16-782 Planning & Decision-making in Robotics

Multi-Robot Planning

Maxim Likhachev
Robotics Institute
Carnegie Mellon University

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 - Decentralized: each robot decides/plans what to do on its own

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Challenges with decentralized planning?

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Challenges with decentralized planning?

How to guarantee that the overall team accomplishes its goal?

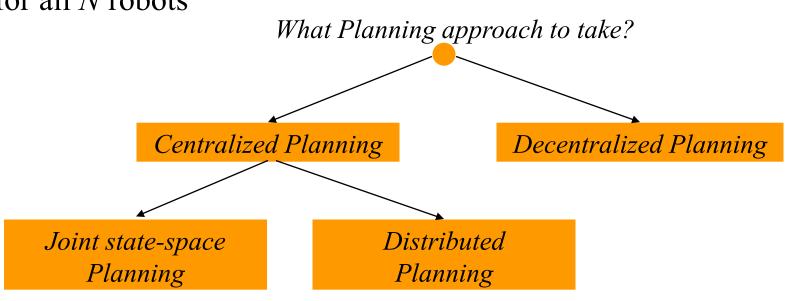
- Multi-robot Path Planning vs. Multi-robot Cooperative Task Planning
 - Multi-robot Path Planning: how to plan paths for N robots so that they don't collide with each other during execution
 - Multi-robot Cooperative Task Planning: how to compute plans for N robots so that they achieve the overall goal that may require cooperation

- Small teams vs. large teams (swarms) of robots
 - Planning for small teams: Compute plans for N (potentially heterogeneous) robots, where N is typically 2-10
 - Planning for (control of) swarms of robots: how to control a swarm of N (usually homogeneous) robots, where N is typically 10-1000

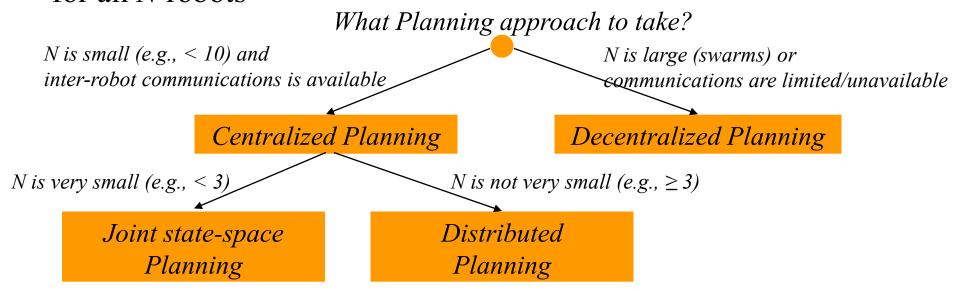
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Control of swarms is typically decentralized

- Joint state-space vs. distributed planning (within centralized)
 - Joint state-space planning: Planning for N robots in a state-space that represents joint configurations of robots
 - **Distributed planning**: Planning is split into N individual planners that share their results (and potentially re-plan) to obtain a final plan for all N robots

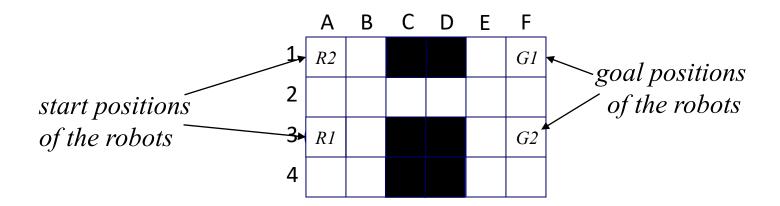


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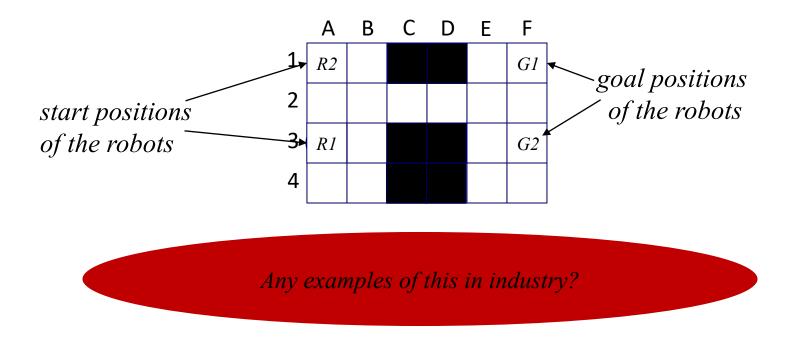
• Path planning for N robots to get to their goals w/o collisions

simple example for two omnidirectional point-size robots



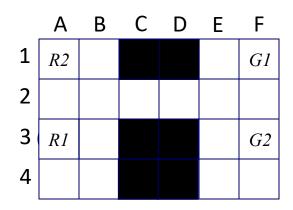
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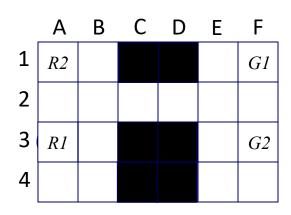
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Joint state-space planning



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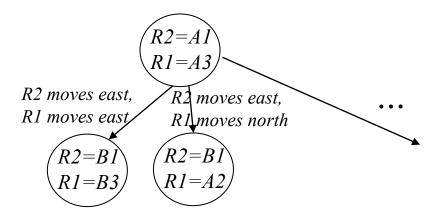
Joint state-space planning

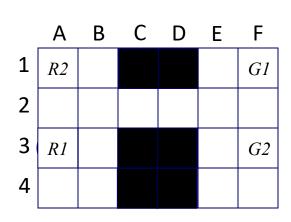


The simplest approach: construct and search a graph, where each state encodes positions of all the robots and each action encodes all possible movements

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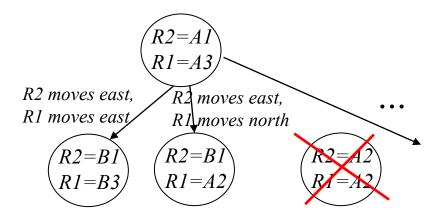


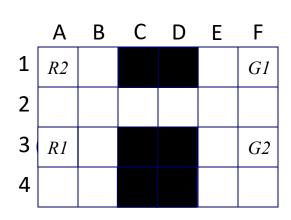


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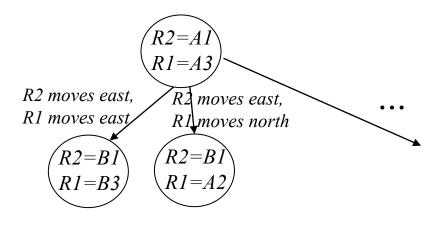


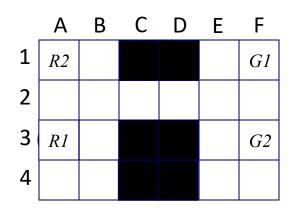


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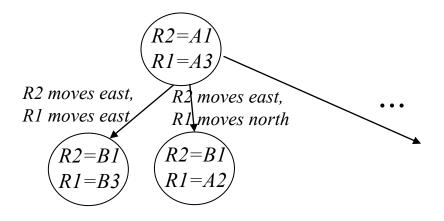
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$$(R2=F3)$$
 goal state

• Path planning for N robots to get to their goals w/o collisions

Assuming 4-connected grid, what is the maximum branching factor (how many actions/successors)?

Joint state-space planning



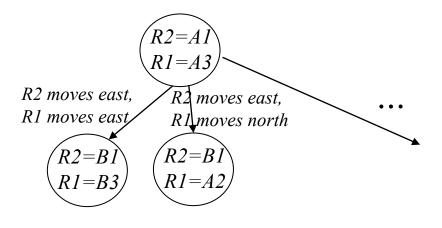
	Α	В	С	D	Ε	F
1	R2					G1
2						
3	<i>R1</i>					G2
4						

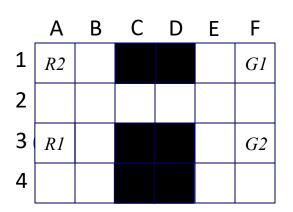
$$\binom{R2=F3}{RI=FI}$$
 goal state

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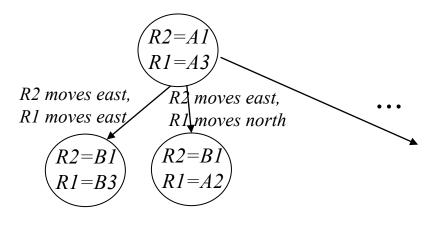


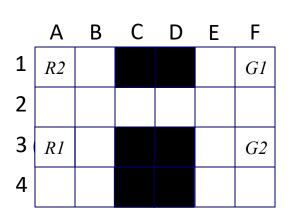
 $\binom{R2=F3}{RI=FI}$ goal state

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Joint state-space planning

What is the size of the graph?





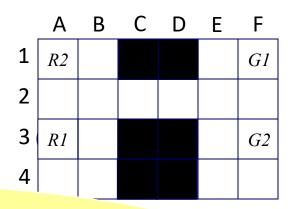
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Scalability w.r.t. N is clearly an issue!

$$(R2=F3)$$
 goal state

• Path planning for N robots to get to their goals w/o collisions

Distributed planning

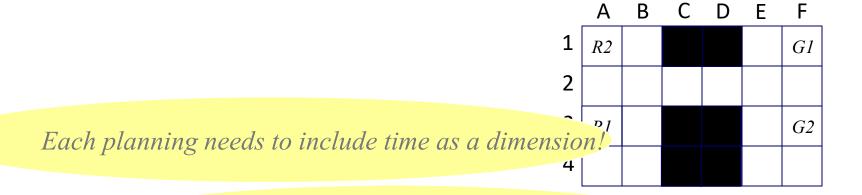


One popular approach: Prioritized Planning
For i = 1:N

Compute path for robot R_i that avoids collisions with paths for robots $R_1..R_{i-1}$

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Distributed planning

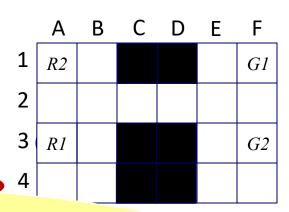


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What will be the plan returned by Prioritized Planning?

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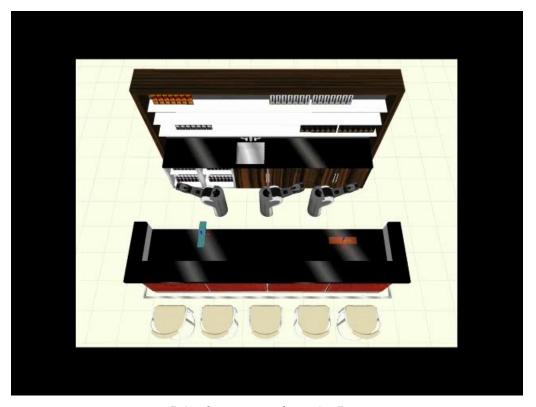
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Distributed planning

complete & optimal approach Conflict-based Search (CBS) $C(Sstart) = \{\}; //no\ constraints\ on\ paths\ at\ the\ start\ state$ $OPEN = \{Sstart\}; g(Sstart) = 0;$ while $(OPEN \neq 0)$ remove S with the smallest g-value from OPEN if no collisions between paths associated with S, then return them as the overall plan for each (or at least one) collision of robots Ri and Rj at vertex v at time t $C1 = C(S) \cup \{Ri(t) \neq v\} / Ri \ can't \ be \ at \ vertex \ v \ at \ time \ t$ compute N paths that satisfy C1 constraints (w/o collision-checking between paths) and the overall cost PLANCOST1 $C2 = C(S) \cup \{Rj(t) \neq v\} / Rj \ can't \ be \ at \ vertex \ v \ at \ time \ t$ compute N paths that satisfy C2 constraints (w/o collision-checking between paths) and the overall cost PLANCOST2 insert state S1 with C(S1) = C1 and g(S1) = PLANCOST1insert state S2 with C(S2) = C2 and g(S2) = PLANCOST2

• Example: planning for N robotic arms to move an object



[Cohen et al., '14] (performs joint state-space planning)

• Example: planning for N robotic arms to move an object



(planning is distributed: plan on Roman platform first, then on PR2)

• Example: planning for multi-robot exploration/mapping

N robots need to explore and build a map of unknown environment

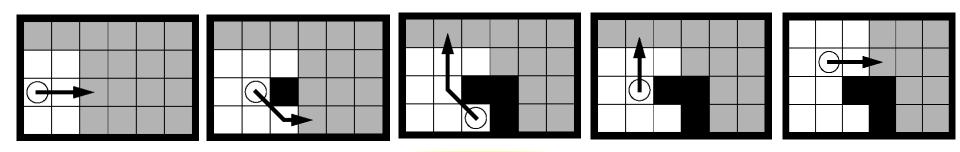
One approach: Distributed Greedy Mapping

For i = 1:N

• Example: planning for multi-robot exploration/mapping

N robots need to explore and build a map of unknown environment

- Greedy Mapping for a single robot:
 - always move the robot on a shortest path to the closest unobserved (or unvisited) cell
 - it always achieves a gain in information.
 - thus, it is guaranteed to map the environment that is reachable (assuming all moves are reversible)



One approach: Distributed Greedy Mapping

For i = 1:N

• Example: planning for multi-robot exploration/mapping

N robots need to explore and build a map of unknown environment



[Butzke et al., '11]

One approach: Distributed Greedy Mapping For i = 1:N

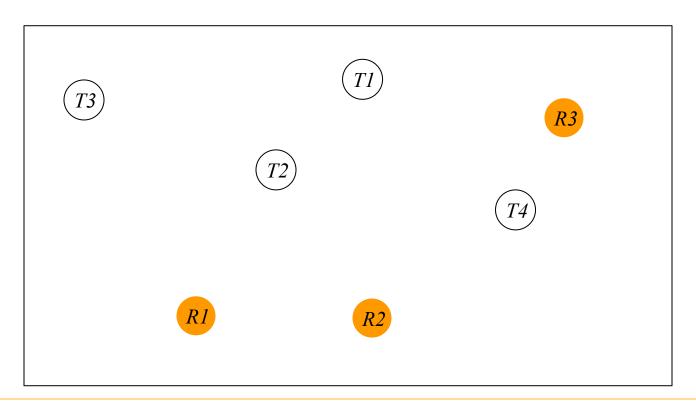
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N robots need to explore and build a map of unknown environment



[Kusner et al., '19]
One approach: Distributed Greedy Mapping
For i = 1:N

- Market-based approach (very popular distributed approach)
 - Consider planning the allocation of tasks to N robots
 - General scheme: robots auction out their tasks to their teammates with the goal of increasing their own revenue

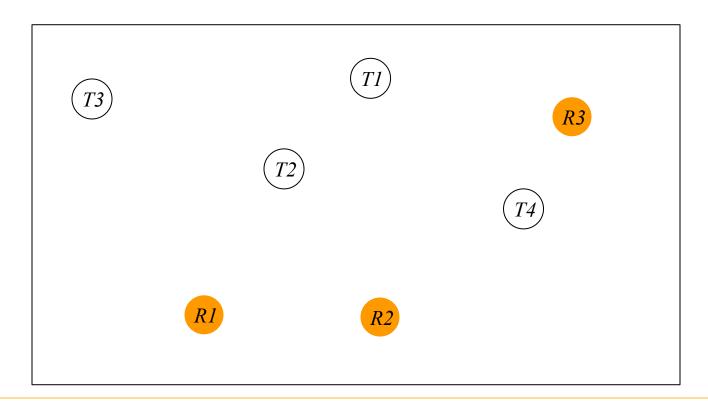


Market-based Approach

Given N robots $R_1...R_N$, M tasks $T_1...T_M$, and C_i^{Rj} – cost of executing task i by robot R_j (cost may depend on other tasks executed by this robot)

Planner needs to decide: Which task gets executed by which robot?

Find a plan(mapping) π^* : $T_i \rightarrow R_j$ such that $\pi^* = argmin \sum C_i^{\pi(T_i)}$



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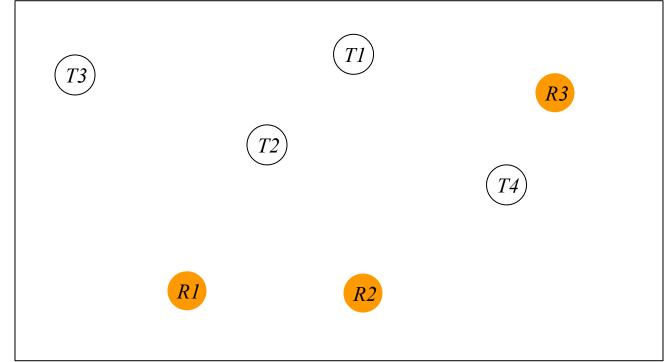
Iterate over steps 1-4 until convergence or planning time expires

Step 1: start with an arbitrary plan π

Step 2: all robots offer their tasks T_i at auction at the max. price of $C_{T_i}^{\pi(T_i)}$ - ϵ

Step 3: all robots R_j bid on the offered tasks T_i with the bid = $C_i^{R_j} + \epsilon$

Step 4: robots sell to the lowest bidders if they are below max. price and get profit: $C_{Ti}^{\pi(Ti)}$ - C_i^{Rj} - ϵ



Market-based Approach

Given N robots $R_1...R_N$, M tasks $T_1...T_M$, and C_i^{Rj} – cost of executing task i by robot R_j (cost may depend on other tasks executed by this robot)

When does it converge in one iteration?

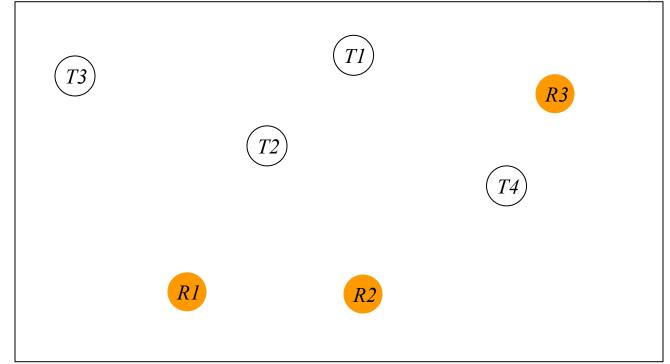
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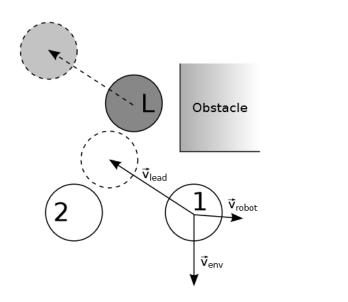
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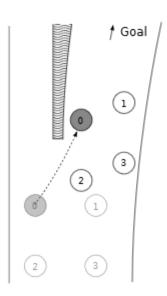
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Planning for Leader-based Coordination

- Fully decentralized approach (doesn't rely on the presence of communication between robots)
- Plan for the "leader" robot (sometimes leader can be just a centroid of the team or some other reference point)
- All other robots execute either "follow the leader" or "follow neighbors within field-of-view" behaviors while avoiding collisions





What You Should Know...

- Different styles of multi-robot planning
 - Centralized vs. decentralized
 - Joint state-space planning vs. distributed planning
 - Multi-robot path planning vs. cooperative task planning
- Prioritized Multi-robot Path Planning
- Market-based Approach to multi-robot planning