

Multi-Robot Path Planning with Due Times

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Introduction: Multi-Robot Path Planning

- multi-robot path planning (MRPP)¹ or multi-agent path finding (MAPF)² is an enabling research field for practical deployment of multi-robot systems



(a)



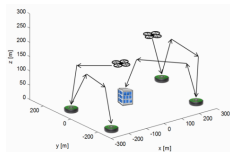
(b)



(c)



(d)



(e)

¹Lynne E. Parker. "Multiple Mobile Robot Teams, Path Planning and Motion Coordination in". In: *Encyclopedia of Complexity and Systems Science* (2009), pp. 5783–5800. DOI: 10.1007/978-0-387-30440-3_344.

²Roni Stern. "Multi-Agent Path Finding – An Overview". In: *Artificial Intelligence*. Springer International Publishing, 2019, pp. 96–115. doi: 10.1007/978-3-030-33274-7_6.

Introduction: Solutions

- MRPP or MAPF can be solved by various methodologies:
 - ▶ reduction-based (SAT, ILP, AST...)
 - ▶ search-based, optimal (CBS, ICBS, M*...)
 - ▶ search-based, bounded suboptimal (ECBS, CBS-HWY...)
 - ▶ search-based, unbounded suboptimal (HCA*, WHCA*, co-WHCA*...)
 - ▶ rule-based (Push&Swap, Push&Rotate, traffic rules...)
 - ▶ decoupled (prioritized planning, path coordination...)
 - ▶ sensor-based or reactive (ORCA...)
 - ▶ learning-based (PRIMAL, PRIMAL2, MAGAT...)
 - ▶ ...

Motivation

- in practice, multi-robot systems have to accomplish tasks with due times can be anticipated
- multi-robot path planning with time and distance-based objectives:³
 - ▶ makespan
 - ▶ total travel times
 - ▶ maximum distance
 - ▶ total distances
- multi-agent path finding with deadline-based objectives:⁴
 - ▶ the number of successful agents
- none of existing problem formulations could reflect soft temporal constraints of tasks⁵

³Jingjin Yu and Steven M. LaValle. "Optimal Multirobot Path Planning on Graphs: Complete Algorithms and Effective Heuristics". In: *IEEE Transactions on Robotics* 32.5 (2016), pp. 1163–1177. ISSN: 15523098. DOI: 10.1109/TR0.2016.2593448. arXiv: 1507.03290.

⁴Hang Ma et al. "Multi-agent path finding with deadlines". In: *IJCAI International Joint Conference on Artificial Intelligence* 2018-July (2018), pp. 417–423. ISSN: 10450823.

⁵Ernesto Nunes et al. "A taxonomy for task allocation problems with temporal and ordering constraints". In: *Robotics and Autonomous Systems* 90 (2017), pp. 55–70. ISSN: 09218890. DOI: 10.1016/j.robot.2016.10.008.

Problem Formulation: MRPP-DT

- inputs:

- ▶ underlying graph $G = (V, E)$
- ▶ robot set R composed of n robots
- ▶ start vertices V_s
- ▶ goal vertices V_g
- ▶ due times D

- outputs:

- ▶ n feasible collision-free paths $P = \{p_1, \dots, p_n\}$

- three due time-related objectives:

- ▶ minimization of maximum lateness L_{max}
- ▶ minimization of total tardiness $\sum_{i=1}^n H_i$
- ▶ minimization of total unit penalties $\sum_{i=1}^n U_i$

Examples and Problem Analysis

Problem: Multi-Robot Path Planning with Due Times

Given (G, R, V_s, V_g, D) as is described previously, find an optimal solution $P = \{p_1, \dots, p_n\}$ composed of n feasible collision-free paths for all robots with objective 1, 2, or 3.

Intractability

The MRPP-DT problem is NP-hard.

Pareto Structure

Any two of seven objectives cannot be optimized simultaneously.

Examples

Examples

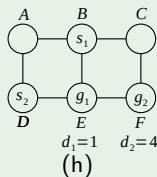
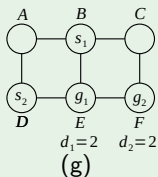
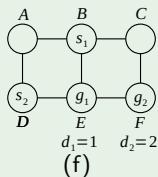
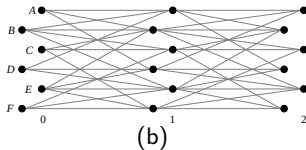
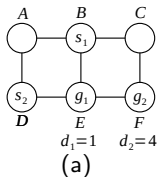


Figure: Three examples with the same start and goal vertices, but different due times. The underlying graph has 6 vertices.

Reduction-based Algorithms



$$\min : t_{max}$$

$$\min : nT - \sum_{i=1}^n d_i - \sum_{i=1}^n \sum_{t=1}^T z_i^t y_i^t$$

$$\min : n - \sum_{i=1}^n \sum_{e_j \in \delta^-(v_i)} x_{i,j}, v_t \in S^*$$

(c)

Completeness Theorem

MINMAXLAT, MINTOTALTARD, MINTOTALUP are all complete.

Instance Design

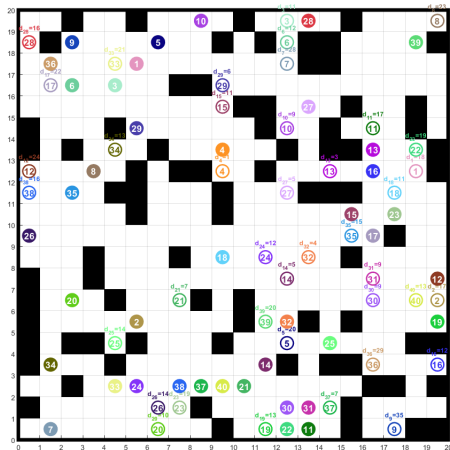


Figure: An MRPP-DT instance with 40 robots on a $20 \times 20 \times 20\%$ grid map.

Results of Objective 1

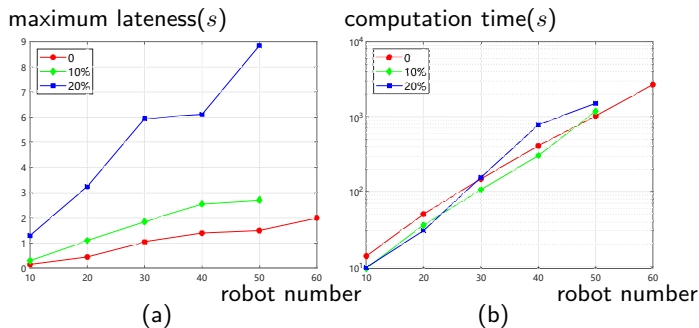


Figure: Performance of MINMAXLAT over MRPP-DT instances.

Results of Objective 2

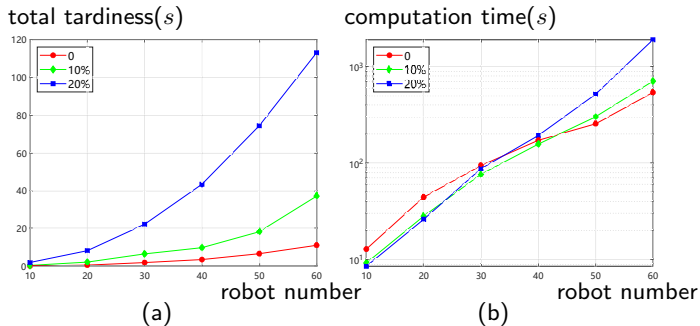
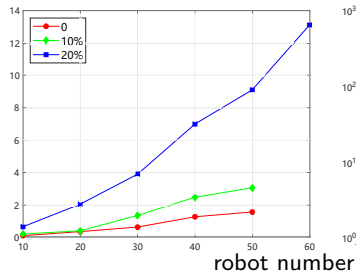


Figure: Performance of MINTOTALTARD over MRPP-DT instances.

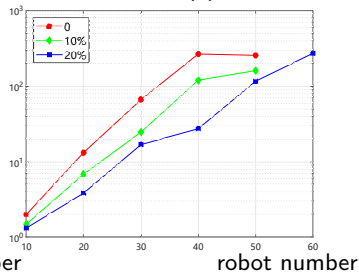
Results of Objective 3

total unit penalties



(a)

computation time(s)



(b)

Figure: Performance of MINTOTALUP over MRPP-DT instances.

Conclusions and Future Work

- conclusions:

- ▶ MRPP-DT is formulated to address task awareness in path coordination level
- ▶ complete reduction-based algorithms are developed
- ▶ proposed algorithms can solve small-scale problems in allocated time, with good optimality performance

- future work:

- ▶ develop fast algorithms
- ▶ apply related ideas in practical applications, i.e., logistics
- ▶ combined task and path planning

Key References

- [1] Lynne E. Parker. “Multiple Mobile Robot Teams, Path Planning and Motion Coordination in”. In: *Encyclopedia of Complexity and Systems Science* (2009), pp. 5783–5800. DOI: 10.1007/978-0-387-30440-3_344.
- [2] Roni Stern. “Multi-Agent Path Finding – An Overview”. In: *Artificial Intelligence*. Springer International Publishing, 2019, pp. 96–115. DOI: 10.1007/978-3-030-33274-7_6.
- [3] Jingjin Yu and Steven M. LaValle. “Optimal Multirobot Path Planning on Graphs: Complete Algorithms and Effective Heuristics”. In: *IEEE Transactions on Robotics* 32.5 (2016), pp. 1163–1177. ISSN: 15523098. DOI: 10.1109/TR0.2016.2593448. arXiv: 1507.03290.
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- [5] Ernesto Nunes et al. “A taxonomy for task allocation problems with temporal and ordering constraints”. In: *Robotics and Autonomous Systems* 90 (2017), pp. 55–70. ISSN: 09218890. DOI: 10.1016/j.robot.2016.10.008.