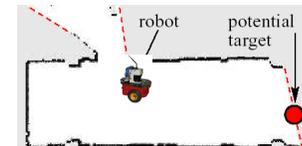


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Robotics

Exploration

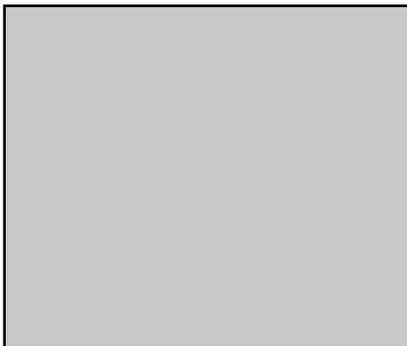
Single Robot Exploration

- Frontiers between free space and unknown areas are potential target locations
- Going to frontiers will gain information



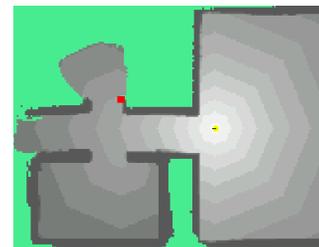
- Select the target that minimizes a cost function (e.g. travel time / distance /...)

Frontier-Based Exploration



Multi-Robot Exploration

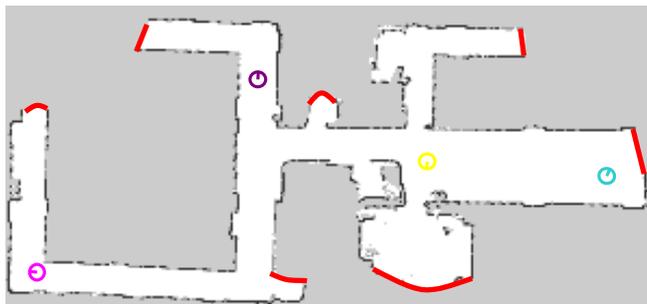
Robot 1:



Robot 2:



Coordinated Exploration



$$C(\theta) = \sum_{(i,j) \in \theta} \text{dist}(i,j)$$

$$U(\theta) = \sum_{(i,j) \in \theta} \text{explore}(i,j)$$

$$\theta^* = \arg \max_{\theta} (U(\theta) - C(\theta))$$

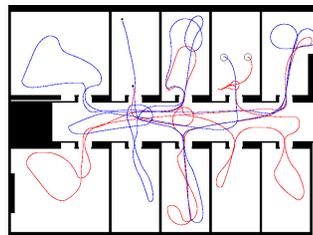
[Burgard et al. 00],
[Simmons et al. 00]

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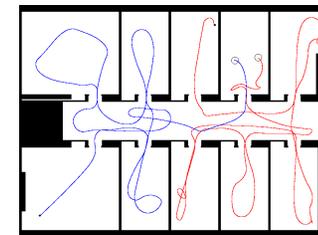
5

Typical Trajectories in an Office Environment

Implicit / no coordination:

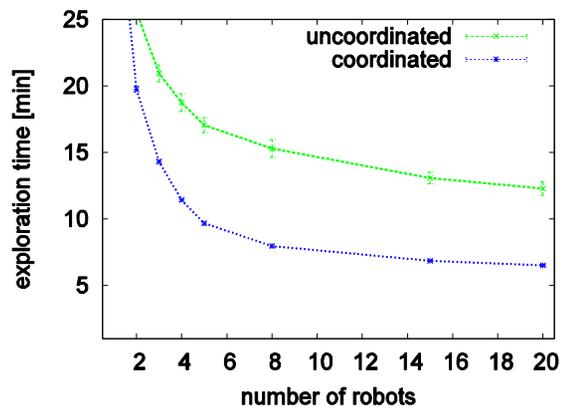


Explicit coordination:



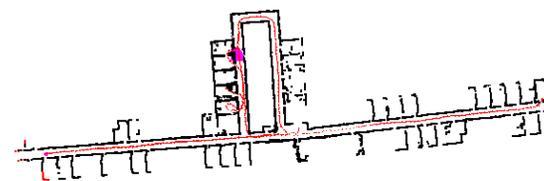
6

Exploration Time



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Multi-Robot Mapping With Known Start Locations



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Multi-Robot Mapping With Known Start Locations



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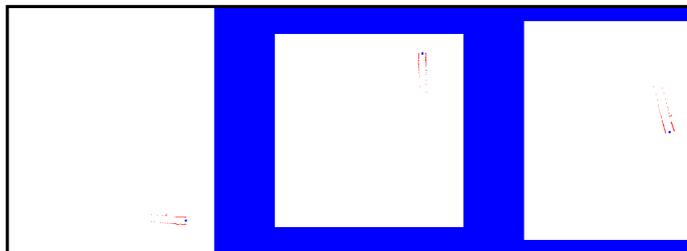
9

Sponsored by DARPA-SDR, NSF, Intel



- ▶ Map an unknown area
- ▶ Search for an “object of value”
- ▶ Set up a surveillance network
- ▶ Track any intruders

Why are Unknown Start Locations Hard?



- ▶ Need to know whether or not maps overlap
- ▶ Need to know how maps overlap

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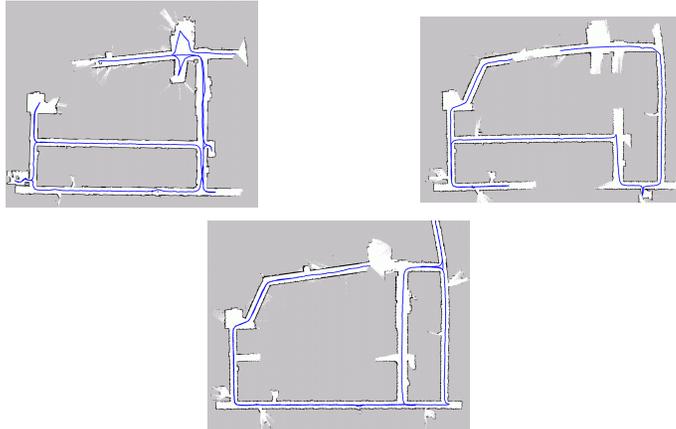
Multi-robot Map Merging

- **Problems**
 - Number of possible merges is **exponential** in number of robots
 - Cannot merge maps by simply **overlaying** them
- **Wanted**
 - **Scalability, robustness**
 - Merge maps **as soon as possible**

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Multi-robot Map Merging



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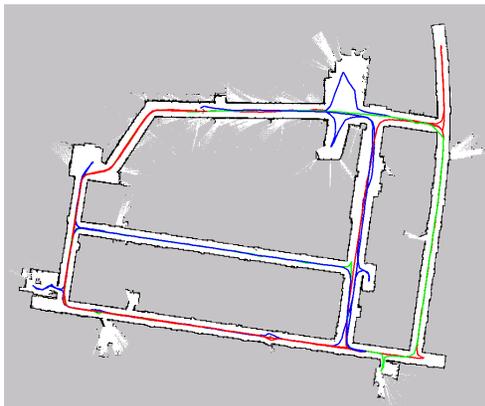
Multi-robot Map Merging



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Multi-robot Map Merging



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Estimating relative locations

- **Idea:** Localize one robot in other robot's map using particle filter
- **Problems:**
 - Only partial map available
 - Other robot might be outside the map
 - Map grows
 - Impossible to keep track of all locations inside and outside the partial map
- **Solution:** Only keep track of trajectories that overlapped map at some time

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Partial map localization (intuition)



- Overlapping trajectories

$$p(x_t | z_{1:t}, u_{1:t-1}) = \alpha_t p(z_t | x_t) \cdot \left[\int p(x_t | x_{t-1}, u_{t-1}) p(x_{t-1} | z_{1:t-1}, u_{1:t-2}) dx_{t-1} + \boxed{p(x_t | n_{t-1}, u_{t-1}) p(n_{t-1} | z_{1:t-1}, u_{1:t-2})} \right]$$

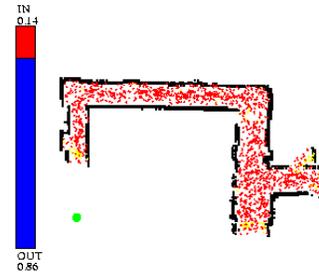
- Non-overlapping trajectories

$$p(n_t | z_{1:t}, u_{1:t-1}) = \alpha_t p(z_t | \text{outside})(1 - \epsilon) p(n_{t-1} | z_{1:t-1}, u_{1:t-2})$$

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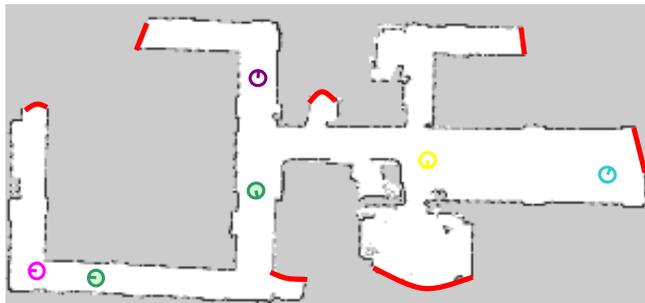
Partial map localization (example)



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Coordination



$$C(\theta) = \sum_{(i,j) \neq \emptyset} \begin{cases} \text{dist}(i,j) & \text{if } j \text{ is frontier} \\ \text{dist}(i,j) + \text{meet}(i,j) & \text{if } j \text{ is hypothesis} \end{cases} \quad U(\theta) = \sum_{(i,j) \neq \emptyset} \begin{cases} \text{explore}(i,j) & \text{if } j \text{ is frontier} \\ p(j) \text{merge}(i,j) & \text{if } j \text{ is hypothesis} \end{cases}$$

Hypotheses become potential goals

[Burgard et al. 00],
[Simmons et al. 00],
[Zlot et al. 02]

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Experimental setup



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Coordinated exploration with three robots
from unknown start locations

The robots are fully autonomous.
All computation is performed on-board.

Shown is the perspective of one robot

CentiBots: Experimental Evaluation

- Rigorously tested by **outside evaluation team**
- **No testing** allowed in 1/2 of environment
- **Limited communication**
- No intervention / observation during experiment
- Comparison to **“ground truth”** map

Control Center and Test Team



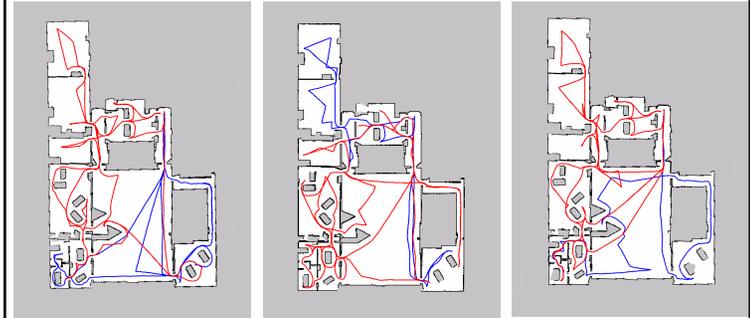
Comparison to “Ground Truth Map”



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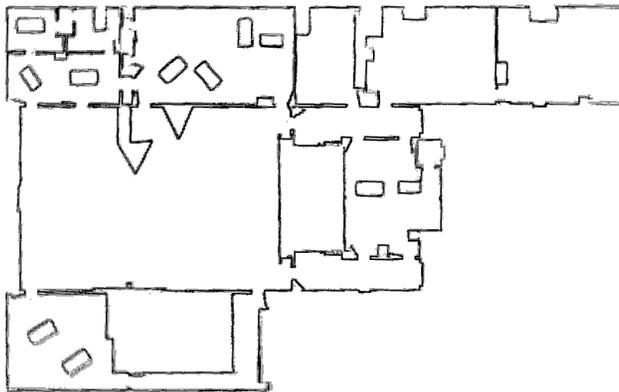
Three Mapping Runs



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Three Overlaid Maps

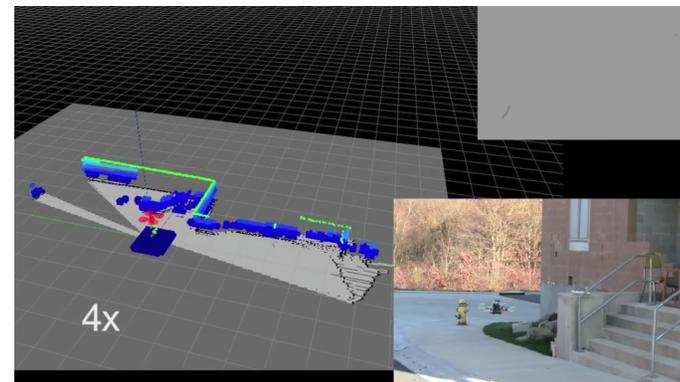


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Courtesy of Vijay Kumar 3D Exploration

[Shen-Michael-Kumar: IJRR-2012]

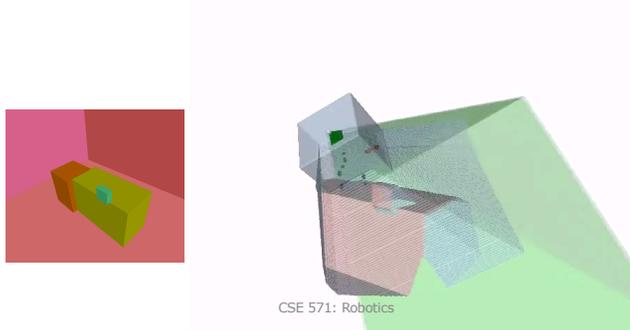


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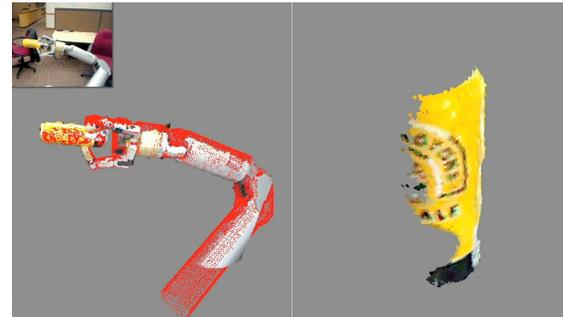
Active Mapping

- View selection for mapping and segmentation



Active Object Modeling: Joint Tracking and Modeling

[Krainin-Henry-Ren-F: IJRR-11]



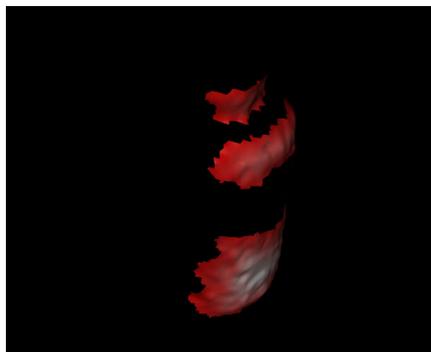
- EKF with articulated ICP over manipulator joint angles, camera pose and pose of (partial) object

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Uncertainty in Object Surface

[Krainin-F-Curless: ICRA-11]



- Signed-distance function voxel grid [Curless '96]
- Surface uncertainty from beam-based noise model

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View Selection Algorithm

- Conceptually similar to Planetarium Algorithm [Connolly '85]
- Procedure:
 - Generate kinematically achievable viewpoints
 - Compute information gain (quality) for each viewpoint
 - Select view as tradeoff between quality and cost

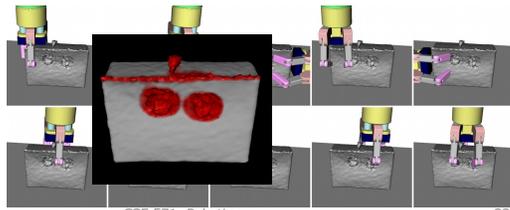


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Re-Grasp Selection

- Generate candidate grasps [Diankov '10]
- Select grasp by maximum information gain, accounting for occlusion caused by grasp

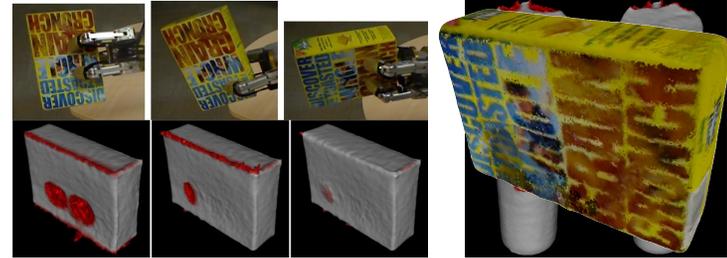


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Multiple Grasp Results

- Evaluated regrasping on four objects
- Includes box with three grasps



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Active Object Modeling

**Next Best View Planning
for 3D In-Hand Modeling**

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