Comparison of Classical and Interactive Multi-Robot Exploration Strategies in Populated Environments

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Multi-robot exploration Context

Formalization Classification

What is multi-robot exploration?

Definition

The Multi-robot exploration aims to design <u>efficient robots control</u> for <u>accurately reconstructing an unknown environment</u>.

Efficient control and accurate reconstruction

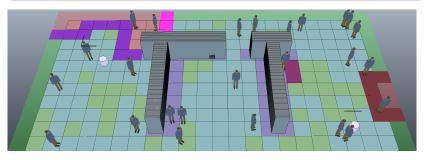
- <u>Control efficiency</u> is addressed at several levels (coverage, time, distance, energy, overlapping, ...)
- <u>Reconstruction accuracy</u> is the degree of closeness to the ground truth

Multi-robot exploration Context Formalization Classification

Populated environments

We investigate

- <u>human-aware</u> exploration,
- how can human presence help to explore dynamic environments?



Multi-Agent System simulated in V-REP [Rohmer et al., 2013]

Multi-robot exploration Context Formalization Classification

Multi-agent system formalization

Formally let...

- \mathcal{E} be an environment
- $\mathcal{R} = \{R_1, .., R_n\}$ be a set of robots
- $\mathcal{H} = \{H_1, ..., H_m\}$ be a set of humans

And for exploration...

- $\mathcal{O}_i^t \subset \mathcal{E}$ be \mathcal{R}_i 's observation at time t
- $\theta_i^{0:t} = \theta_i^{0:t-1} \cup \mathcal{O}_i^t$ be \mathcal{R}_i 's local history

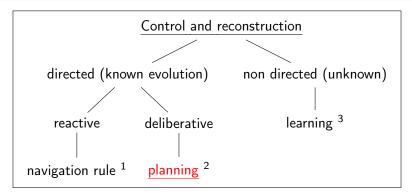
Exploration terminates when...

• $\Theta^{0:t} = igcup_{i=1}^n heta_i^{0:t}$ be the global history

•
$$\nexists \mathcal{O}_i^{t+1} \not\subset \Theta^{0:t}$$

Multi-robot exploration Context Formalization Classification

Multi-Robot Exploration



MRE classification example

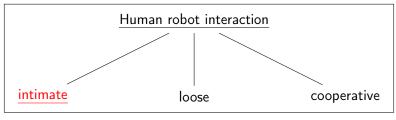
¹[Baronov and Baillieul, 2007, Morlok and Gini, 2007]
 ²[Yamauchi, 1997, Faigl et al., 2012, Bautin et al., 2012, Burgard et al., 2005, Macedo and Cardoso, 2004, Moorehead et al., 2001]
 ³?

Multi-robot exploration Context Formalization Classification

Human-robot Interaction

Definition ⁴

The HRI problem is to understand and shape the interactions between one or more humans and one or more robots.



HRI classification ⁵

⁴[Goodrich and Schultz, 2007] ⁵[Takeda et al., 1997] Problem definition Task Allocation Experimentation Conclusion Classification



Our work bridges together <u>Multi-Robot Exploration Planning</u> and <u>Human-Robot Intimate Interaction</u> into a task allocation framework.

Framework Tasks Costs

Definition

Robots, tasks, costs, assignments

- $\bullet \ \mathcal{R}$ be a set of robots
- $\bullet \ \mathcal{T}$ be a set of tasks
- $c_{\mathcal{R}_i \mathcal{T}_j}$ be the cost for \mathcal{R}_i to accomplish \mathcal{T}_j

•
$$a_{\mathcal{R}_i \mathcal{T}_j} = \begin{cases} 1 \text{ if } \mathcal{R}_i \text{ must accomplish } \mathcal{T}_j \\ 0 \text{ otherwise} \end{cases}$$

$$\begin{array}{c|c} c_{\mathcal{R}_{i}\mathcal{T}_{j}} & \mathcal{T} \\ \hline \mathcal{R} & \mathcal{C}_{\mathcal{R}\mathcal{T}} \end{array} \xrightarrow{\text{opt.}} & \overrightarrow{\mathcal{R}} & \mathcal{A}_{\mathcal{R}\mathcal{T}} \end{array}$$

Framework **Tasks** Costs

Our approach: Mixed Exploration

We consider the following tasks/targets

- frontiers to reach
- humans to interact with (opening doors, etc.)

Frontier based ${}^6 \ \mathcal{F} \subset \mathcal{T}$

A frontier is the observed boundary between explored and unexplored space.

Interaction based 7 $\mathcal{H}\subset\mathcal{T}$

Human-robot interaction is defined as the reciprocal influence between a human and a robot, followed by one or more effects.

⁶[Yamauchi, 1997] ⁷[Kaldé et al., 2014]

Framework Tasks **Costs**

Positive interaction

What kind of interaction takes advantage of human knowledge in populated environment?

Assumption : Humans have a natural adaptive navigation heuristic. Interaction : Robots can interact implicitly by following humans

How to define a human-robot interaction cost to speed up exploration?

Framework Tasks **Costs**

Cost formula

Cost formula $f : \mathcal{R} \times \mathcal{T} \to \mathbb{R}$

f combines target inactivity time t, distance to target d and reorientation to target o.

$$f(x, y) = \alpha \cdot g(x, y) + (1 - \alpha) \cdot h(x, y)$$

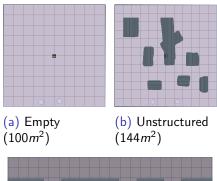
$$g(x, y) = d$$

$$h(x, y) = \begin{cases} \sigma \cdot (t + o) & \text{if } y \in \mathcal{F} \\ (1 - \sigma) \cdot (t + o) & \text{if } y \in \mathcal{H} \end{cases}$$

 $\alpha \in [0, 1]$, weight for immediate costs g and penalty heuristic $h \sigma \in [0, 1]$, weight for frontier or interaction penalties

Protocol Evaluation Results

Test environments





(c) Structured $(242m^2)$

Protocol Evaluation Results

Parameters

Parameters are as follows:

- Human density (% of env.): [0,30]
- Robot range of view: 2m
- Costs optimization strategy:
 - individual greedy
 - group greedy
- Modulators: $(\alpha, \sigma) \in [0, 0.25, 0.5, 0.75, 1]^2$

Protocol Evaluation Results

Metrics

Multi Robot Exploration metrics

Each scenario is evaluated with classical MRE metrics:

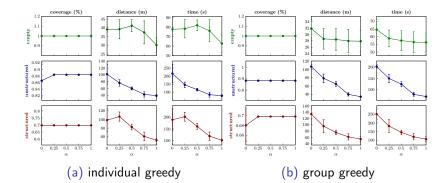
- coverage,
- distance,
- time,
- and number of allocations.

HRI metric [Olsen and Goodrich, 2003, Steinfeld et al., 2006].

We use a common metric in HRI, called the 'Robotic Attention Demand' (RAD). Here we consider the number of interactions initiated during exploration.

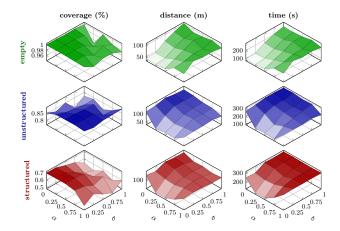
Protocol Evaluation Results

Results in non-populated environment



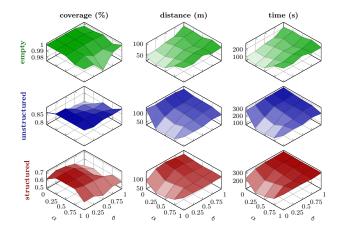
Protocol Evaluation Results

Results in populated environment 1/2 (individual greedy)



Protocol Evaluation Results

Results in populated environment 2/2 (group greedy)



Conclusion and perspectives

Conclusion

- our heuristic can improve exploration performances,
- our cost function cannot promote human-robot interactions.

Perspectives

- improve cost function to promote interactions,
- robot-(robot/object) interactions,
- perform real life experiments,
- learn to adapt exploration,
- dynamic parameter tuning.

Thank you for your attention.

References I

Baronov, D. and Baillieul, J. (2007). Reactive exploration through following isolines in a potential field. In <u>Proceedings of the American Control Conference</u>.



Bautin, A., Simonin, O., and Charpillet, F. (2012). MinPos : A Novel Frontier Allocation Algorithm for Multi-robot Exploration. In Proceedings of the 5th International Conference on Intelligent Robotics and Applications.



Burgard, W., Moors, M., Stachniss, C., and Schneider, F. E. (2005). Coordinated multi-robot exploration. Robotics, IEEE Transactions on.



Faigl, J., Kulich, M., and Preucil, L. (2012).Goal assignment using distance cost in multi-robot exploration.In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems.



Goodrich, M. A. and Schultz, A. C. (2007). Human-robot interaction: a survey. Foundations and trends in human-computer interaction.

References II



Kaldé, N., Charpillet, F., and Simonin, O. (2014).

Comparaison de stratégies d'exploration multi-robot classiques et interactives en environnement peuplé.

In JFSMA - Journées Francophones sur les Systèmes Multi-Agents, Loriol-sur-Drôme, France. Cépaduès. [accepté].



Macedo, L. and Cardoso, A. (2004).

Exploration of unknown environments with motivational agents. In Proceedings of the Third International Joint Conference on Autonomous Agents and Multiagent Systems.



Moorehead, S. J., Simmons, R., and Whittaker, W. L. (2001). Autonomous exploration using multiple sources of information. In Proceedings of the IEEE International Conference on Robotics and Automation.



Morlok, R. and Gini, M. (2007). Dispersing robots in an unknown environment. In Distributed Autonomous Robotic Systems. Problem definition Task Allocation Conclusion

References III

Olsen, D. R. and Goodrich, M. A. (2003). Metrics for evaluating human-robot interactions. In <u>Proceedings of PERMIS</u> .
Rohmer, E., Singh, S. P., and Freese, M. (2013).
V-rep: A versatile and scalable robot simulation framework.
In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots
and Systems.
Steinfeld, A., Fong, T., Kaber, D., Lewis, M., Scholtz, J., Schultz, A., and
Goodrich, M. (2006).
Common metrics for human-robot interaction.
In Proceedings of the 1st ACM SIGCHI/SIGART conference on Human-robot
interaction.
Takeda, H., Kobayashi, N., Matsubara, Y., and Nishida, T. (1997).
Towards ubiquitous human-robot interaction.
In Proceedings of the Working Notes for LICAL Workshop on Intelligent

In Proceedings of the Working Notes for IJCAI Workshop on Intelligent Multimodal Systems.

References IV



Yamauchi, B. (1997). A frontier-based approach for autonomous exploration. In <u>Proceedings of the IEEE International Symposium on Computational</u> Intelligence in Robotics and Automation.