Comparison of exploration strategies for multi-robot search



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Content

- Problem definition (exploration x search)
- Goal-selection strategies
- Experimental evaluation
- Conclusion



Search

A process of autonomous navigation of a team of mobile robots in an unknown environment in order to find an object of interest (placed randomly) with minimal resources used.

Frontier based algorithm (Yamauchi)

while the object not found do read current sensor information update the map with the obtained data

determine new goal candidates assign the goals to the robots plan paths for the robots move the robots towards the goals



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- The robots are equipped with a sensor with a limited range able to detect an object.
- ► Finding the object: the object is firstly detected by the sensor.
- Goal: to minimize the expected time T_f when this occurs:

$$T_f = \mathbb{E}(T|\mathcal{R}) = \int_0^\infty t p(t) \, \mathrm{d}t,$$

- where $p(t) = \frac{A_t^R}{A_{total}}$ is the probability of finding the object at time t.
- ► A^R_t is the area newly sensed at time t when the robots follow the trajectory R
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- where $p(t) = \frac{A_t^k}{A_{total}}$ is the probability of finding the object at time *t*.
- ► A^R_t is the area newly sensed at time t when the robots follow the trajectory R
- ► *A_{total}*, the area of the whole environment the robot operates.
- The objective is to find trajectories \mathcal{R}^{opt} minimizing $\mathbb{E}(\mathcal{T}|\mathcal{R})$:

$$\mathcal{R}^{opt} = \arg\min_{\mathcal{R}} \mathbb{E}(T|\mathcal{R}) = \arg\min_{\mathcal{R}} \sum_{t=0}^{\infty} tA_t^{\mathcal{R}}$$



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Exploration





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Exploration





Exploration





ACBD is better

Exploration Search





ACBD is better

Search

B = 2 C = 2 + 3 + 7 = 12 B = 2 C = 2 + 3 = 5 D = 2 + 3 + 7 = 12 $\mathbb{E}(T) = \frac{2 + 5 + 12}{3} = 6.33$

Exploration



ACBD is better

B = 2 C = 2 + 3 + 7 = 12 B = 2 C = 2 + 3 = 0 D = 2 + 3 = 0 $E(T) = \frac{2 + 5}{2}$

Exploration

$$B = 2$$

$$C = 2 + 3 = 5$$

$$D = 2 + 3 + 7 = 12$$

$$\mathbb{E}(T) = \frac{2+5+12}{3} = 6.33$$

Search



$$C = 3$$

$$B = 3 + 3 = 6$$

$$D = 3 + 3 + 5 = 11$$

$$\mathbb{E}(T) = \frac{3+6+11}{3} = 6.66$$

ACBD is better

Exploration

B = 2 C = 2 + 3 + 7 = 12 B = 2 C = 2 + 3 = 5 D = 2 + 3 + 7 = 12 $E(T) = \frac{2 + 5 + 12}{2} = 6.33$



C = 3 B = 3 + 3 = 6 D = 3 + 3 + 5 = 11 $\mathbb{E}(T) = \frac{3 + 6 + 11}{3} = 6.66$

Search

ACBD is better

ABCD is better

Goal Assignment Strategies for Exploration

1. Greedy Assignment

Yamauchi B, Robotics and Autonomous Systems 29, 1999

Randomized greedy selection of the closest goal candidate

2. Broadcast of Local Eligibility

Werger B, Mataric M, Distributed Autonomous Robotic Systems 4, 2001 while any robot remains unassigned do

find the robot-goal pair (i, j) with the highest utility

assign the goal \boldsymbol{j} to the robot \boldsymbol{i} and remove them from the consideration

3. Hungarian Assignment

- Optimal solution of the task-allocation problem for assignment of n goals and m robots in O(n³)
- 4. K-means Clustering

Solanas A, Garcia M. A. IROS, 2004

Cluster an unknown space



Kuhn, 1955

Evaluation Methodology

Experimental setup

- 4,6,8 robots, 4 goal-assignment strategies, 4 environments, 10-30 runs
- sensor range: 5 m, FOV: 270°, SND driver
- ▶ planning period: 1 sec (speeded up simulation ~→ 3 sec)
- CPU 4x3.3GHz, 8GB RAM, x86_64 GNU/Linux kubuntu 3.0.0-20, ROS electric, polygon-based mapping
- total number of runs: 700

Performance metrics

$$T_f = \mathbb{E}(T|\mathcal{R}) = \sum_{t=0}^{\infty} t \frac{A_t^{\mathcal{R}}}{A_{total}} \approx \sum_{t=0}^{\infty} t A^{\mathcal{R}_t}$$

Results

Empty map 50 \times 50 m







Results

Hospital-section map 138×110.75 m



Conclusion

- The problem of multi robot search in an unknown space formulated.
- Several distance-cost-only strategies presented and statistically evaluated.
- Sophisticated methods outperformed the simple ones.
- Hungarian approach seems to be best (statistical significance not evaluated).
- Incorporating gain of visiting a goal will be interesting.
- Do we need a methods designed especially for multi-robot search?

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Thank You your attention!



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