

# Multi-Robot Motion Planning

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## 1 Multi-Robot Motion Planning

- Introduction
- Decoupled Approach
- Centralized Approach

## 2 Centralized Path Planning for Multiple Robots

- Preliminaries
- Incremental Discovery of Coupling
- Results
- Project

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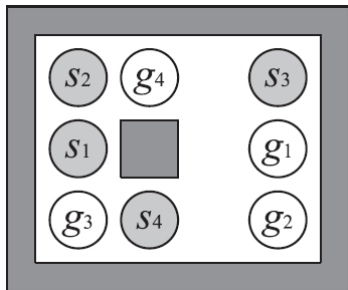
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# Main Approaches

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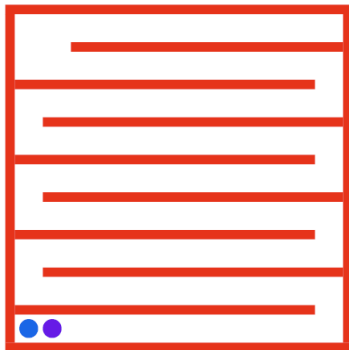
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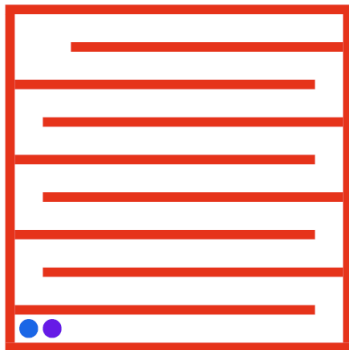
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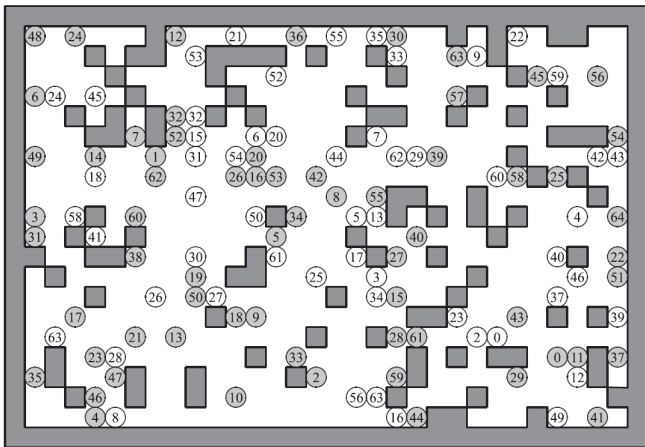
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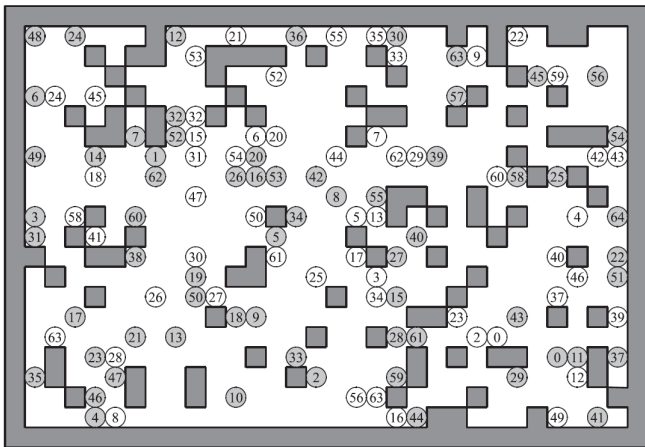
- The robots are treated as a group and sometimes as one composite robot.
- Usually yields a *complete* planner.
- Known techniques applicable only for small-scale problems with low degrees of freedom.

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# Introduction

## Centralized Path Planning for Multiple Robots: Optimal Decoupling into Sequential Plans

Jur van den Berg   Jack Snoeyink   Ming Lin   Dinesh Manocha

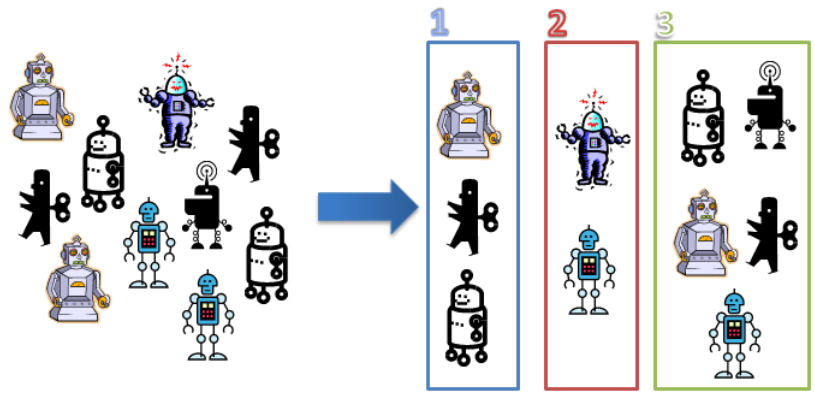
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- In these sub-problems, individual robots will be coupled and considered as composite robots.

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## Goal

Compute a path  $\pi : [0, 1] \rightarrow C(r_1) \times \dots \times C(r_n)$  such that initially  $\pi(0) = (s_1, \dots, s_n)$ , finally  $\pi(1) = (g_1, \dots, g_n)$ , and for every  $t \in (0, 1)$   $\pi(t)$  describes a collision free (obstacles, robots) position.

# Coupled Relation

## Definition

For a robot  $r_i$  the *Active Interval*  $\tau_i$  is defined as the open interval from the first time  $r_i$  leaves its start position to the last time it reaches its goal position.

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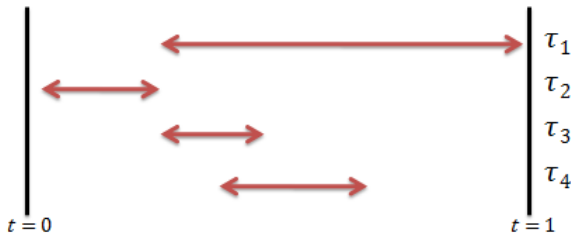
Two robots  $r_i, r_j$  are coupled if their active intervals intersect,  $\tau_i \cap \tau_j \neq \emptyset$ .

# Example

Consider the following active intervals for a given path  $\pi$ :

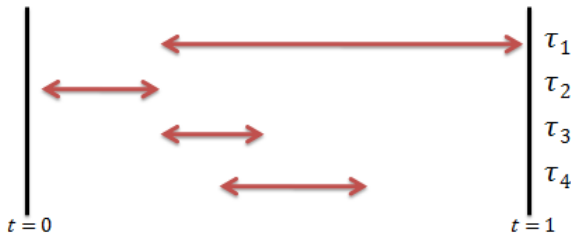
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The coupled robots are  $\{r_1, r_3\}$ ,  $\{r_1, r_4\}$ ,  $\{r_3, r_4\}$ .

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## Definition

*Execution Sequence* is an ordered partition of the  $n$  robots into a sequence  $S = (R_1, \dots, R_k)$  of composite robots, such that

- $R_1 \cup \dots \cup R_k = \{r_1, \dots, r_n\}$
- For every  $i \neq j$   $R_i \cap R_j = \emptyset$ .

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$S$  is *valid* if it describes the *coupled* relation of the robots in some solution path.

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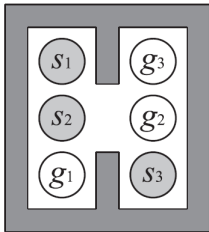
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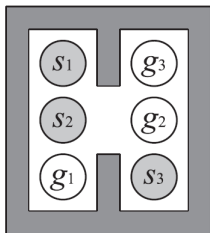
## Definition

$S$  is *optimal* if it is the solution sequence with the lowest degree of its largest *composite robot*.

# Example

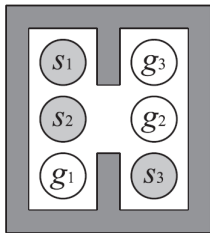


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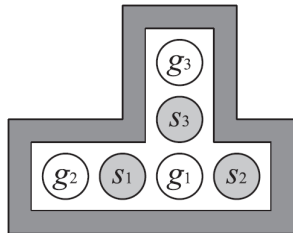


$$S = [r_3, r_2, r_1]$$

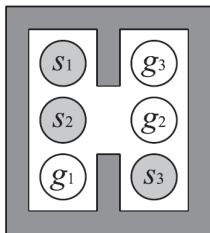
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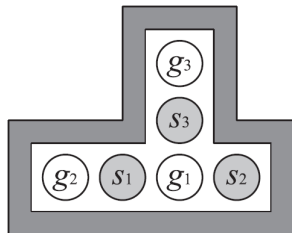
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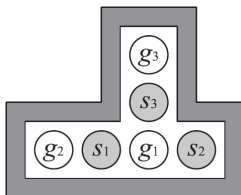
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For example,  $S = [r_3, r_1 r_2]$  is valid for the scene

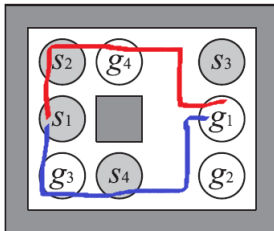


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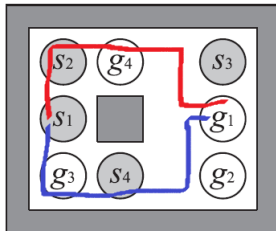
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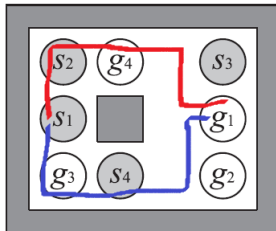
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- Assume we found the blue trajectory for  $r_1$  and we want to use it in the global solution.
- What will be the order of  $r_1, r_3, r_4$  in the solution sequence?

# Order Constraints

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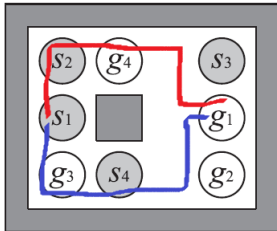
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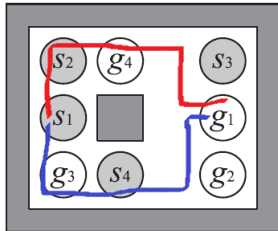
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# Example



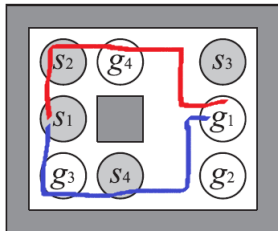
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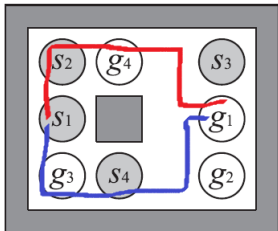
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And combined:

$$P(r_1) = (r_2 \prec r_1 \wedge r_1 \prec r_4) \vee (r_1 \prec r_3 \wedge r_4 \prec r_1)$$

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### Observation

*An execution sequence  $S = (R_1, \dots, R_k)$  is a solution sequence iff  $S$  satisfies the constraints expression  $P(R_1) \wedge \dots \wedge P(R_k)$ .*

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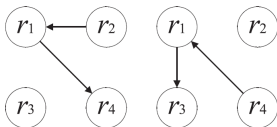
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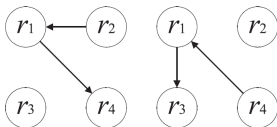
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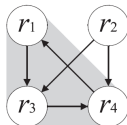


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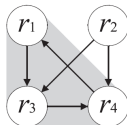
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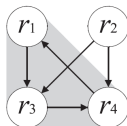
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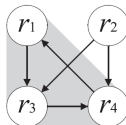


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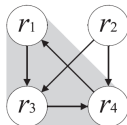
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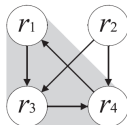
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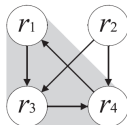
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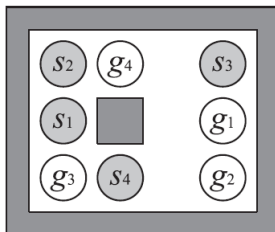


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- We stop when there exists  $J \in E$  for which all robots in  $S(J)$  have been planned.

# Incrementally Building the Execution Sequence: Example

We will run the algorithm on the problem:



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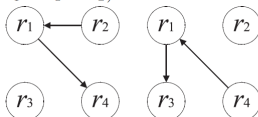


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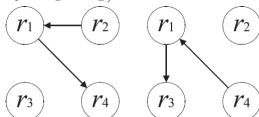


Iteration 2.  $L = \{r_1\}$ ,  $R_{\min} = r_2$ ,  $\mathcal{P}(r_2) = r_2 \prec r_4 \vee (r_1 \prec r_2 \wedge r_2 \prec r_3 \wedge r_4 \prec r_2)$ .

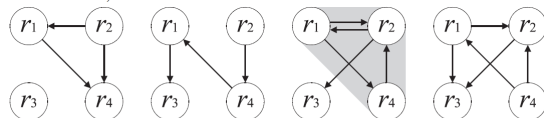
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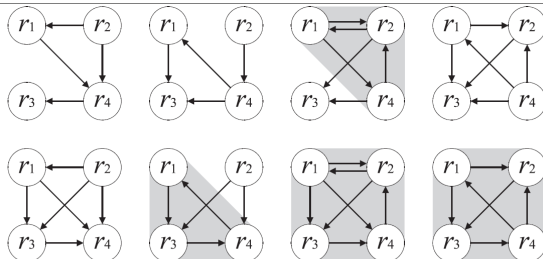


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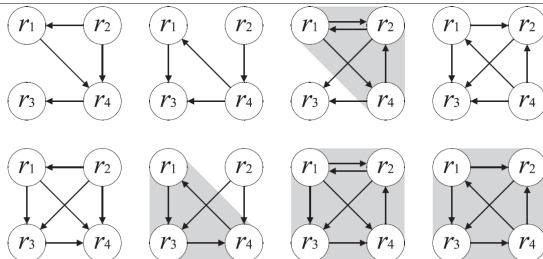
Iteration 3.  $L = \{r_1, r_2\}$ ,  $R_{\min} = r_3$ ,  $\mathcal{P}(r_3) = r_4 \prec r_3 \vee (r_3 \prec r_4 \wedge r_2 \prec r_3 \wedge r_1 \prec r_3)$ .

# Incrementally Building the Execution Sequence: Example



Iteration 4.  $L = \{r_1, r_2, r_3\}$ ,  $R_{\min} = r_4$ ,  $\mathcal{P}(r_4) = \top$ , so no change to graphs.

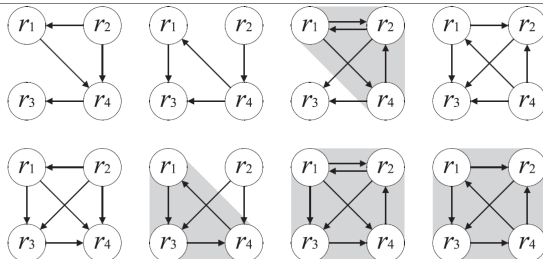
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- After the 4th iteration,  $L = \{r_1, r_2, r_3, r_4\}$ .

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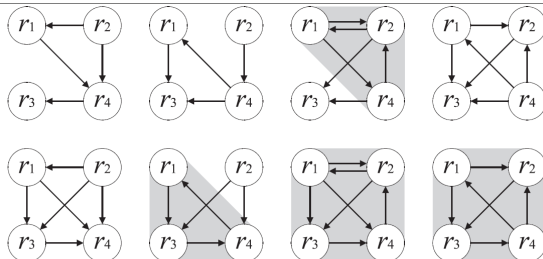


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- $J_1, J_2, J_4, J_5$  have all their SCC in  $L$ .



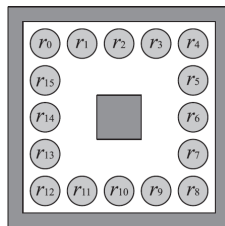
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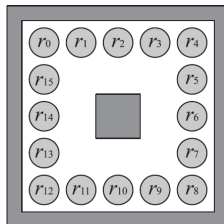
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- After the 4th iteration,  $L = \{r_1, r_2, r_3, r_4\}$ .
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- The algorithm terminates and returns  $S(J_1)$ .

# Scene 1

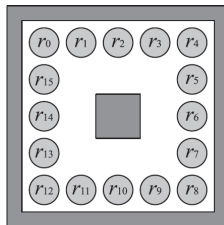


# Scene 1



- Each robot  $r_i$  changes position with  $r_{((i+8) \bmod 16)}$

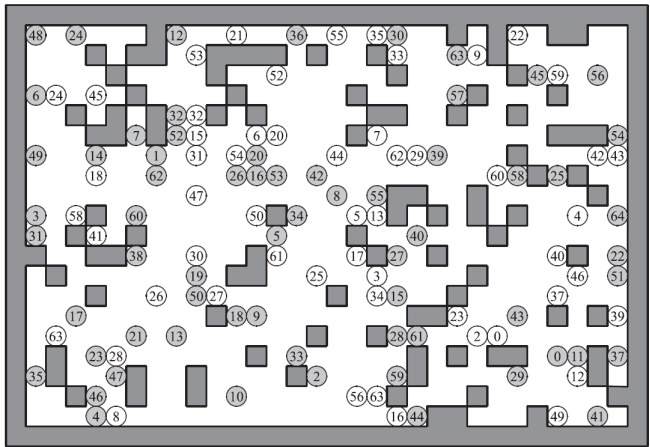
# Scene 1



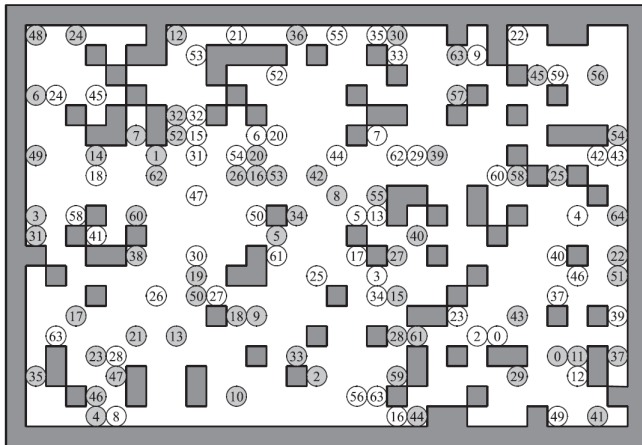
- Each robot  $r_i$  changes position with  $r_{((i+8) \bmod 16)}$
- The algorithm returned

$(r_0 r_7 r_8 r_{15}, r_1 r_9, r_2 r_{10}, r_3 r_{11}, r_4 r_5 r_{12} r_{13}, r_6 r_{14}, r_7 r_{15})$

# Scene 2

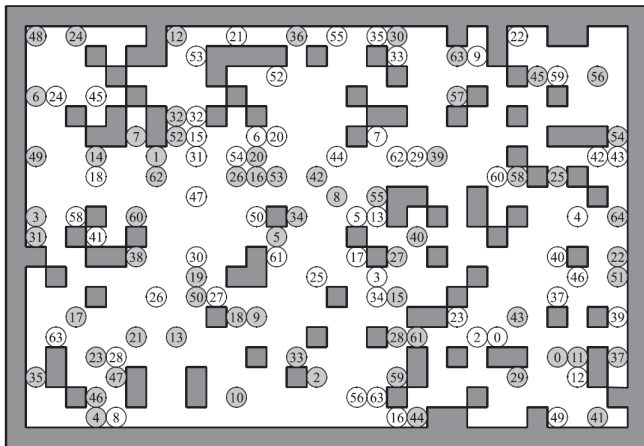


## Scene 2



- 65 robots

## Scene 2



- 65 robots
- Solution sequence involves only **individual** robots!

# Project

- Two possible settings, two possible CR Planners:



# Project

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  - ▶ Grid workspace: motion planning on graphs.

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  - ▶ Polygonal robots and obstacles (harder): sampling-based techniques.

# Project

- Two possible settings, two possible CR Planners:
  - ▶ Grid workspace: motion planning on graphs.
  - ▶ Polygonal robots and obstacles (harder): sampling-based techniques.
- In both cases, the main algorithm remains the same.

# The End

”... and the robots lived happily ever after.”