

Massive Self-Organized Shape Formation in Grid Environments

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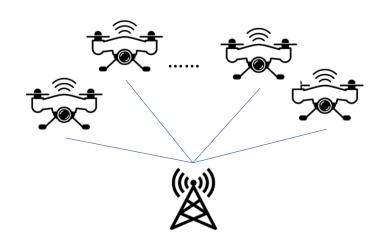
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Overview

- IntroductionOur Approach
 - > Preliminary Results



Introduction





Increasingly Large-scale
Collaborative Tasks in the Connected
World with Autonomous Agents.



Smart Warehouse



Intelligent Transportation



Drone Delivery



Light Show with Drones



Introduction

In multi-agent collaboration scenarios, integrated task assignment and path planning is the core task, and this task is also referred to as a shape formation task.

Current Approaches' Demerits:

- Centralized approaches based on task assignment and path planning suffer from poor scalability concerning the number of involved agents due to the high computational cost;
- Edge-following based approach manifests low efficiency due to the low parallelism among agents;
- Artificial potential field (APF) based approaches exhibits poor stability due to the high risk of an agent's falling in local minima.

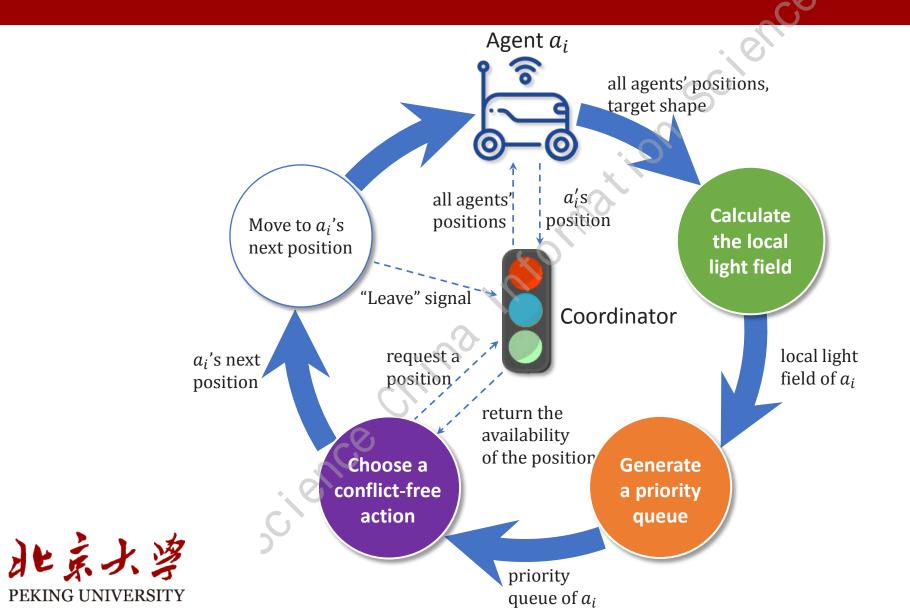


lacking an <u>efficient</u>, <u>scalable</u>, and <u>stable</u> mechanism for shape formation!

The shape formation phenomenon in nature, e.g. ants build a bridge, is a collective intelligence (CI) phenomenon, which indicates that shape formation problem can be solved in highly self-organized ways based on the CI theory.

- The essence of our approach is a continuously executing loop of information exploration, integration, and feedback among agents in a collective, following a constructive model for collective intelligence.
- An artificial light eld (ALF) is introduced and superimposed on the grid environment, serving as a carrier for information integration and feedback.
- A mutual feedback process emerges between the ALF and the agent collective: the current positions of all agents in the grid environment determine the current state of the ALF, which in turn drives agents to change their current positions.







Calculate the local Light Field.

The **FOV** of a_1 0.5 a_1 0.5 0.5 A Red Light

 a_1 is an un-target agent.

 a_1 decides its next action based on local red light field.

Move towards a neighbor grid with the highest red light intensity.

Upper Right





Source

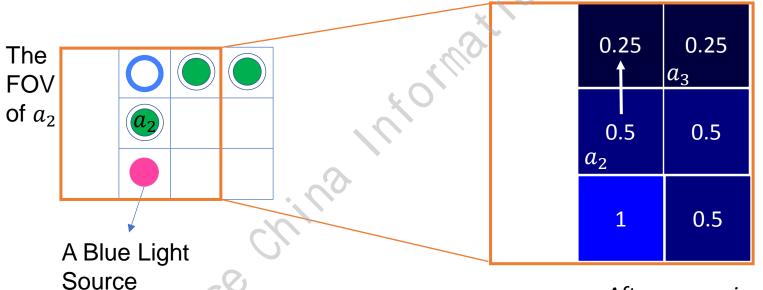




Calculate the local Light Field.

 a_2 has occupied a grid.

 a_2 decides its next action based on local blue light field.



Move towards a neighbor grid with the lowest blue light intensity (and unoccupied).

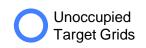
Up

After a_2 moving up, the edge node of the connected target area is vacated

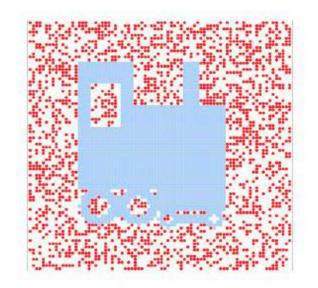


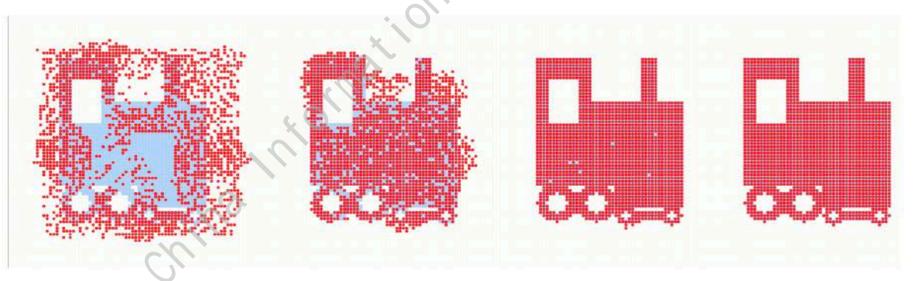






Preliminary Results

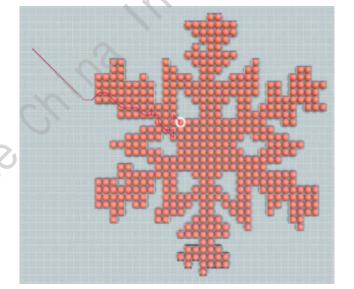






Preliminary Results





The trajectory of an agent



Thank You!

