

Constrained Geometric Approximation Approach for Robot Planning and Decentralized Formation Algorithm for Multi-Robot Systems

Yang Song, Jason M. O'Kane
song24@email.sc.edu
jokane@cse.sc.edu

Dept. of Computer Science and Engineering
University of South Carolina





Outlines

CGA for Robot Planning

Overview

Problem Statement

Range Space

Range Space Definition

Disk Range Space

Rectangle Range Space

Double-Rectangle Range Space

Experiments

Introduction

Lattice Graph

Algorithm

Robot Authority

Results

Conclusions

Constrained Geometric Approximation



3

► **Goal:**

For an extremely simple robot with:

- ▶ computation limitations
- ▶ moving and sensing uncertainties

represent and reason about uncertainty
in its own states efficiently.

► **Basic Idea:**

Explicitly represent what the robot
knows as an information state (*I-state*).

► **Intuition:**

Accelerate time-consuming operations
by maintaining only an
overapproximation of the true *I-state*,
and constraining this approximation to
have a simple geometric form.



SRV-1 Surveyor Robot

Robot Model



Assume that current real state of the robot could not be observed directly. The robot could maintain an *I-state* η_k , to make its decisions.

- ▶ Robot state at stage k : $x_k \in X$,
- ▶ State transition function: $F(x_k, u_k)$.
- ▶ Robot action at stage k : u_k .
- ▶ Information state (*I-state*) at stage k : η_k is a set of possible states at stage k

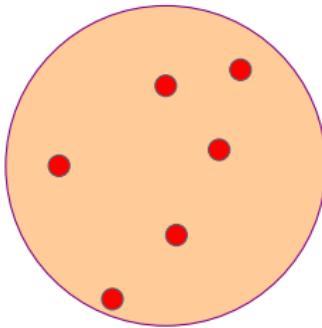


Figure: I-state η_k contains all possible states

Prior Work



5

- ▶ Prior research done by (B. Tovar and S. M. LaValle) and (J. van den Berg, P. Abbeel, and K. Goldberg) used probabilistic representations for planning
- ▶ Prior work by the J.O'Kane has used preliminary versions of the constrained geometric approximation method using specific, fixed range spaces.

New contributions

1. A careful formulation of the operations in the range space \mathcal{R} .
2. Algorithms for double-rectangle range space \mathcal{R}_{direct} .
3. A series of experiments for effectiveness comparison of different range spaces.



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Range Space



Definition

A **range space** $\mathcal{R} \subseteq \mathcal{I}$ is a set of I-states, contains approximation of I-states, $A(\eta_k) \in \mathcal{R}$, equipped with two operations:

1. An *approximate observation update function*

$O : \mathcal{R} \times Y \rightarrow \mathcal{R}$, such that if
 $\eta_k \subseteq A(\eta_k)$, then

$$\eta_k \cap H(y_k) \subseteq O(A(\eta_k), u_k)$$

2. An *approximate action update function* $T : \mathcal{R} \times U \rightarrow \mathcal{R}$, such that if $\eta_k \subseteq A(\eta_k)$, then

$$\bigcup_{x_k \in \eta_k} F(x_k, u_k) \subseteq T(A(\eta_k), u_k)$$

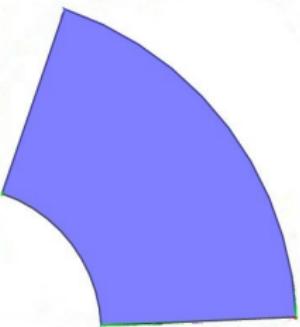


Figure: An I-state η_k

Range Space



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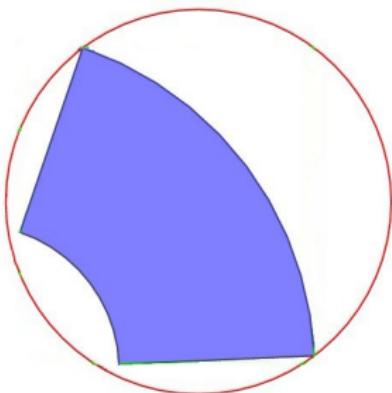


Figure: Disk overapproximation $A(\eta_k) \in \mathcal{R}_{disk}$ in red.

Range Space



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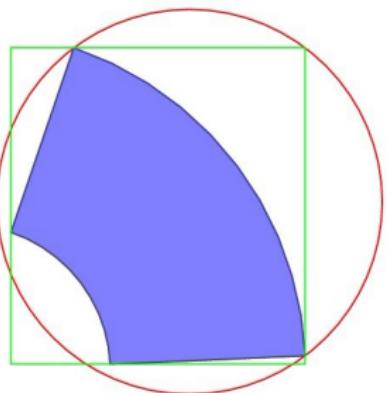


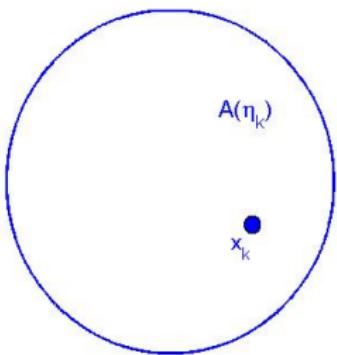
Figure: Rectangle overapproximation $A(\eta_k) \in \mathcal{R}_{rect}$.

Observation Update in \mathcal{R}_{disk}



10

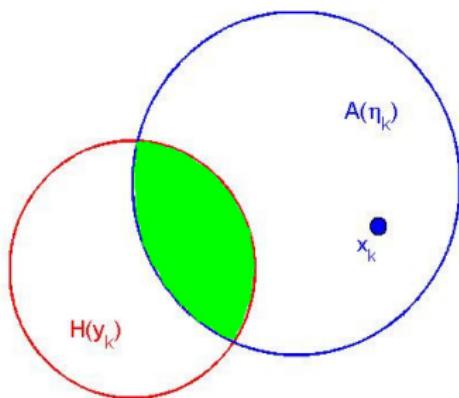
Approximated I-state $A(\eta_k) \in \mathcal{R}_{disk}$, where x_k denotes the real state but unknown to robot.



Observation Update in \mathcal{R}_{disk}



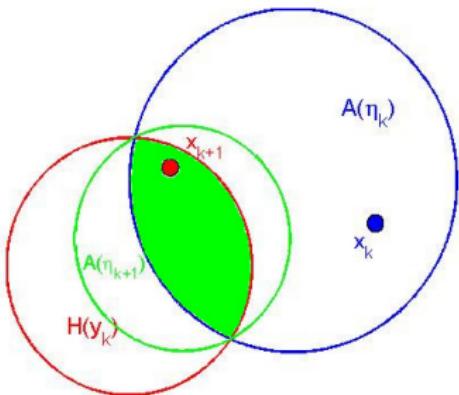
Approximated I-state $A(\eta_k) \in \mathcal{R}_{disk}$ intersects with observation preimage $H(y_k)$.



Observation Update in \mathcal{R}_{disk}



The green region of intersection is the updated I-state η_{k+1} , and the green disk is the approximation $A(\eta_{k+1})$ of η_{k+1} .



Observation Update in \mathcal{R}_{rect}

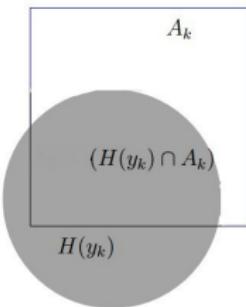


Definition

$AABB(S)$: For any compact set $S \subset \mathbb{R}^2$, let $AABB(S)$ denote the its smallest “axis-aligned bounding box.”

In \mathcal{R}_{rect} , computing approximate observation update function O_{rect} takes $O(1)$ time:

$$O_{rect}(A_k, y_k) = AABB(H(y_k) \cap A(\eta_k)), A_k = A(\eta_k) \in \mathcal{R}_{rect}$$



Observation Update in \mathcal{R}_{rect}



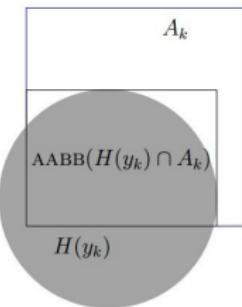
14

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Action Update in \mathcal{R}_{rect}



In \mathcal{R}_{rect} , computing approximate action update function T_{rect} :

$$T_{rect}(A(\eta_k), u_k) = AABB(X_{free} \cap [A(\eta_k) \oplus \{u_k\} \oplus AABB(\Theta(u_k))])$$

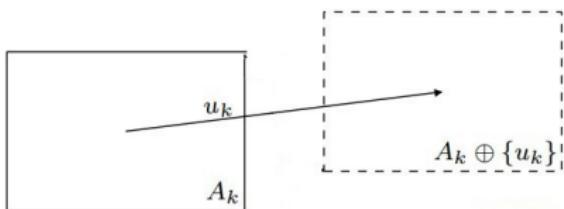


Figure: Transition of $A(\eta_k)$ given action u_k , \oplus is Minkowski addition of two sets.

Action Update in \mathcal{R}_{rect}



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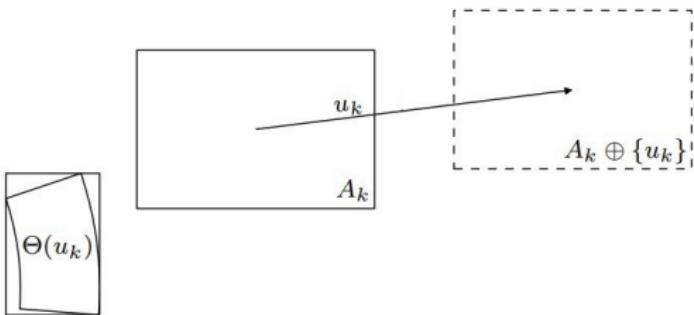


Figure: Consider approximation of the bounded motion uncertainty $\Theta(u_k)$.

Action Update in \mathcal{R}_{rect}



In \mathcal{R}_{rect} , computing approximate action update function T_{rect} :

$$T_{rect}(A(\eta_k), u_k) = AABB(X_{free} \cap [A(\eta_k) \oplus \{u_k\} \oplus AABB(\Theta(u_k))])$$

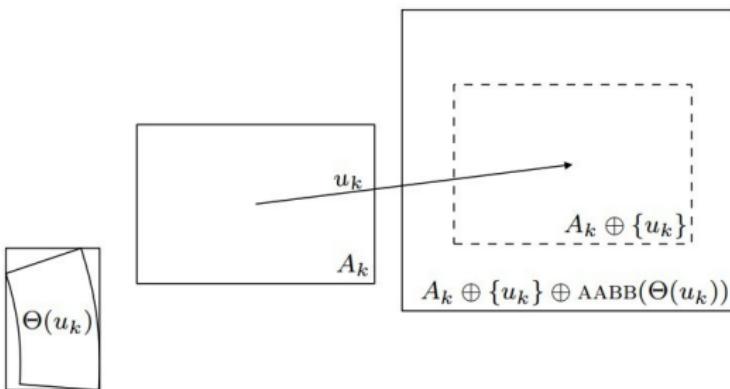


Figure: Find Minkowski sum of bounded noise and approximation transition

Action Update in \mathcal{R}_{rect}



In \mathcal{R}_{rect} , computing approximate action update function T_{rect} :

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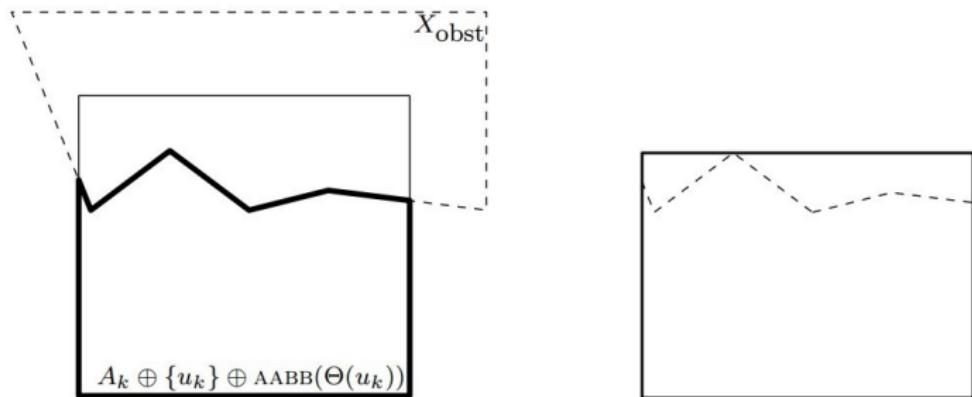


Figure: If there is obstacles, intersect with X_{free} first and then find the bounding box of the intersection.

Double-Rectangle approximated I-state



- ▶ For better overapproximation quality for non-convex *I*-states, we proposed a more expressive range space of *double rectangles*:

$$\mathcal{R}_{direct} = \{R_1 \cup R_2 \mid R_1, R_2 \in \mathcal{R}_{rect}\} \quad (1)$$

- ▶ Aims to improve the approximation quality.

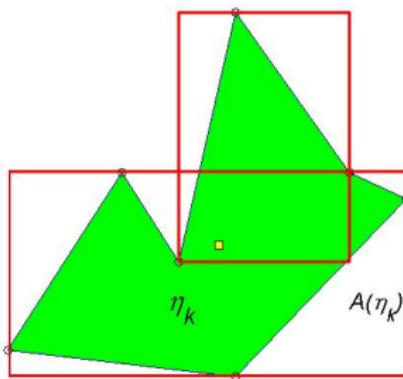


Figure: Non-convex *I*-state

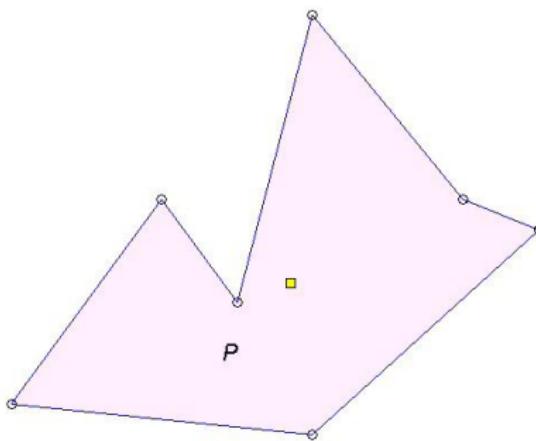
DRAP Algorithm



20

“Double Rectangle Around Polygon” (**DRAP**) algorithm:

- ▶ input is a n -edge polygonal region of the plane
- ▶ output is a small double rectangle containing that polygon
- ▶ run time: $O(n^3)$



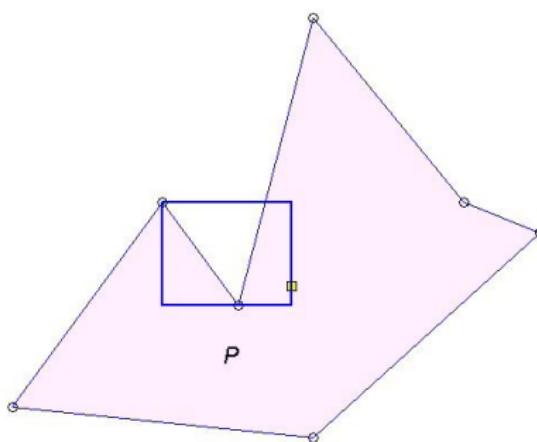
DRAP Algorithm



21

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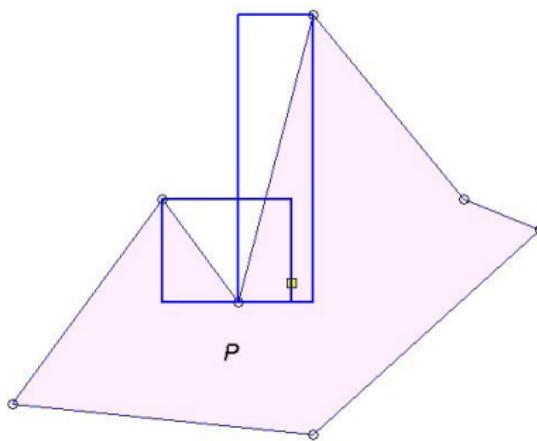
DRAP Algorithm



22

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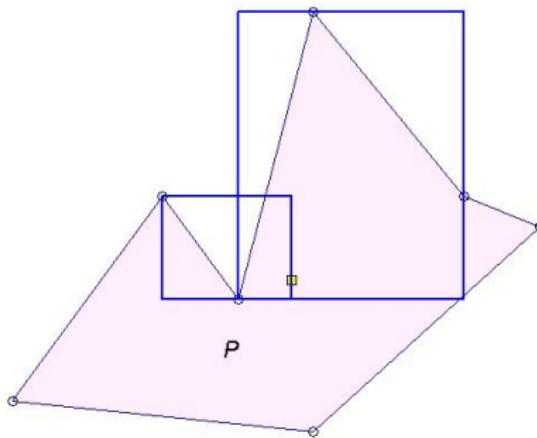
DRAP Algorithm



23

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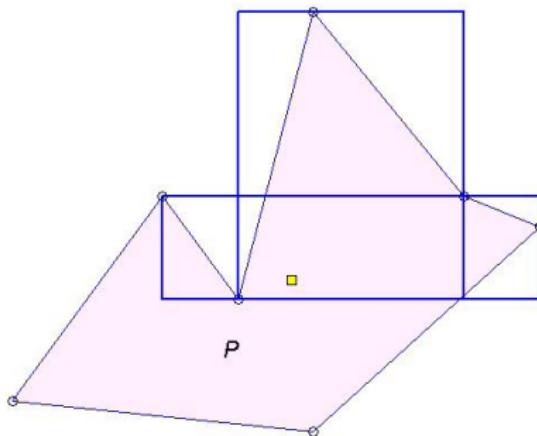
DRAP Algorithm



24

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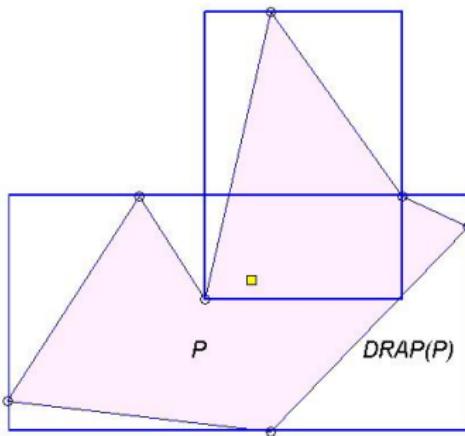


DRAP Algorithm



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26

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Experiments in Various Environments



27

Comparison with using the true I-state, we conducted experiments using 3 environments, and 3 range spaces \mathcal{R}_{disk} , \mathcal{R}_{rect} , and \mathcal{R}_{direct} .

ASSUMPTIONS:

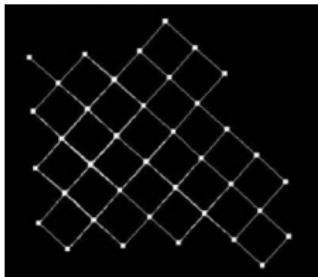
- ▶ Robot is guided by centroid point of the approximated *I-state*
- ▶ Robot can detect presence but not distance to the settled waypoints
- ▶ Landmarks are pseudo-randomly generated (Number Matters)
- ▶ Initial I-state η_0 is given.

Related Work

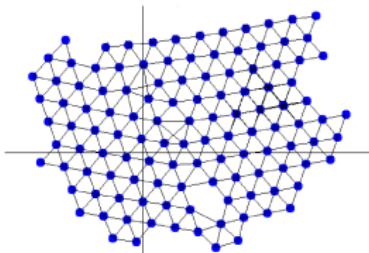
Formation using virtual force



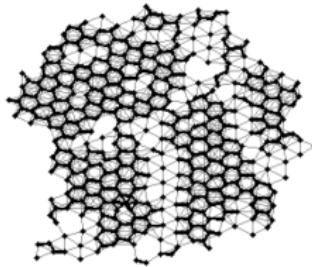
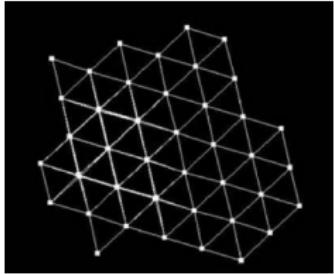
W. Spears, D. Spears, J. Hamann and R. Heil, 2004



I. Navarro, J. Pugh, A. Martinoli, and F. Matia, 2008



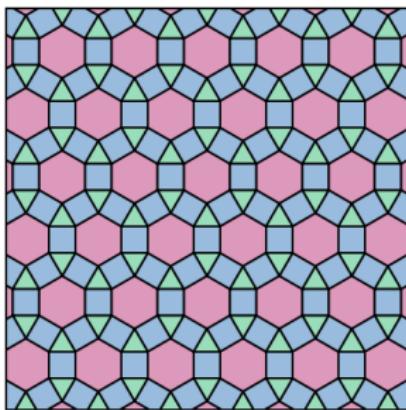
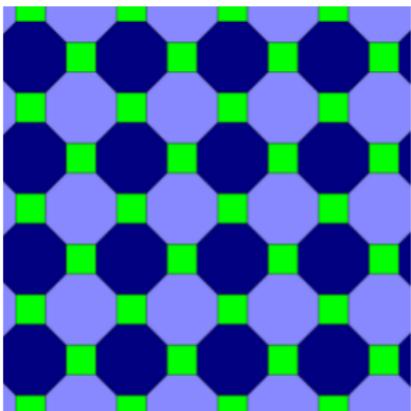
S. Prabhu, W. Li, J. McLurkin, 2012



Motivation



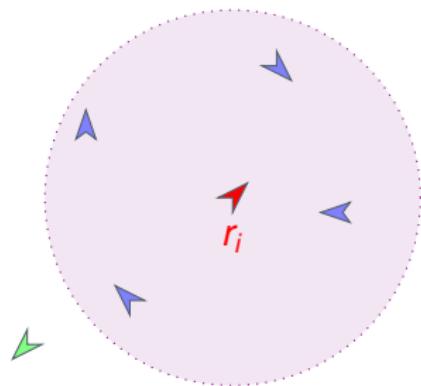
Question: How to use one algorithm to generate various (repeating) lattice pattern formations?



Robot Model



- ▶ Differential Drive robots.
- ▶ Each robot has an unique **ID**.
- ▶ Use a vector $p = [x, y, \theta]^T$ to represent robot's **pose**.
- ▶ Each robot has a **range** within which it can sense and communicate with other robots.
- ▶ Each robot gets **observation** of its neighbors' IDs and relative poses in its body frame.



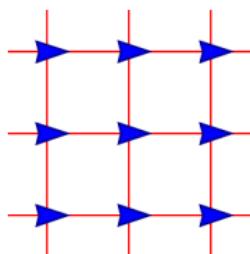
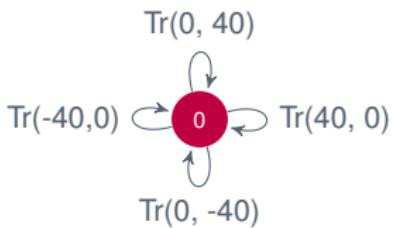
Robot r_i has 4 neighbors

Input: Lattice Graph



Definition

A **lattice graph** is a strongly connected directed multigraph in which each edge e is labeled with a rigid body transformation $T(e)$ and each $v \xrightarrow{T(e)} w$ has an inverse edge $w \xrightarrow{T(e)^{-1}} v$.

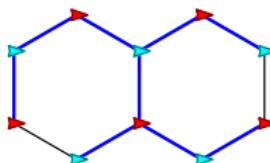
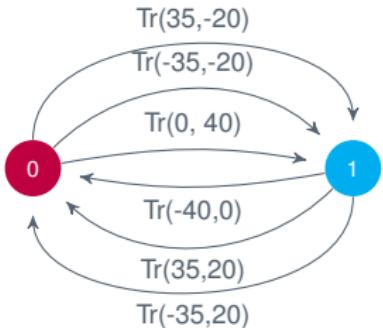




Definition

Given a lattice graph $G = (V, E)$ and a set of robots $R = \{r_1, \dots, r_n\}$, R **satisfies** G if there exists a role function $f : R \rightarrow V$ that preserves the neighborhood structure of G .

Specifically, for any i and j , if r_i and r_j are neighbors, there must exist an edge $e_{ij} : f(r_i) \longrightarrow f(r_j)$ in E , such that $T(r_j) = T(r_i)T(e_{ij})$.





General Description

Robot broadcasts message containing its

- ▶ authority
 - ▶ matching.
1. Form tree structure.
 2. Use tree structure to computer local task assignment.
 3. Make movement decision.



Define authority and comparison operator



An **authority** is an ordered list of robot IDs

$$(id_1, \dots, id_k)$$

The first ID in the list, id_1 is called the **root** ID. The final ID in the list, id_k is called the **sender** ID.



Authority A_2 is **higher than** A_1 if:

- ▶ root ID of $A_2 >$ root ID of A_1 , or
- ▶ length of $A_2 <$ length of A_1 if they have the same root, or
- ▶ sender ID of $A_2 >$ sender ID of A_1 if they have the same root and length.

1. Construct Authority Tree

Decide to be root or descendant



35

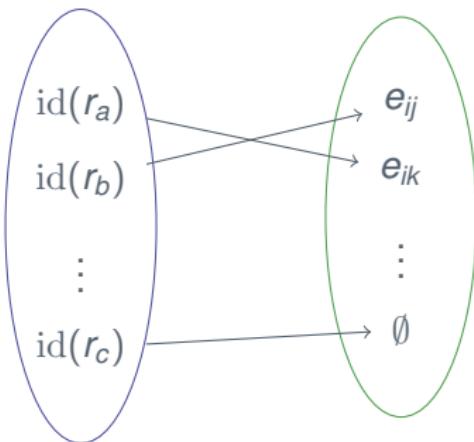
The robots use these authorities to establish a collection of authority trees

1. Discards any message in which the authority contains its own ID.
2. Forms an authority containing only its own ID, compares it with the authorities of remaining messages and selects the highest authority.
 - ▶ If its authority is the highest, then it is a **root**;
 - ▶ Otherwise, it selects the one who sends the highest authority as its parent. Append its own ID to the highest authority to create its own authority.

Matching



A **matching** for a robot is a function $\eta : \{\text{id}(r_a), \text{id}(r_b), \dots\} \rightarrow \{\emptyset, e_{ij}, e_{ik}, \dots\}$ that associates each neighbor ID with either a lattice graph edge from its role vertex or with the null value \emptyset .



2. Local Task Assignment

Hungarian Algorithm



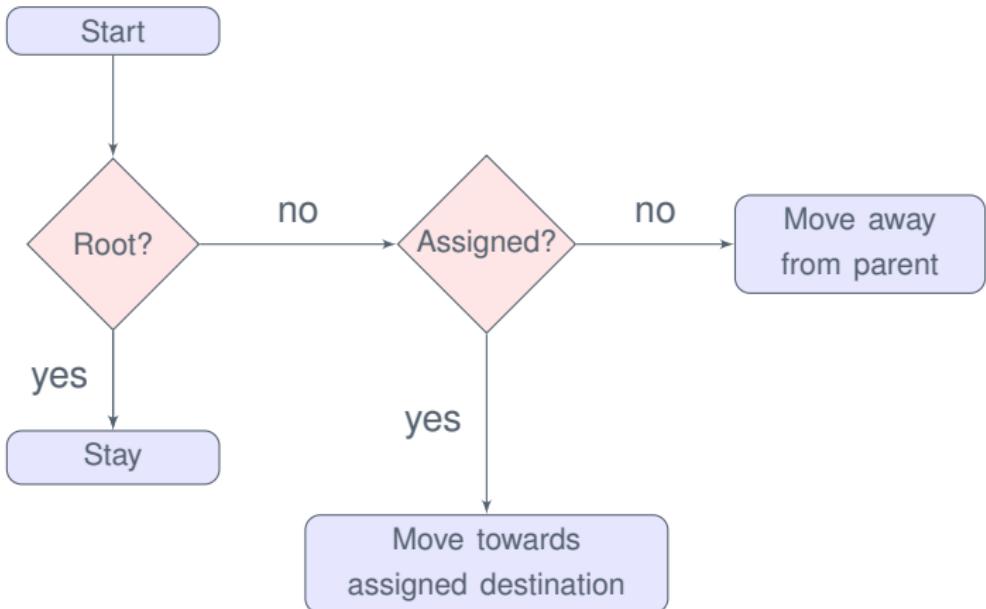
37

To compute an optimal matching of a robot with N neighbors and E out-going edges of its role in the lattice graph, define a weight matrix of size $N \times \max(N, E)$ and apply **Hungarian Algorithm** (Harold W. Kuhn, 1955).

1. Each row corresponds to a neighbor;
2. Each column corresponds to an out-going edge of robot's role or a null value \emptyset .
3. The entries of the matrix are the Euclidean distance between current position of each neighbor and the desired position if matched with a lattice graph edge.

5 neighbors, 4 out-going edges.

3. Robot Movement Strategy



Bounded Movement



Goal: let descendant stay in the range of its parent

- ▶ Within the set O (**Red circle**), the parent is guaranteed to get observation at next stage.
- ▶ Descendant can reach anywhere in set P (**blue circle**) at next stage.
- ▶ The real destination for descendant is the closest point on the boundary of the intersection ($O \cap P$) to the assigned destination.

Simulation

Octagon-square formation with 100 robots



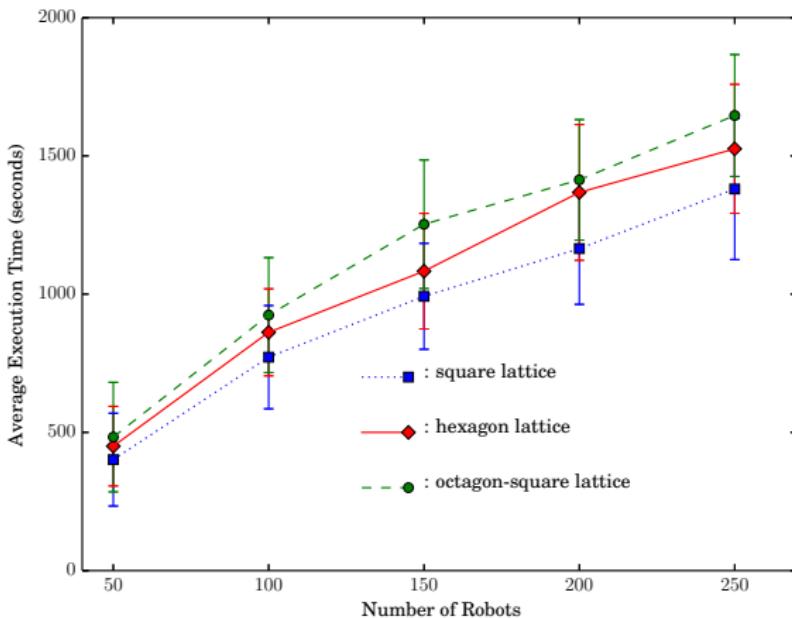
Simulation

Hexagon formation with 100 robots



Experiment Results

on three kinds of repeated lattice patterns



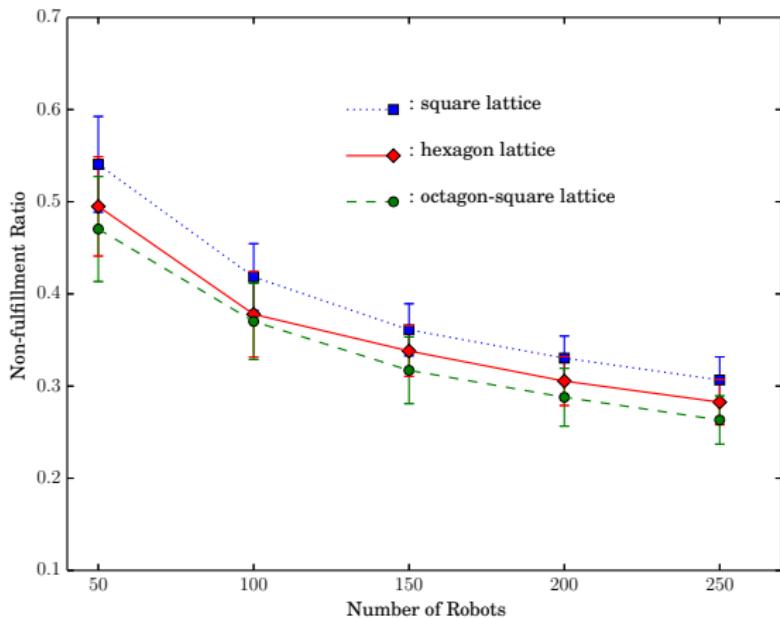
Average time to the static position with 50 trials, uniform distribution.

Experiment Results

on three kinds of repeated lattice patterns



43



Average non-fulfillment ratio $\Gamma = \frac{1}{n} \sum_{i=1}^n \frac{E_i - N_i}{E_i}$ with 50 trials, uniform distribution.

Conclusions



Summary

- ▶ Robots can form different types of geometric formations, including the repeated lattice patterns
- ▶ Algorithm scales reasonably well with increasing numbers of robot
- ▶ Algorithm is robust to the situation when some robots are removed from or with more robots added to the system

Future Work

- ▶ Improve the motion strategy
- ▶ Prove the convergence
- ▶ Nonholonomic constraints

Questions



45

