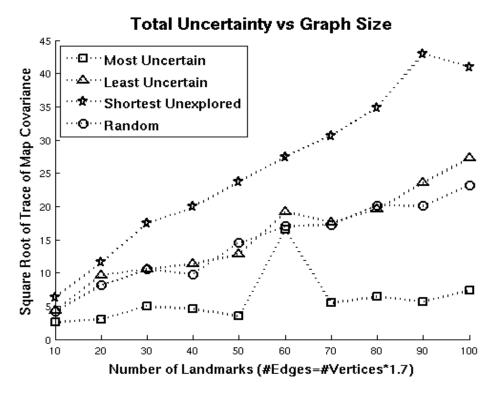
Experimental Results Bounded Uncertainty





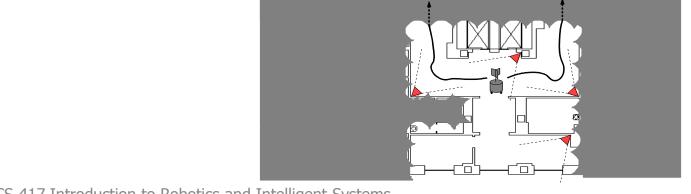
Experimental Results Different Strategies





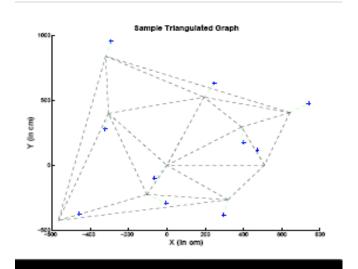
Planning Exploratory Steps

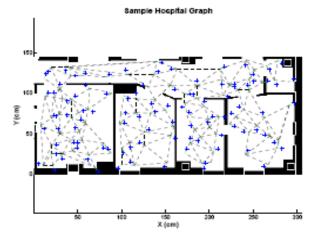
- Choose motion in unexplored space to locate additional camera nodes
- Planner cannot simulate these paths
- Evaluated 2 strategies: 1) nearest camera and 2) a randomly selected camera



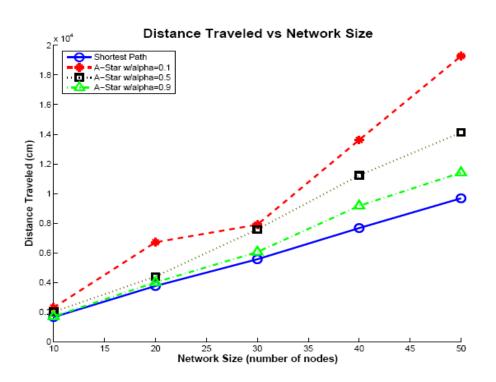
Simulation Results

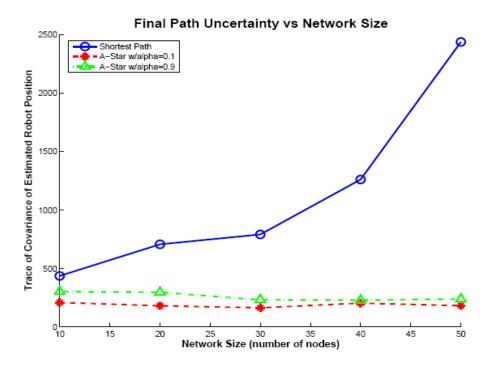
- Compared planners over many trials
- 3 realistic network types (2 shown)
- 3 methods for comparison:
 - Depth-first
 - Return to origin
 - Return to nearest explored



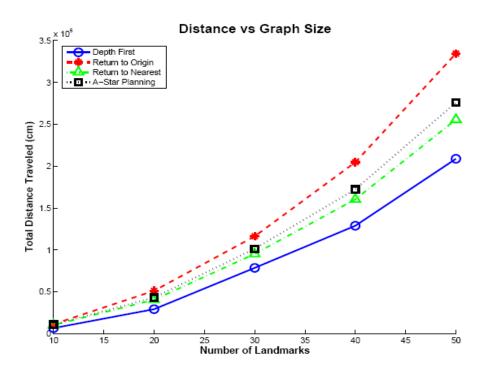


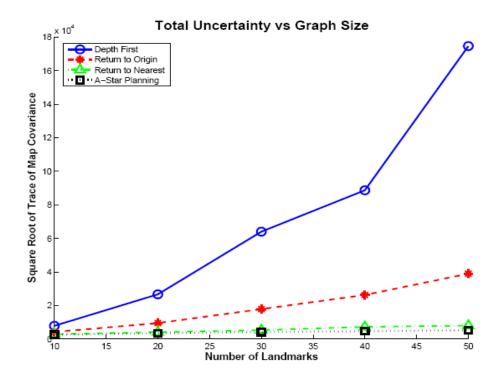
Simulated Relocalization Results





Simulated Exploration Results





Key Points

- Mapping requires exploration
- Exploration strategies depend on the representation
- Topological representations are the most convenient for exploration
- Two objectives:
 - Explore new territory
 - Improve the accuracy by relocalization

References

- B. J. Kuipers and Y.-T. Byun. "A robot exploration and mapping strategy based on a semantic hierarchy of spatial representations". In *Journal of Robotics and Autonomous Systems*, 8: 47-63, 1991.
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- R. Martinez-Cantin, N. de Freitas, A. Doucet, and J. Castellanos, "Active policy learning for robot planning and exploration under Uncertainty". In Robotics: Science and Systems, 2007.
- D. Meger, I. Rekleitis, and G. Dudek. "Heuristic Search Planning to Reduce Exploration Uncertainty". In *IEEE/RSJ International Conference on Intelligent Robots and Systems*, pp 3382-3399, 2008.

QUESTIONS?

