# CSE-571 Probabilistic Robotics

## Multi-robot Mapping and Exploration

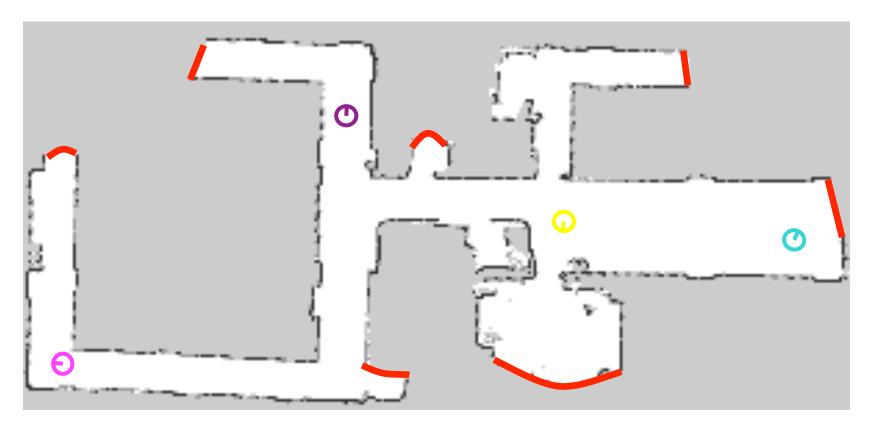


- Map an unknown area
- Search for an "object of value"
- Set up a surveillance network
- Track any intruders

### Mapping the Allen Center



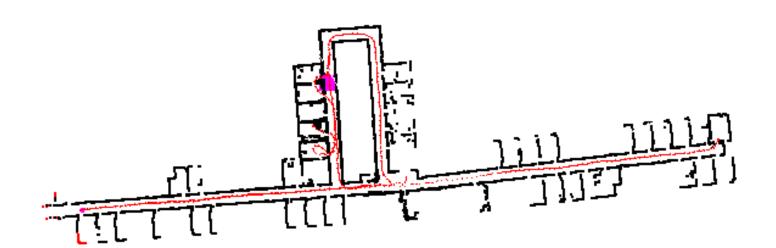
#### Coordinated Exploration



$$C(\theta) = \sum_{(i,j) \in \theta} \operatorname{dist}(i,j) \qquad U(\theta) = \sum_{(i,j) \in \theta} \operatorname{explore}(i,j)$$
 
$$\theta^* = \arg\max \bigl( U(\theta) - C(\theta) \bigr) \qquad \text{[Burgar Simmer of the property of$$

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

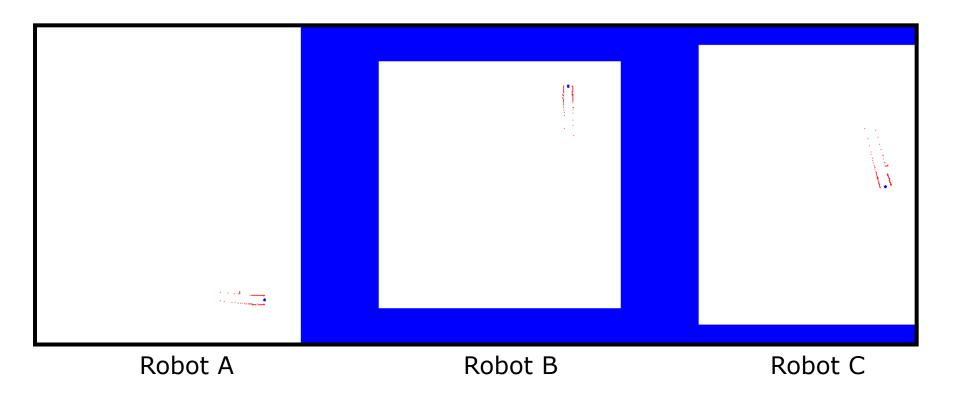
## Multi-Robot Mapping With Known Start Locations



#### Multi-Robot Mapping With Known Start Locations



#### Why are Unknown Start Locations Hard?



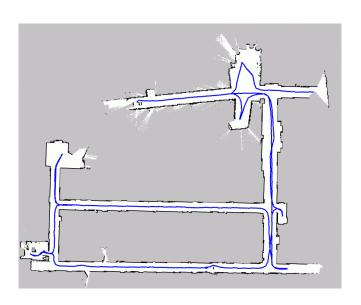
- ▶ Need to know whether or not maps overlap
- Need to know how maps overlap

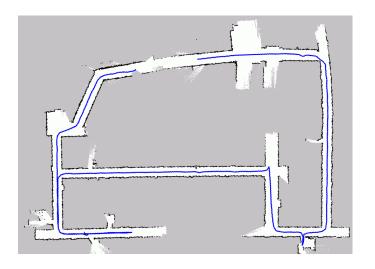
#### Problems

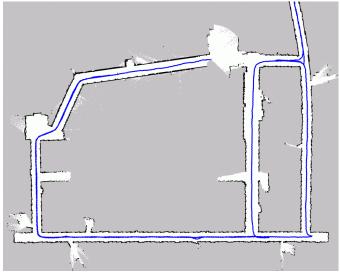
- Number of possible merges is exponential in number of robots
- Cannot merge maps by simply overlaying them

#### Wanted

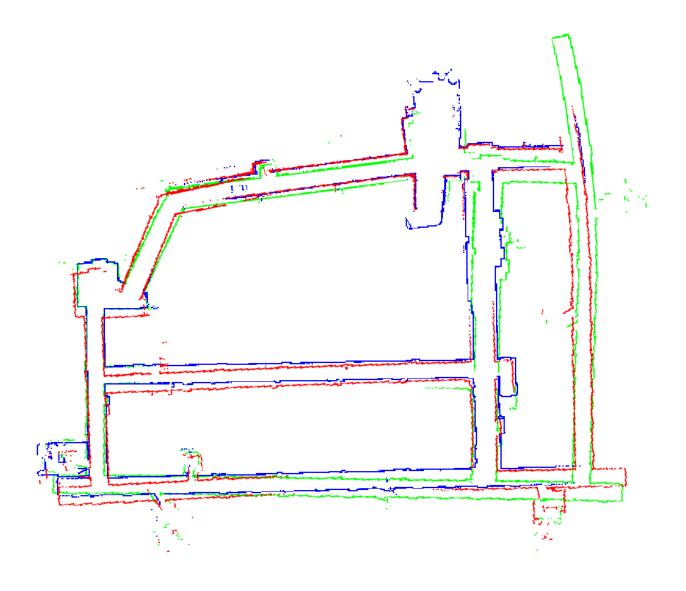
- Scalability, robustness
- Merge maps as soon as possible

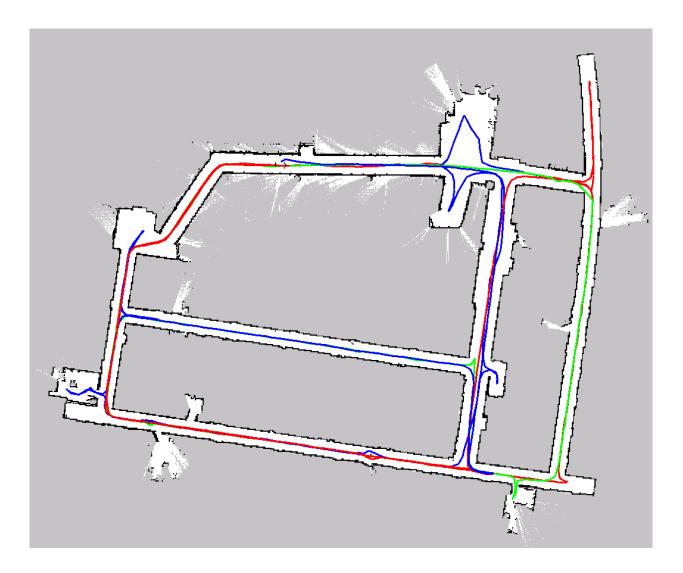






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### Estimating relative locations

 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

#### Partial map localization (intuition)



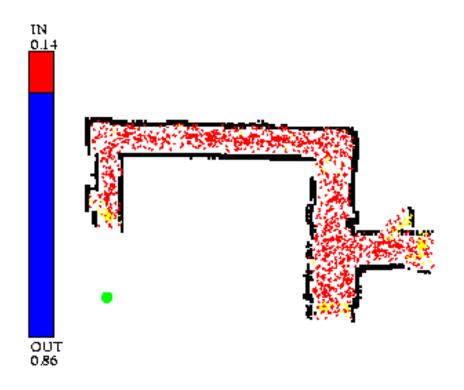
Overlapping trajectories

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

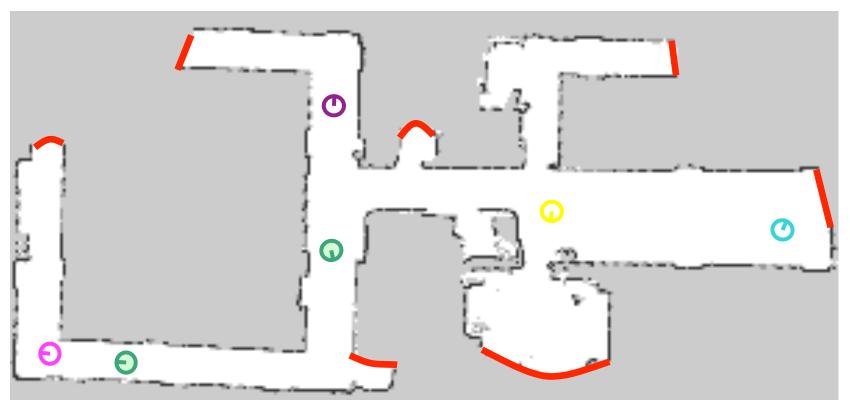
Non-overlapping trajectories

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

#### Partial map localization (example)



#### Coordination



$$C(\theta) = \sum_{(i,j) \in \theta} \begin{cases} \operatorname{dist}(i,j) & \text{if } j \text{ is frontier} \\ \operatorname{dist}(i,j) + \operatorname{meet}(i,j) & \text{if } j \text{ is hypothesis} \end{cases}$$

$$U(\theta) = \sum_{(i,j) \in \theta} \begin{cases} \exp(\operatorname{lore}(i,j)) & \text{if } j \text{ is frontier} \\ p(j) \operatorname{merge}(i,j) & \text{if } j \text{ is hypothesis} \end{cases}$$

#### Hypotheses become potential goals

[Burgard et al. 00], [Simmons et al. 00], [Zlot et al. 02]  $_{15}\,$ 



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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







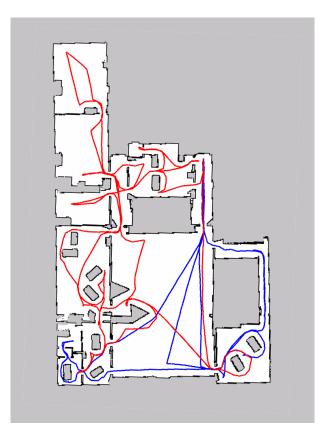


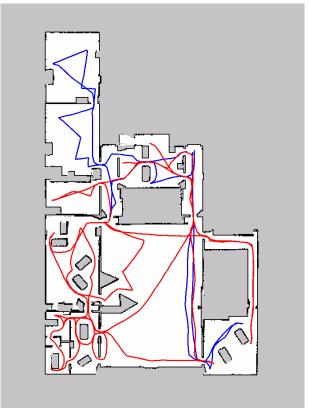
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## Comparison to "Ground Truth Map"



## Three Mapping Runs







## Three Overlayed Maps

