

CS 460/560

Introduction to Computational Robotics
Fall 2019, Rutgers University

Lecture 16

Multi-Robot Path Planning

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

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Outline

Multi-robot path planning

- ⇒ Applications
- ⇒ Formulations

Feasibility of graph-theoretic MPP

- ⇒ The 15-puzzle variant
- ⇒ The synchronous rotation variant

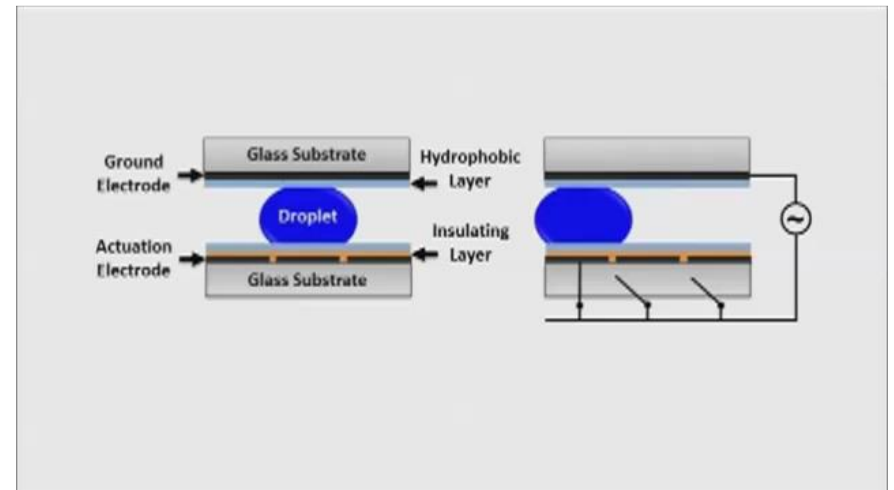
Structure and complexity of optimal MPP

Applications – Order Fulfillment



Applications

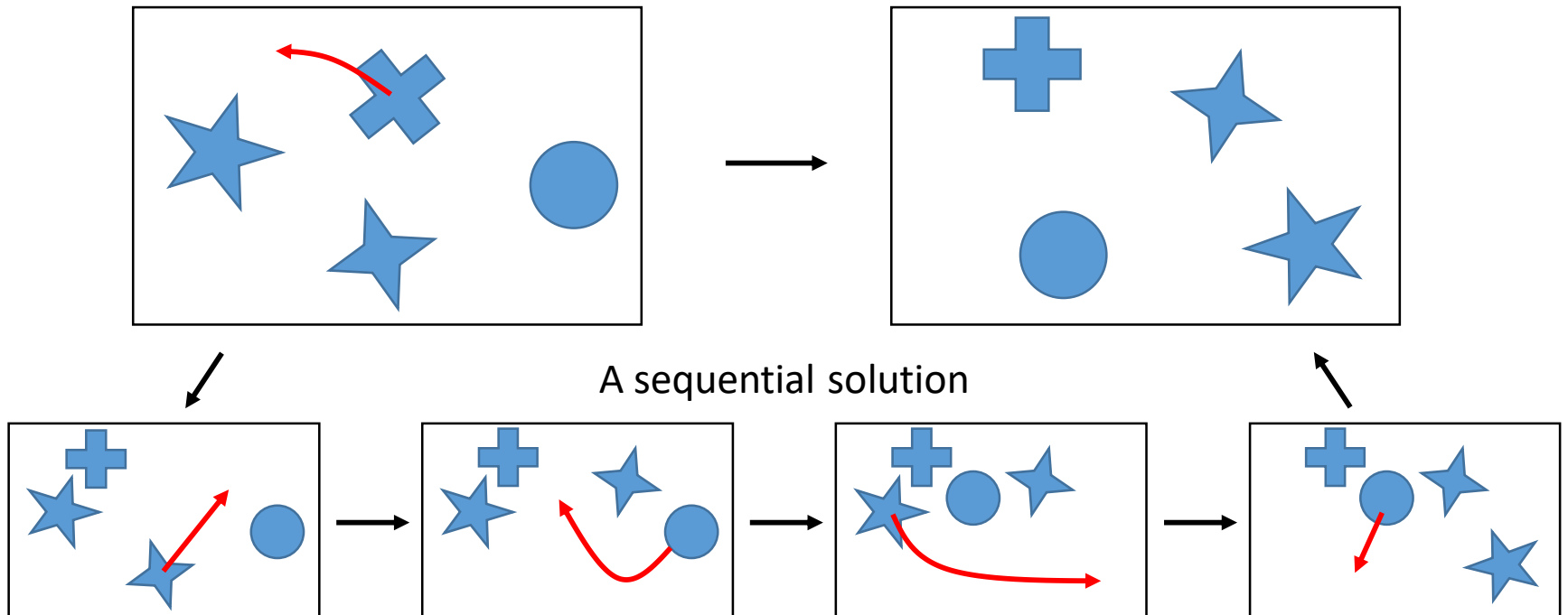
- ⇒ Container ports
- ⇒ Fulfillment centers
- ⇒ Delivery drones
- ⇒ Microfluidics (chemical/medical)
- ⇒ Future autonomous vehicles
- ⇒ Many others



Continuous Domain

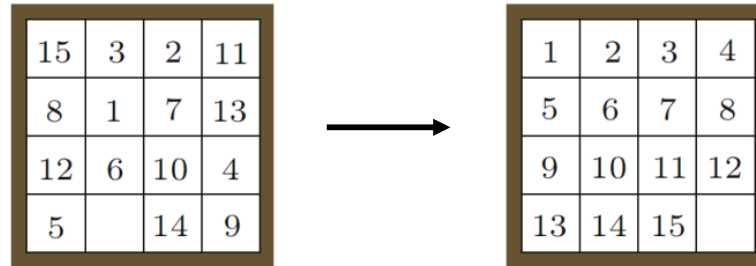
Continuous formulation

- ⇒ Some n objects, usually close to each other (crowded)
- ⇒ Usually in bounded region (why?)
- ⇒ The objects must be moved from one configuration to another
- ⇒ No collision is allowed
- ⇒ This problem (finding any solution) is PSPACE-hard

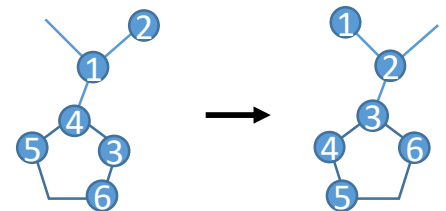
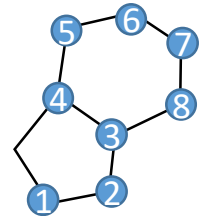


Discrete Domain: Single Move per Step

Origin: the 15-puzzle

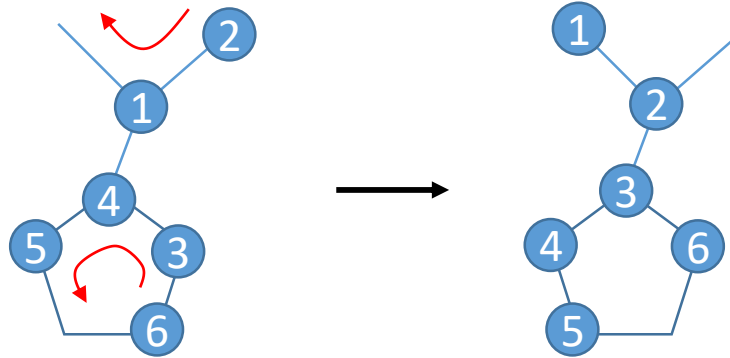


- ⇒ 4×4 game board, 15 square game pieces
- ⇒ Only pieces adjacent to the empty space can move
- ⇒ This is generalized first to $(N^2 - 1)$ -puzzle
- ⇒ Then generalized to **2-connected graph** with n vertices and $n - 1$ robots
- ⇒ Then generalized to general graph with n vertices and up to $n - 1$ robots
- ⇒ In these problems only one robot may move at a time
- ⇒ Let's call these problems MPP_s
- ⇒ Also known as the pebble motion problem (*PMG*)



Discrete Domain: Parallel Moves

For MPP_s on a n -vertex graph, moves may be parallelized

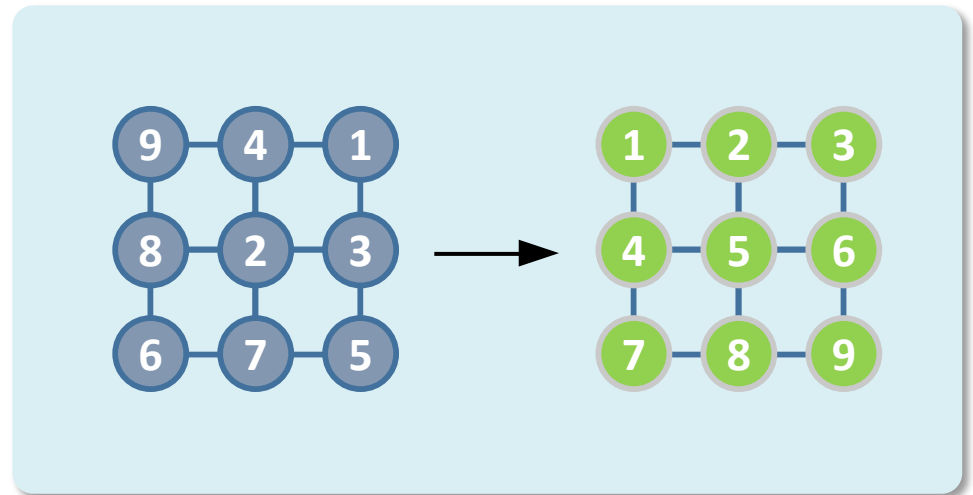
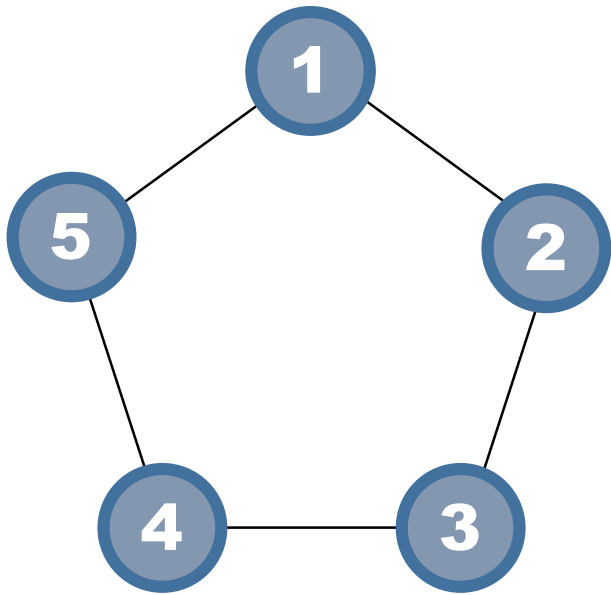


- ⇒ Essentially, there are multiple “move sequences”
- ⇒ Each sequence requires one empty vertex in the head
- ⇒ It is possible that all robots move in the same step
- ⇒ Let's call this MPP_p

Discrete Domain: Parallel Moves w/ Rotation

MPP_p still does not reflect full capabilities of modern robots

⇒ Robots should not always require empty vertex to move, e.g.,



⇒ We can further allow **rotations** in addition to parallel moves

⇒ Let's call this MPP_r

⇒ Problems can be feasible even when there are no empty vertices!

Feasibility of the 15-Puzzle

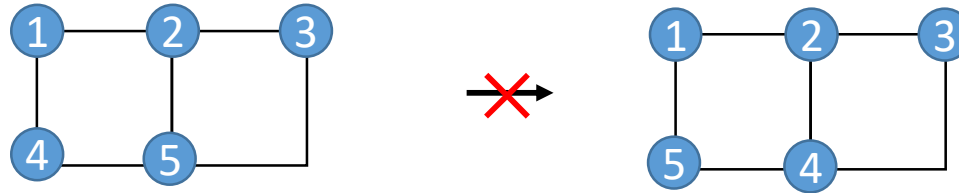
15	3	2	11
8	1	7	13
12	6	10	4
5		14	9

→

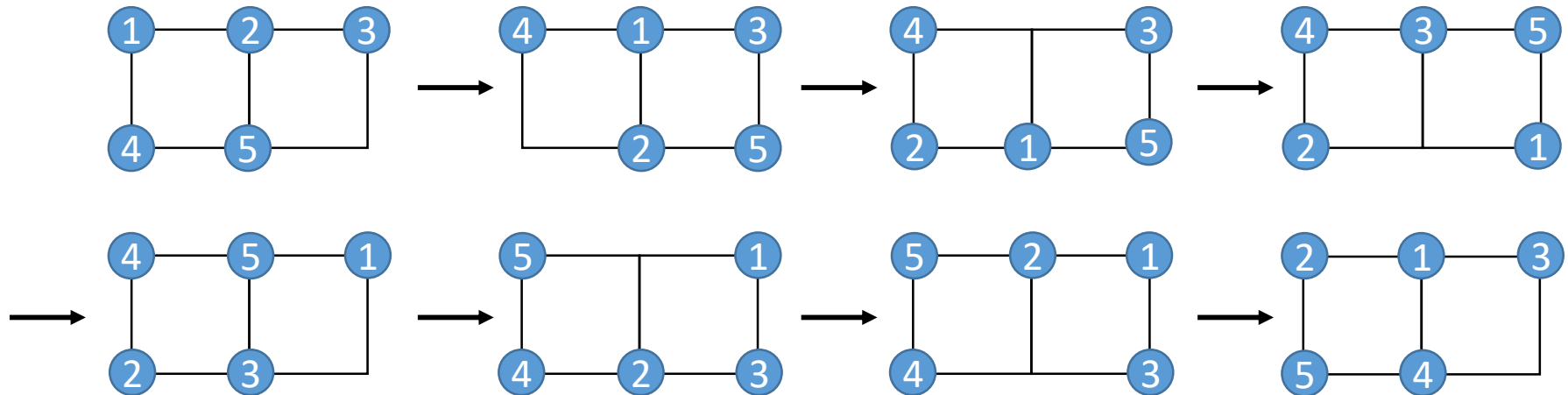
1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	

The 15-puzzle (and $(N^2 - 1)$ -puzzle in general) is not always feasible

⇒ In particular, if two robots are “swapped”, it defines an infeasible problem

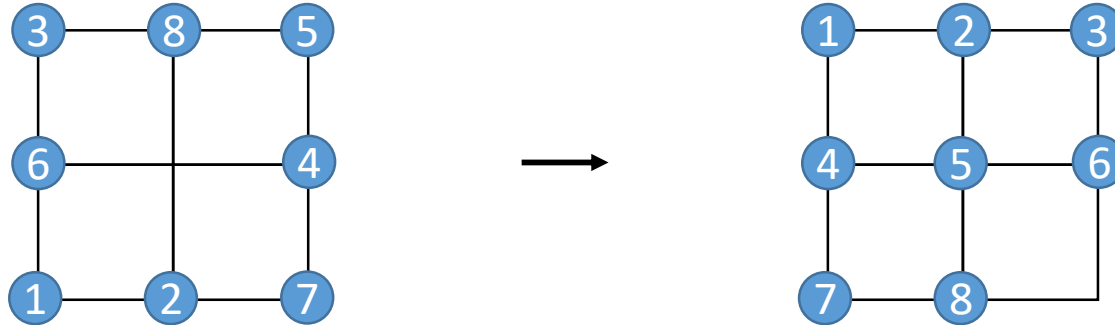


⇒ What happens if we legally swap 4 and 5?



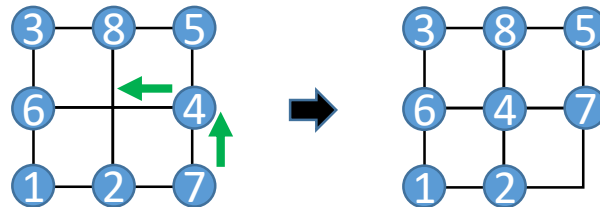
⇒ It also (always) flip at least one other pair (in this case, 1 and 2)

A Simple Method for $N^2 - 1$ Puzzle Feasibility

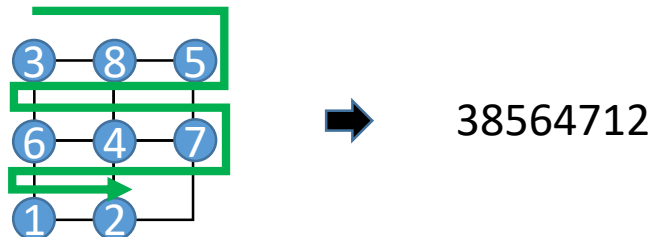


Steps

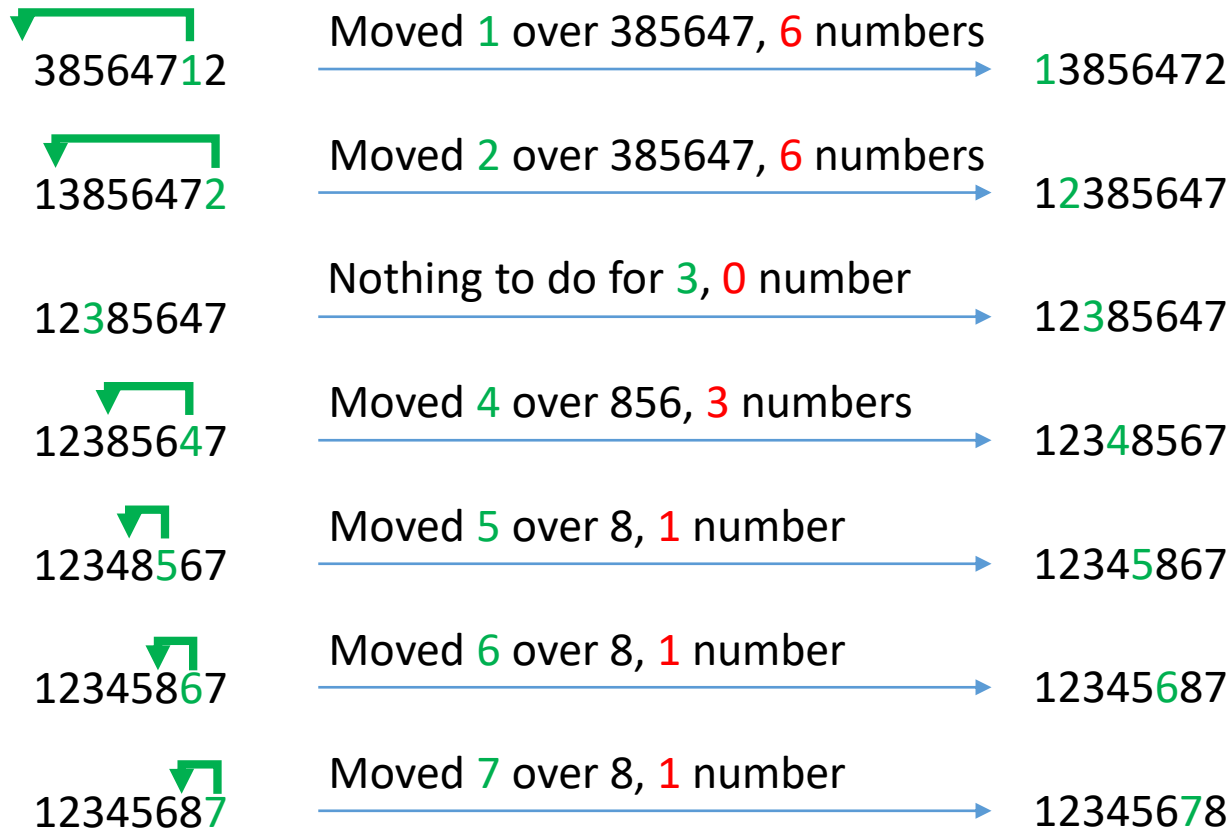
1. Move the empty spot to the lower right (doesn't matter how you do it)



2. Flatten the square row by row



3. Bubbling each number from 1 and count number of moves



4. Sum up $X = 6 + 6 + 0 + 3 + 1 + 1 + 1 = 18$

5. Odd – infeasible. Even – feasible. $X = 18$, instance is feasible

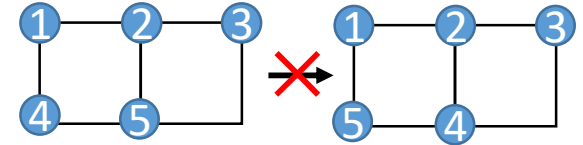
Solving of the 15-Puzzle (I)

For $(N^2 - 1)$ -puzzles

⇒ The only source of infeasibility comes from this

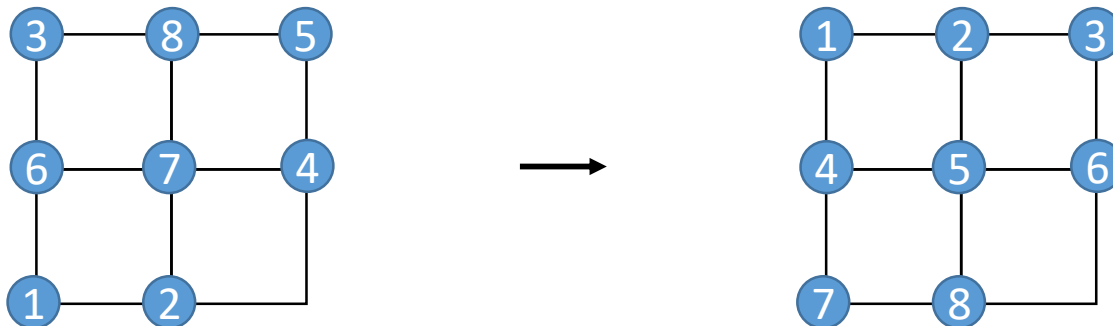
⇒ How can we detect this?

⇒ Through counting the number of moves for each robot



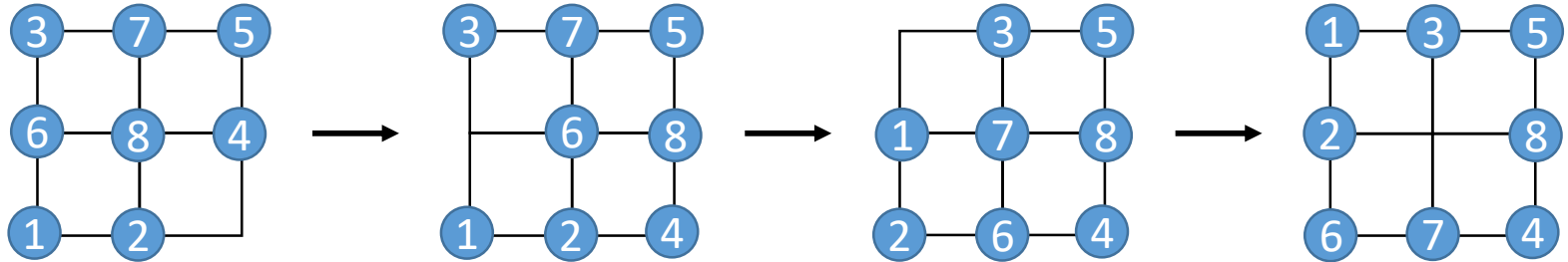
Now suppose we have a feasible problem, how do we solve it?

⇒ We will use the 8-puzzle as an example

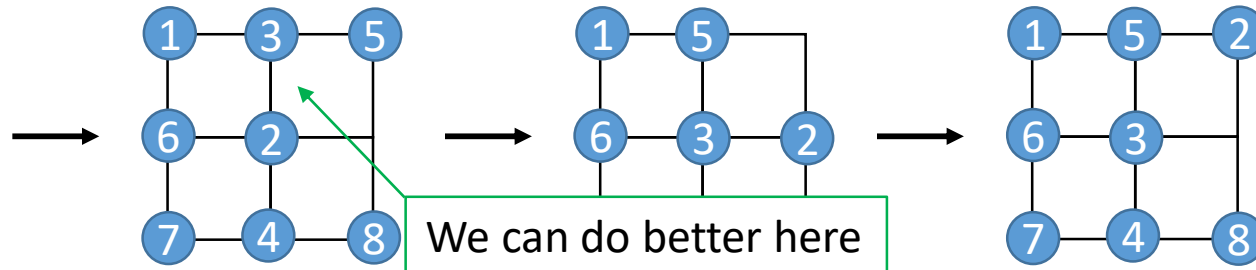


Solving the 15-Puzzle (II)

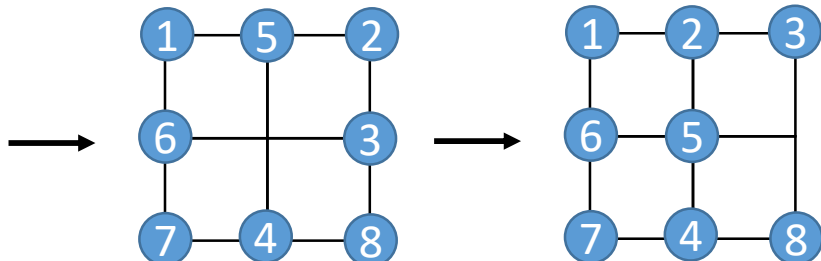
First, move 1 to its goal



Then, move 2 to 3's goal



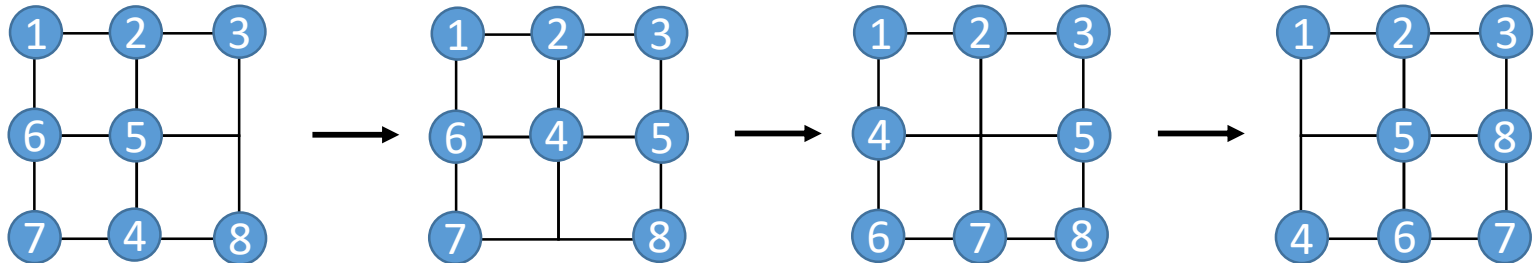
Then, move 3 to 6's goal, followed by moving 2,3 to their goals



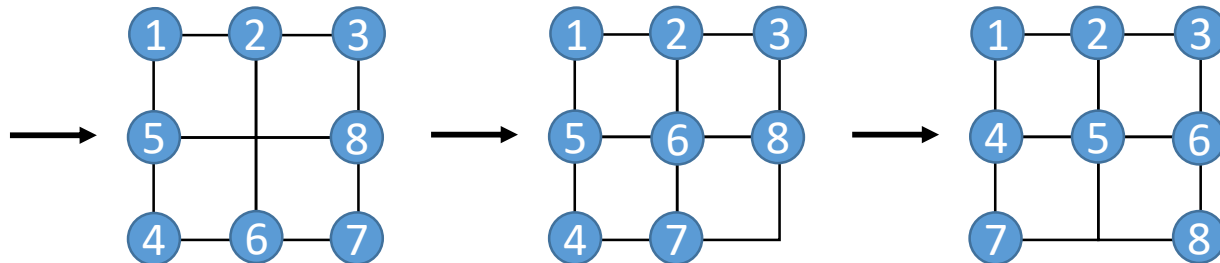
First row is solved!

Solving the 15-Puzzle (III)

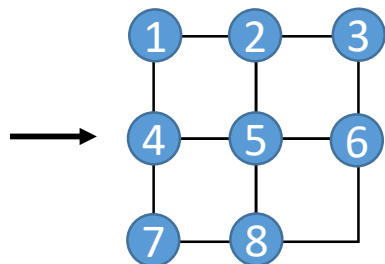
Then, solve the first column similarly. In this case, move 4 to 7's goal



Then, move 7 to 8's goal, and solve 4 and 7



Then, the last piece should be readily solvable



Solving the 15-Puzzle (IV)

Procedure for solving the $(N^2 - 1)$ -puzzle

- ⇒ Check feasibility
- ⇒ Solve the first (top) row
- ⇒ Solve the first (left most) column, now we have a $((N - 1)^2 - 1)$ -puzzle
- ⇒ Repeat the above steps until we get a 3-puzzle
- ⇒ The 3-puzzle should be readily solvable

The same procedure applies to $N \times M$ -puzzles where $N \neq M$

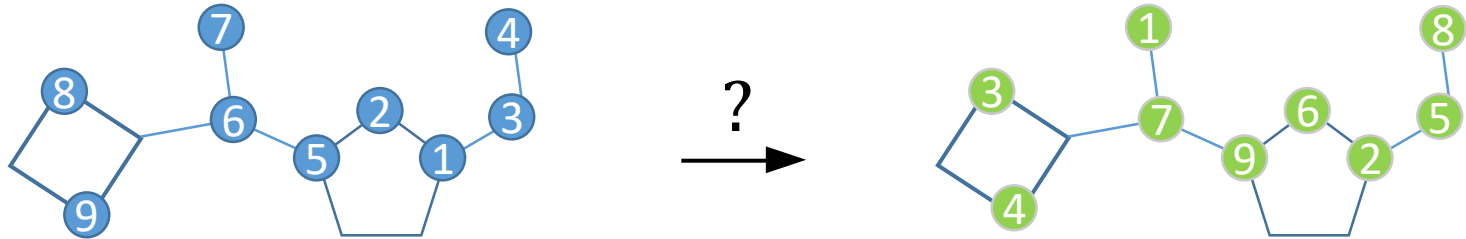
How many moves are needed?

- ⇒ For each robot, may need to move $O(N)$ robots
- ⇒ Each robot needs to be moved $O(1)$ times
- ⇒ So $O(N)$ total moves
- ⇒ For all $N^2 - 1$ robots, $O(N^3)$ total moves

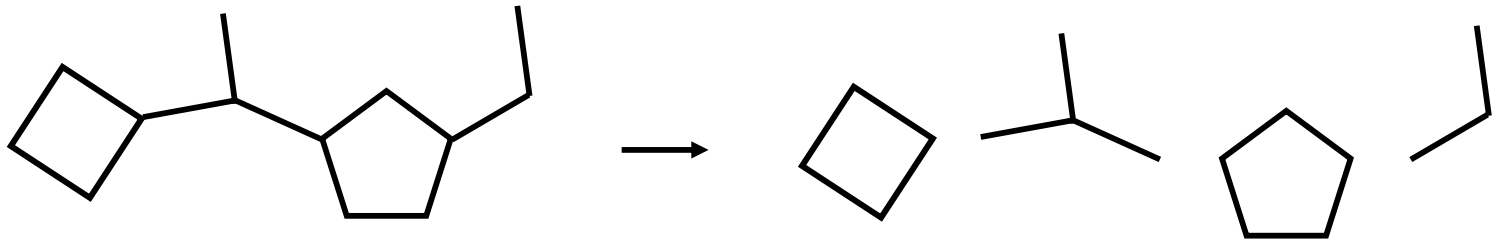
The method generalizes to 2-connected graphs

General Graphs

How about general graphs?



⇒ We can break it into trees and 2-connected graphs



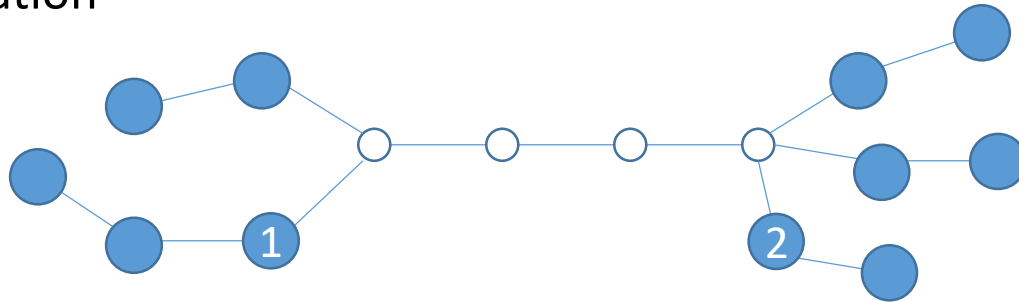
⇒ We already can solve 2-connected graphs

⇒ How about trees?

Partitioning Robots on Trees

We can solve the problem by partitioning robots into **equivalence classes**

- ⇒ Two robots belong to the same equivalence classes if they can exchange locations
- ⇒ It turns out this is easy to determine (well, intuitively)
- ⇒ Two robots 1 and 2 that are “adjacent” (i.e., no other robots between them on the tree) cannot exchange (i.e., not equivalent) if they have the following configuration



- ⇒ If there is one more empty vertex, then 1 and 2 are equivalent
- ⇒ This can be checked (in linear time)
- ⇒ Combined with 2-connected graph solver, solves MPP_s on general graph

Feasibility with Parallel Moves

MPP_p feasibility is the same as MPP_s

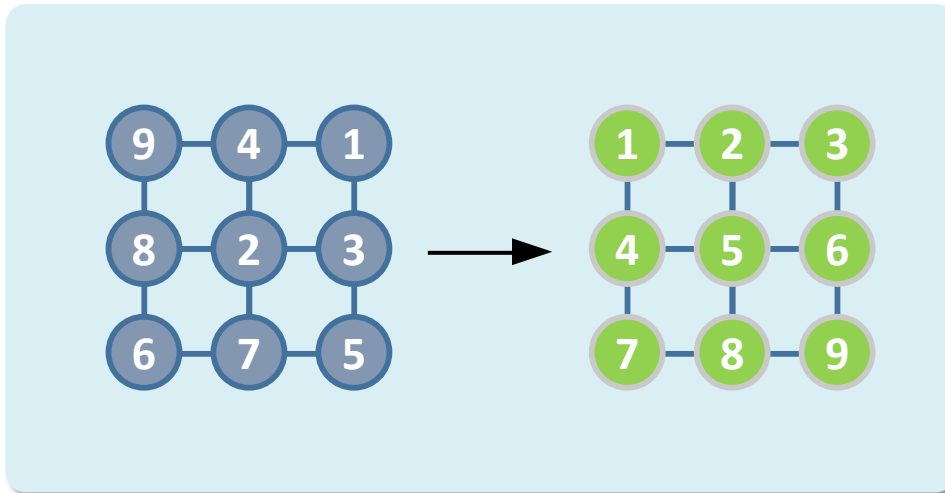
⇒ Why?

⇒ Every parallel move can be carried out as step-by-step single robot moves

MPP_r , it's slightly different

⇒ We can decompose the problem into trees and 2-edge-connected components

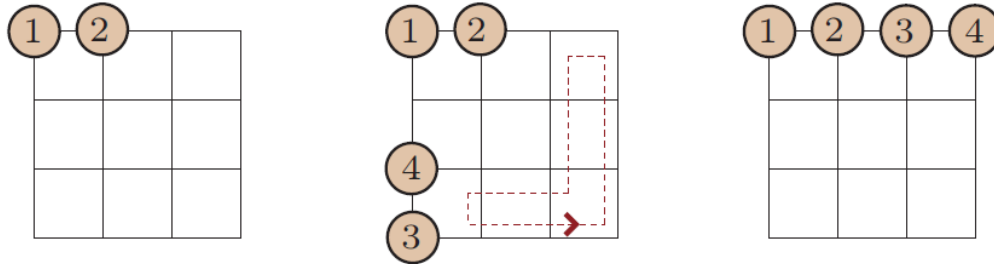
⇒ But, we need some methods for solving problems like this (N^2 -puzzle)



Solving N^2 puzzles

N^2 -puzzle (MPP_r) is similar to $(N^2 - 1)$ -puzzle (MPP_s)

\Rightarrow We can first solve the top row (and left most column)

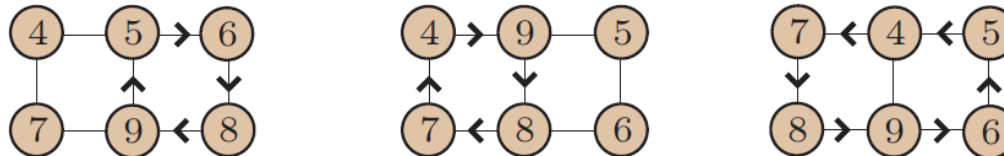


\Rightarrow This yields an $(N - 1)^2$ -puzzle

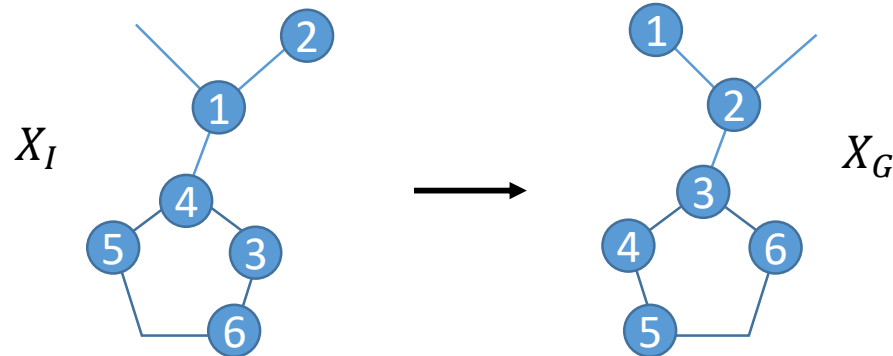
\Rightarrow We do this until we get to 3×3 , and solve the top row

\Rightarrow This leaves us with a 2×3 puzzle, which can be solved

\Rightarrow E.g., exchange 8 and 9



Optimal Formulations



MPP Problem: (G, X_I, X_G) , solution: collision free $P = \{p_1, \dots, p_n\}$

Optimality objectives (minimization):

\Rightarrow Max time (makespan): $\min_{P \in \mathcal{P}} \max_{p_i \in P} \text{time}(p_i)$

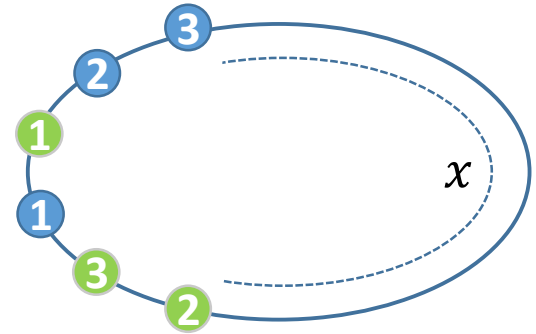
\Rightarrow Total time: $\min_{P \in \mathcal{P}} \sum_{p_i \in P} \text{time}(p_i)$

\Rightarrow Max distance: $\min_{P \in \mathcal{P}} \max_{p_i \in P} \text{length}(p_i)$

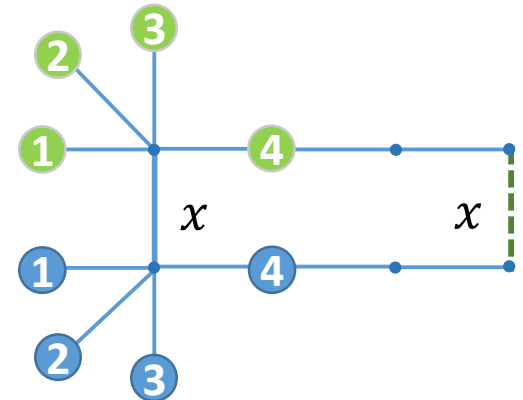
\Rightarrow Total distance: $\min_{P \in \mathcal{P}} \sum_{p_i \in P} \text{length}(p_i)$

Incompatibility of the Formulations

	Makespan	Total time
Clockwise	$x + 1$	$2x + 3$
Counterclockwise	$x + 4$	$x + 12$



	Total distance	Total time
Left path only	$4x + 8$	$4x + 14$
Using right path	$4x + 10$	$4x + 13$



A pair of the four MPP objectives on makespan, total time, max distance, and total distance demonstrates a Pareto-optimal structure.