

# Motion Planning for Unlabeled Discs with Optimality Guarantees

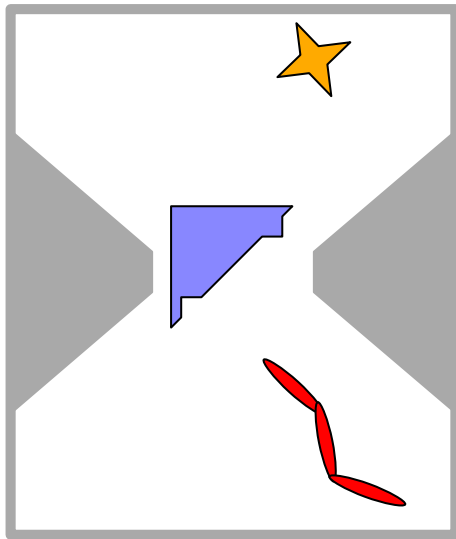
Kiril Solovey

Tel Aviv University, Israel

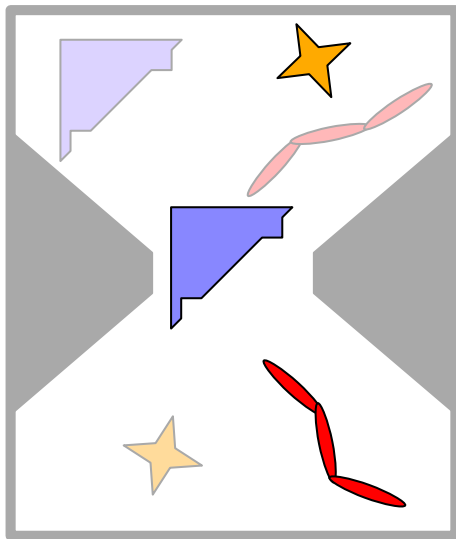
Israeli Conference on Robotics, 2016

\* Joint work with Jingjin Yu, Or Zamir and Dan Halperin

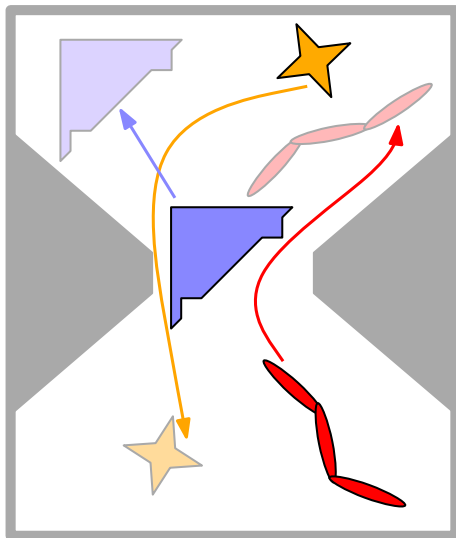
# Multi-robot motion planning



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## Complete combinatorial techniques, related work

Planning using exact analytic methods, often using an explicit construction of the free space.

- Fixed number of robots:
  - ▶ Two discs in  $O(n^3)$  [Schwartz and Sharir, 83]
  - ▶ Two discs in  $O(n^2)$  [Yap, 83]
  - ▶ Two discs (and other types) in  $O(n^2)$  [Sharir and Sifrony, 91]

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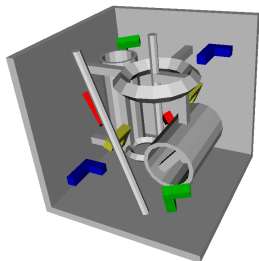
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- Multiple robots:
  - ▶ PSPACE-hardness for translating rectangles [Hopcroft et al., 84], [Hearn and Demaine, 05]
  - ▶ NP-hardness for discs [Spirakis and Yap, 84]

# Sampling-based methods for multi-robot, related work

Capturing the connectivity of the free space by random sampling.

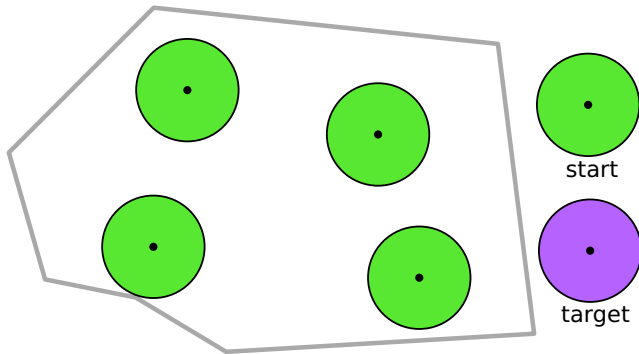
- Sampling-based planners that are tailored for the multi-robot case have been proposed [Švestka and Overmars, 98], [van den Berg et al., 09], [Wagner and Choset, 11], [S. et al., 14] ...
- Sampling-based planners can cope with complex types of scenarios and robots
- However, they provide only (weak) asymptotic guarantees



## Unlabeled multi-robot motion planning

A variant of the multi-robot problem in which the robots are **identical** and **indistinguishable** [Kloder and Hutchinson, 06].

**Goal:** Move the robots such that each target will be occupied by *some* robot, **while avoiding collisions**.

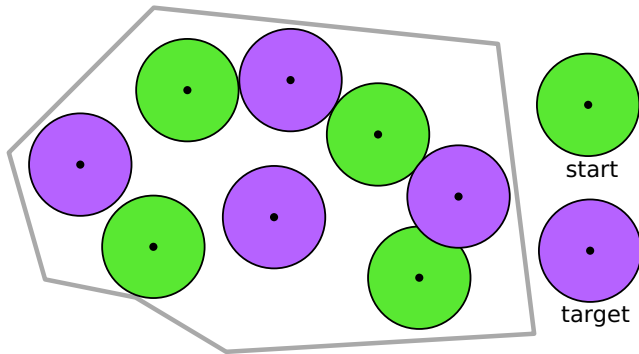




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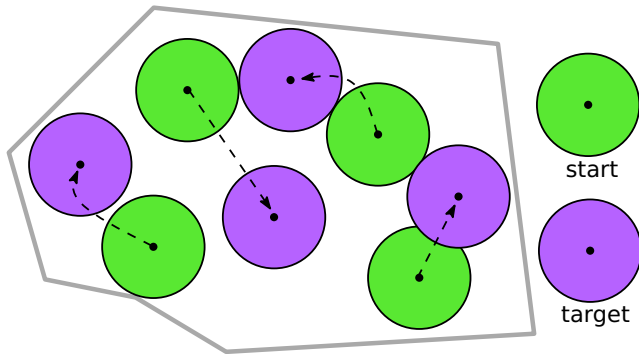
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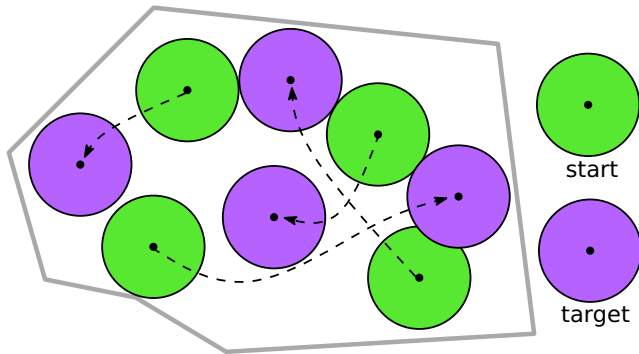
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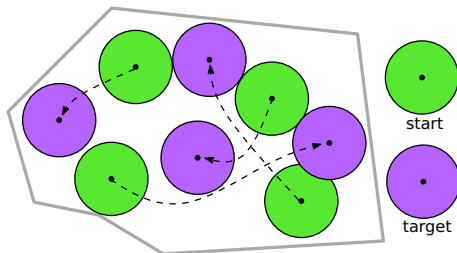


## Unlabeled multi-robot motion planning: related results

In general, the unlabeled problem is **PSPACE-hard** [S. & Halperin, 15].

However, by making certain **separation assumptions** the unlabeled problem can be solved efficiently. The following algorithms work in the setting of  $m$  unit-disc robots in an environment of complexity  $n$ :

- $O(m^2 + mn)$ -time algorithm [Adler et al., 14]
- $\tilde{O}(m^4 + m^2n)$ -time algorithm that returns the **optimal** solution in terms of the **longest path length** [Turpin et al., 13]



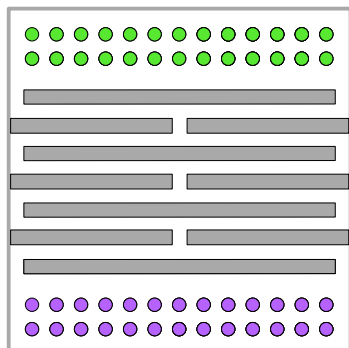
# Contribution

We consider the specific setting of unit-disc robot operating in a workspace cluttered with polygonal obstacles.

We develop an  $\tilde{O}(m^4 + m^2 n^2)$ -time algorithm, where  $m$  is the number of robots and  $n$  is the complexity of the workspace.

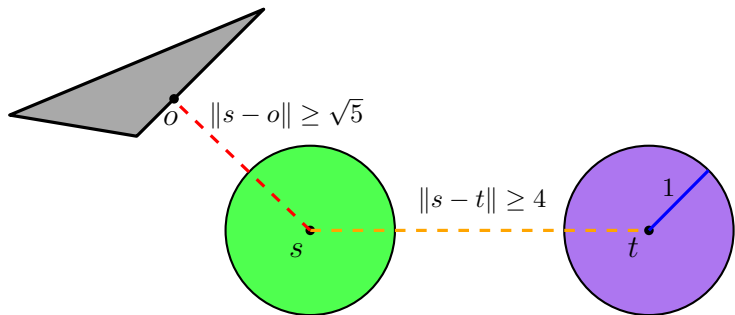
The algorithm **minimizes the total path length**, i.e., the sum of lengths of the individual paths of the robots.

Specifically, the algorithm is guaranteed to return a solution whose cost is at most  $\text{OPT} + O(m)$ .



## Our setting

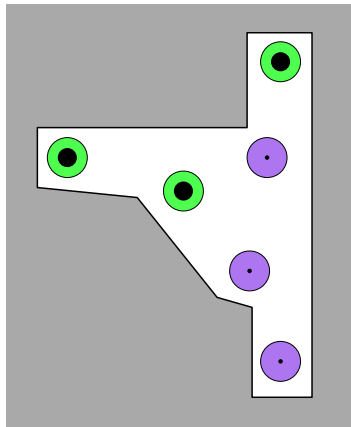
We study the unlabeled problem for **unit discs** moving amid **polygonal obstacles**, and make two simplifying assumptions:



## Standalone targets

The main ingredient in our algorithm is the following observation.

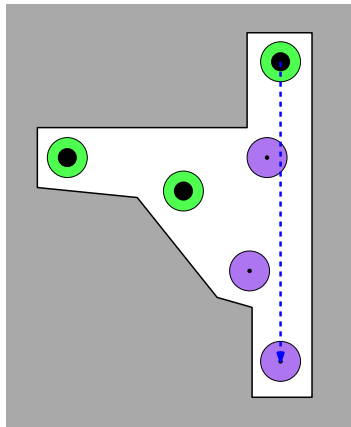
- Suppose that the robots are resting at their start positions
- One of the robots can be moved to one of the targets positions while the remaining robots stay put
- The other robots would still be able to reach their targets afterwards



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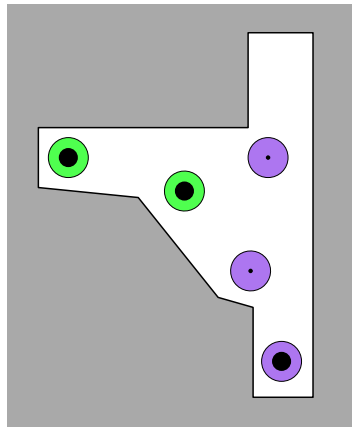




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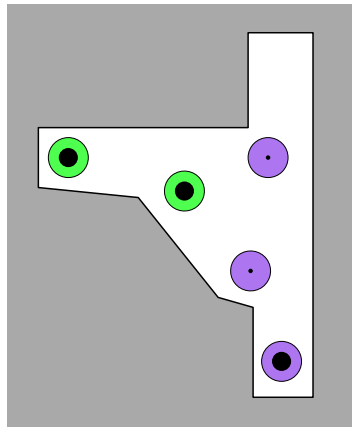
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## Algorithmic implications

We perform the following steps in each iteration of the algorithm:

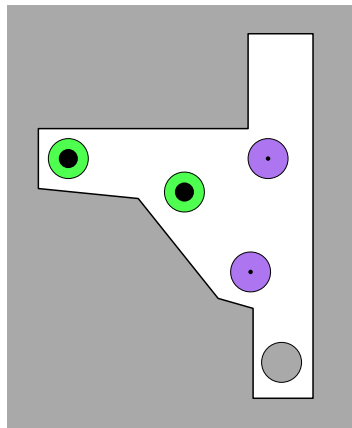
- 1 Find a robot that can be moved to a target without blocking consecutive moves and execute the motion
- 2 This robot will stay put for the rest of the execution, i.e., will be treated as an obstacle
- 3 Repeat



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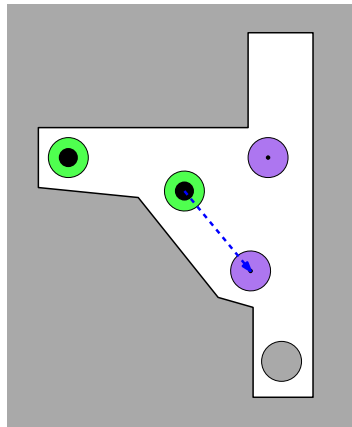
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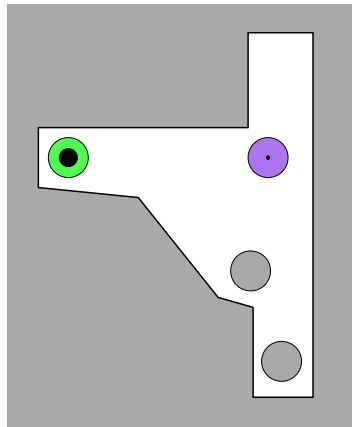
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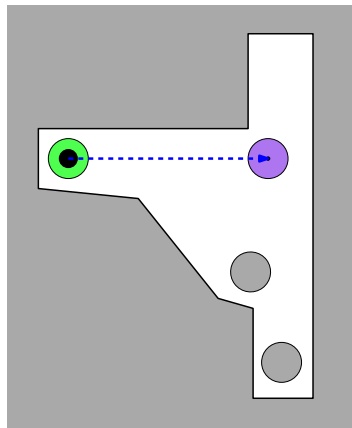
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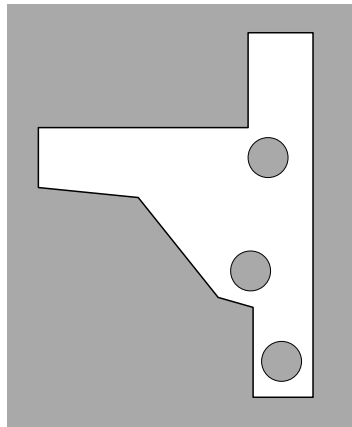
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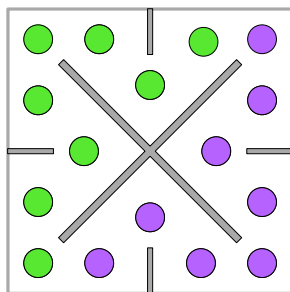
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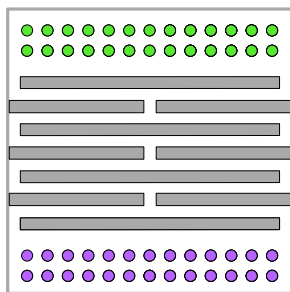


## Experimental results

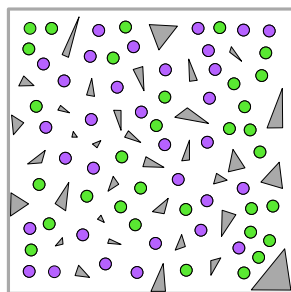
We implemented the algorithm in C++ with exact geometric methods that are provided by CGAL. As such, our implementation is complete, robust, deterministic and parameter free.



8 robots  
19 sec.



26 robots  
800 sec.



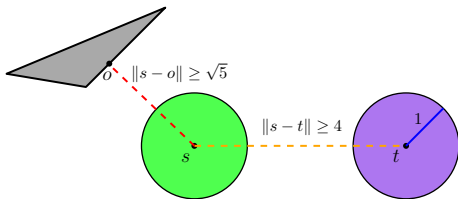
32 robots  
6000 sec.



# Motion Planning for Unlabeled Discs with Optimality Guarantees

Solovey, Yu, Zamir & Halperin

We study the unlabeled problem in the setting of **unit-disc** robots translating amid polygonal obstacles, with **separation constraints**:



We develop an  $\tilde{O}(m^4 + m^2 n^2)$ -time algorithm that returns a solution whose cost is at most  $\text{OPT} + 4m$ , in terms of the **total path length**.

# Thank you!