

# Lifelong Multi-Agent Path Finding for Online Pickup and Delivery Tasks

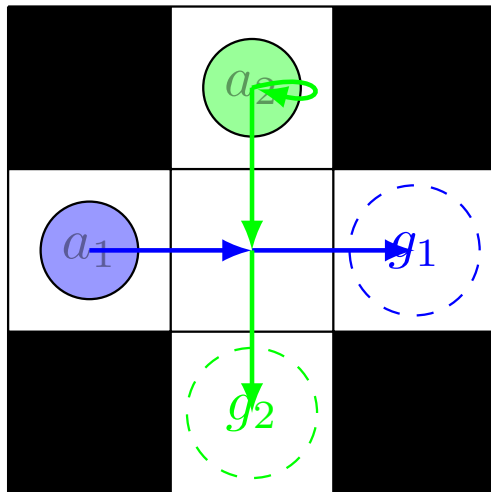
Hang Ma   Jiaoyang Li   T. K. Satish Kumar   Sven Koenig  
University of Southern California  
Tsinghua University

May 11, 2017  
AAMAS



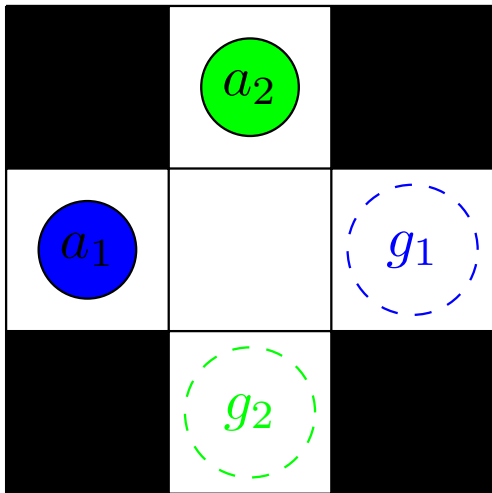
# Multi-Agent Path Finding

Find collision-free paths for all agents from their current locations to their predefined goal locations in a known environment.



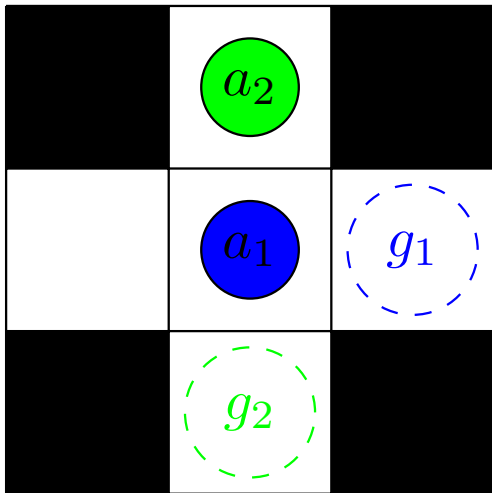


# Time Step 0



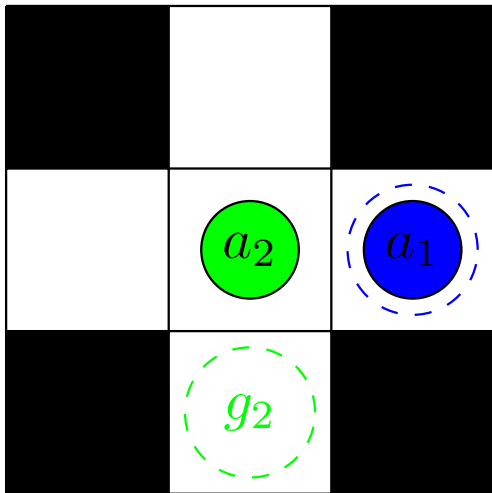


# Time Step 1



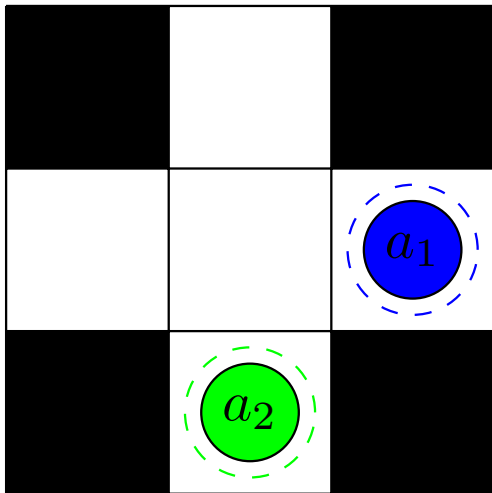


## Time Step 2





## Time Step 3





# Motivated by Real-World Applications:

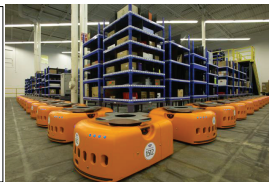
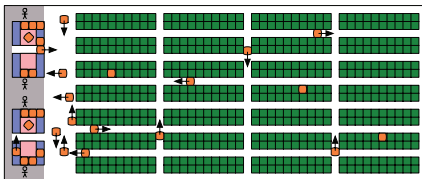
Automated aircraft-towing vehicles, warehouse robots, office robots, and game characters in video games.





# Amazon Warehouse Robots<sup>1</sup>

Tasks: Move inventory shelves from storage locations to inventory stations or vice versa.



<sup>1</sup> P. R. Wurman, R. D'Andrea, and M. Mountz. "Coordinating Hundreds of Cooperative, Autonomous Vehicles in Warehouses". In: *AI Magazine* 29.1 (2008), pp. 9–20.





## Multi-Agent Pickup and Delivery (MAPD) Problem

- ▶ Existing research on multi-agent path finding — a “one-shot” version:  
One pre-determined task for each agent — navigates to its goal location.
- ▶ MAPD — a “lifelong” version of multi-agent path finding:
  - ▶ A task can enter the system at any time.
  - ▶ Agents have to constantly attend to a stream of new tasks.



# MAPD Algorithms

## 1. Decoupled Task Assignment and Path Finding

- ▶ Token Passing (**TP**): Greedy task assignment and no task reassignment.
- ▶ Token Passing with Task Swaps (**TPTS**): Local task reassignment between two agents.

## 2. Centralized Task Assignment and Path Finding

### **CENTRAL**

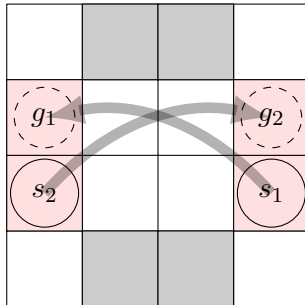
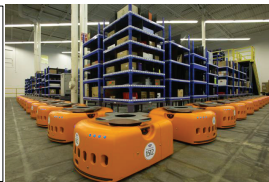
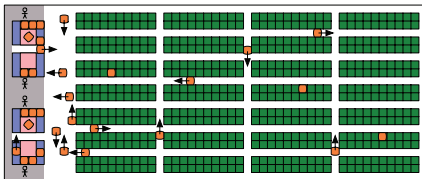
Roughly:

- ▶ Effectiveness:  $TP < TPTS < CENTRAL$
- ▶ Efficiency:  $CENTRAL < TPTS < TP$





# Tasks

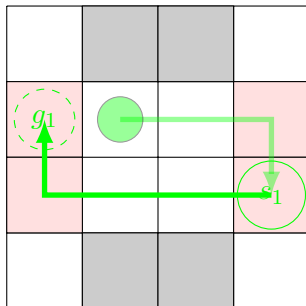




# Executing Task

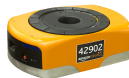
In order to execute a task, the agent has to move from its current location via the pickup location to the delivery location:

1. When the agent reaches the pickup location, it starts to execute the task.
2. When it reaches the delivery location, it finishes the task.

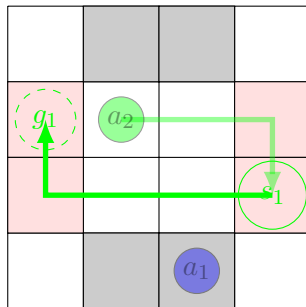




# Free Agents



**Free Agents:**

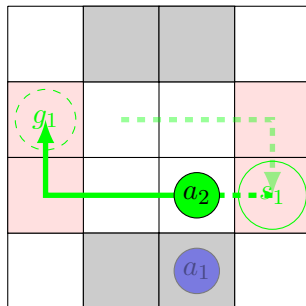




# Occupied Agents



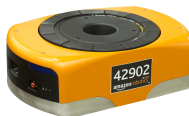
**Occupied Agents:**





# Assignment of Agents to Tasks

- ▶ A free agent can be assigned to any unexecuted task.



- ▶ An occupied agent has to finish executing its current task.





# Objective of MAPD

Finish executing each task as quickly as possible.





# Effectiveness of a MAPD algorithm

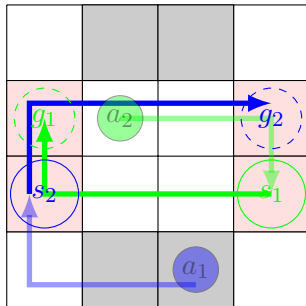
**Service time:** the average number of timesteps needed to finish executing each task after it enters the system.

An algorithm solves a MAPD instance  $\iff$  Service time of all tasks is bounded.





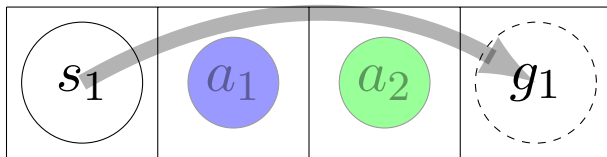
Service time:  $\frac{7+7}{2} = 7$





# Solvability

Not every MAPD instance is solvable.





# Well-Formed MAPD Instances

Being **well-formed** (based on [M. Cáp et al 2015]<sup>2</sup>): a sufficient condition that makes MAPD instances solvable.

Intuition: agents should only be allowed to rest (that is, stay forever) in locations, called **parking locations**, where they cannot block other agents.

---

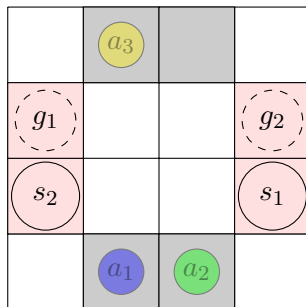
<sup>2</sup>M. Cáp, J. Vokřínek, and A. Kleiner. "Complete Decentralized Method for On-Line Multi-Robot Trajectory Planning in Well-formed Infrastructures". In: *International Conference on Automated Planning and Scheduling*. 2015, pp. 324–332.





# Parking Locations

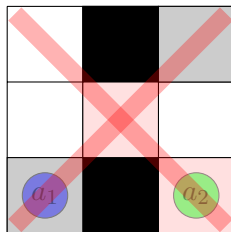
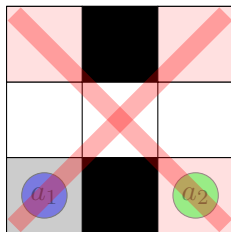
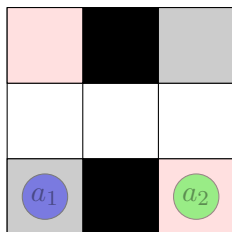
- ▶ **Task Parking Locations:** all pickup and delivery locations of tasks  
(storage locations, inventory stations, etc.)
- ▶ **Non-task Parking Locations:**
  - ▶ All initial locations of agents
  - ▶ Additional designated parking locations





# Well-Formed MAPD Instances

1. # tasks is finite;
2. # non-task parking locations  $\geq$  # agents;
3. For any two parking locations, there exists a path between them that traverses no other parking locations.







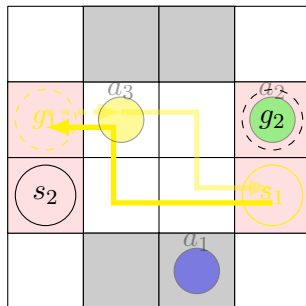


# A Running Example

Unexecuted Tasks:  $task_1, task_2$ .

Agent  $a_1$  and agent  $a_2$  are resting.

Agent  $a_3$  is assigned to  $task_1$  and on the way to the pickup location  $s_1$ .





# Token Passing (TP)

Based on an idea similar to Cooperative A\*<sup>3</sup>:

- ▶ Token: a synchronized shared block of memory that contains the current paths of all agents, set of unexecuted task, and agent assignments.
- ▶ Only one agent has access to the token at each time.
- ▶ Each agent assigns itself a task, plan its path, and passes the token to the next agent.

---

<sup>3</sup>D. Silver. "Cooperative Pathfinding". In: *Artificial Intelligence and Interactive Digital Entertainment*. 2005, pp. 117–122.





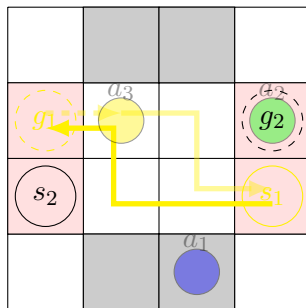




# TP: Running Example

Task Available for Assignment:  $task_2$ .

Agent  $a_1$  and agent  $a_2$  request for token.

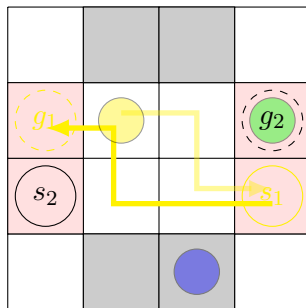




# TP: Agent $a_1$ 's Turn

## Agent $a_1$ Has Token

1. it cannot assign itself to any task because agent  $a_2$  rests in  $g_2$ , the only task available to it;
2. it has to rest in a parking location that will not create any deadlock;
3. it can continue to rest in its current location.

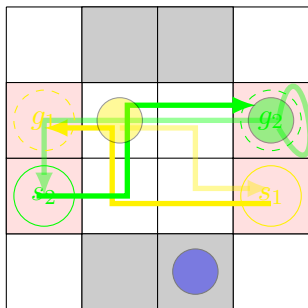




# TP: Agent $a_2$ 's Turn

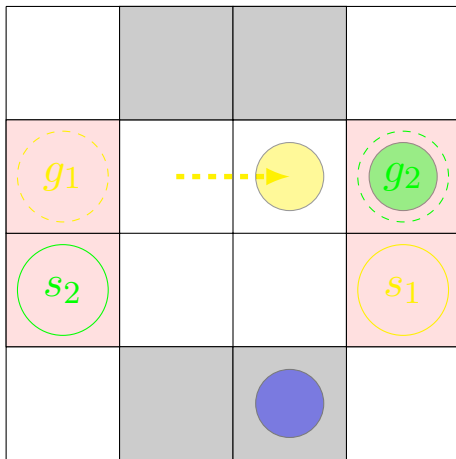
## Agent $a_2$ Has Token

1. it assigns itself to  $task_2$ ;
2.  $task_2$  is no longer available to other agents;
3. it plans a cost-minimal collision-free path to execute  $task_2$ .



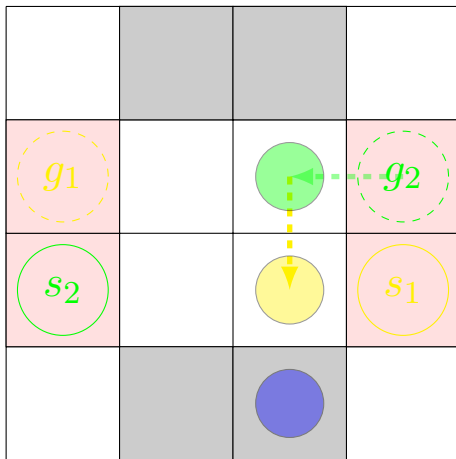


# TP: Animation



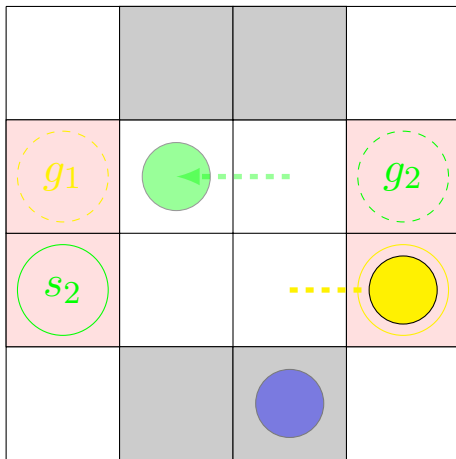


# TP: Animation



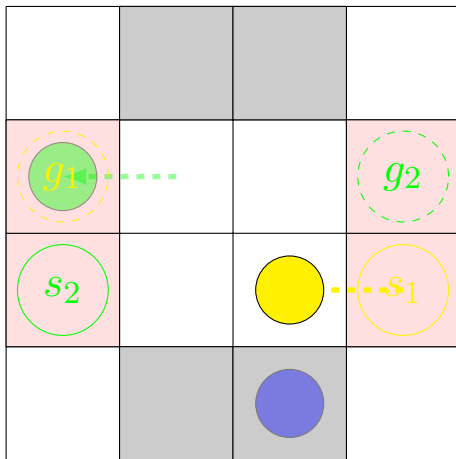


# TP: Animation



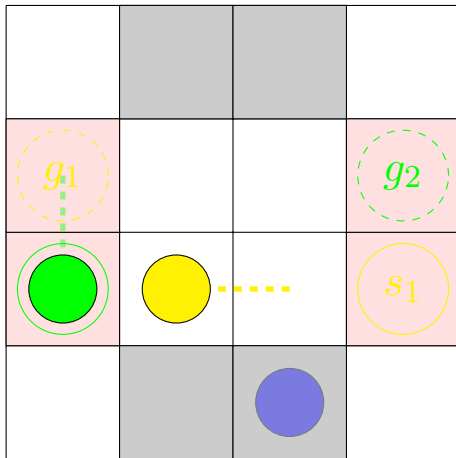


# TP: Animation



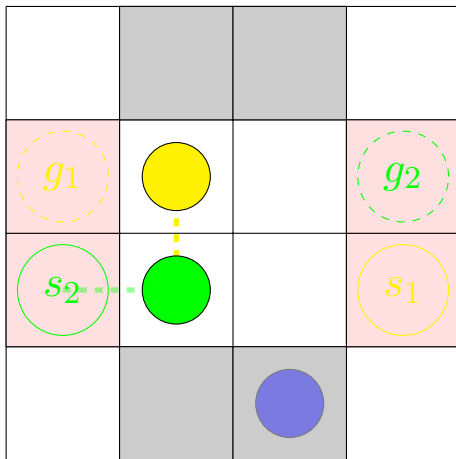


# TP: Animation



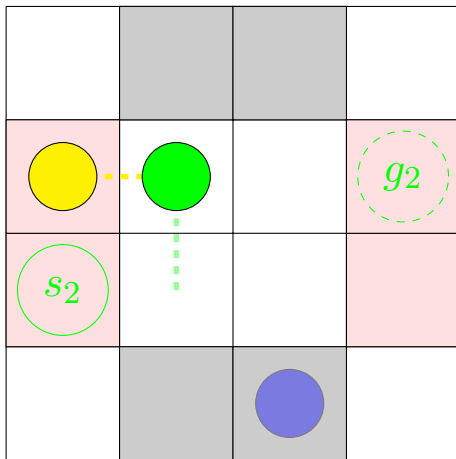


# TP: Animation



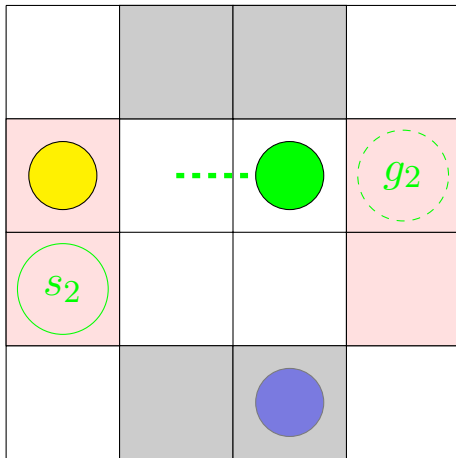


# TP: Animation



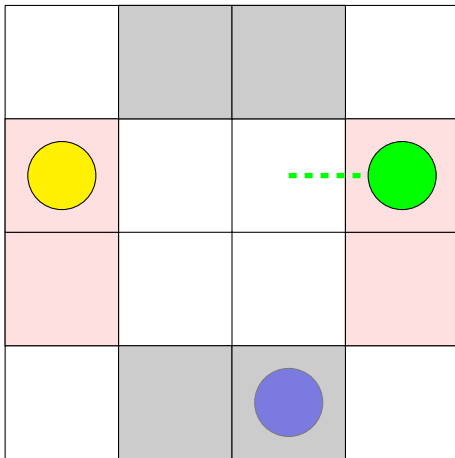


# TP: Animation





# TP: Animation





# TP: Completeness

## Theorem

*All well-formed MAPD instances are solvable, and TP solves them.*





# Improving the Effectiveness of TP

TP is simple but can be made more effective:

- ▶ A task with an assigned agent can be assigned a new agent (as long as the task has not been executed).





# Token Passing with Task Swaps (TPTS)

An agent is allowed to grab a task from another agent if it can finish the task earlier.

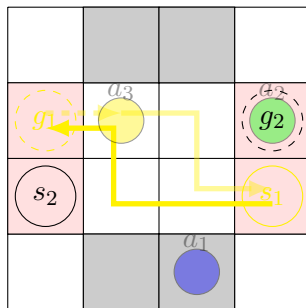




# TPTS: Running Example

Tasks Available for Assignment:  $task_1$ ,  $task_2$ .

Agent  $a_1$  and agent  $a_2$  request for token.

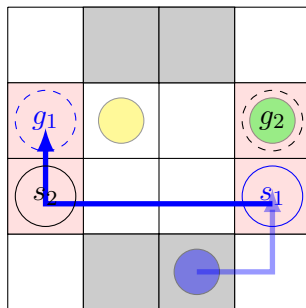




# TPTS: Agent $a_1$ 's Turn

Agent  $a_1$  has token.

Agent  $a_1$  grabs  $task_1$  from agent  $a_3$ .

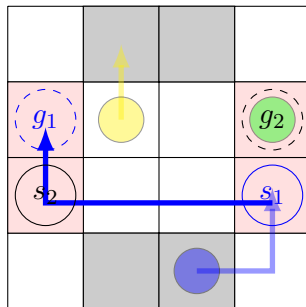




# TPTS: Agent $a_3$ Making Decisions

Agent  $a_3$  has token.

Agent  $a_3$  moves to a parking location that will not create any deadlock in the future.

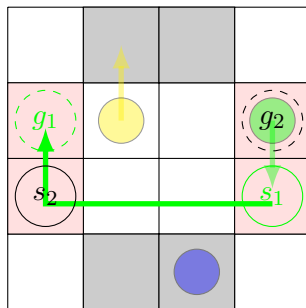




# TPTS: Agent $a_2$ 's Turn

Agent  $a_2$  has token.

Agent  $a_2$  grabs  $task_1$  from agent  $a_1$ .

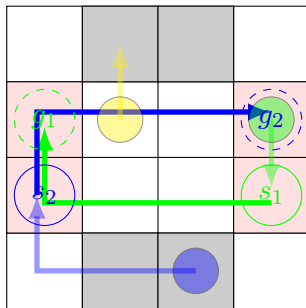




# TPTS: Agent $a_1$ Making Decisions

Agent  $a_1$  has token.

Agent  $a_1$  assigns itself to  $task_2$ .





# TPTS: Completeness

## Theorem

*TPTS solves all well-formed MAPD instances.*





# Centralized MAPD Algorithm: CENTRAL

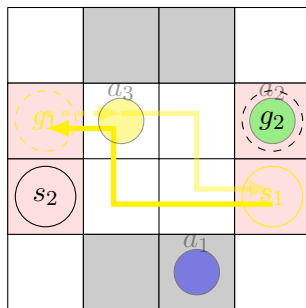
CENTRAL assigns agents to tasks in a centralized way:

1. assigns parking locations to all free agents using Hungarian method;
2. plans paths for all of them from their current locations to their assigned parking locations by solving the resulting “one-shot” multi-agent path-finding problem.



# CENTRAL: Running Example

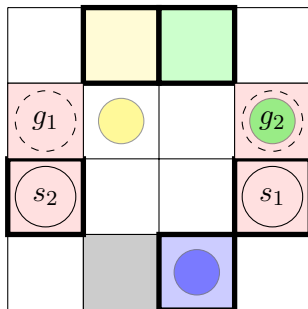
Tasks available for assignment:  $task_1$ ,  $task_2$





# CENTRAL: Candidate Parking Locations

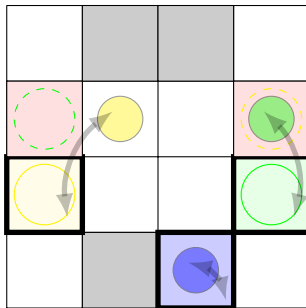
Pickup locations  $s_1$  and  $s_2$  + three additional “good” parking locations, one for each agent:





# CENTRAL: Assignment of Parking Locations to Agents

CENTRAL uses Hungarian method to find a cost-minimal assignment from parking locations to agents (pickup locations have priority over other parking locations):

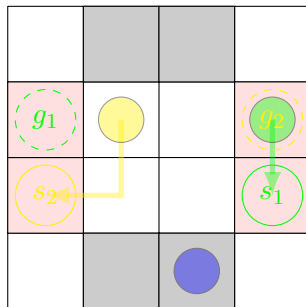




# CENTRAL: Path Finding

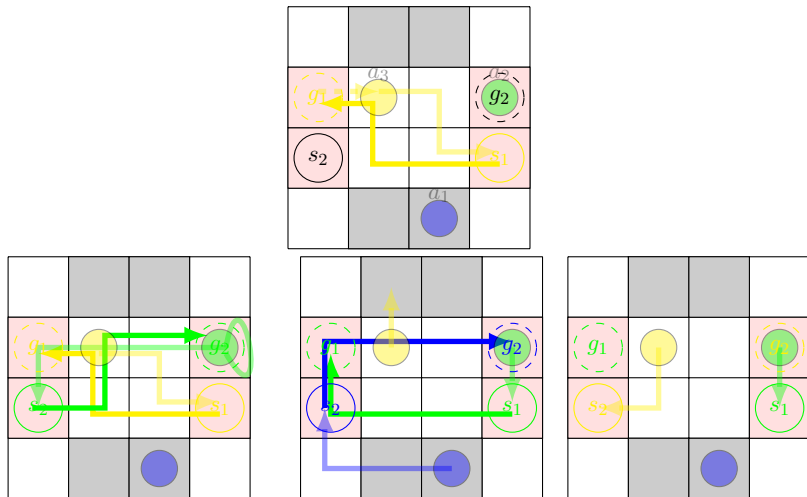
CENTRAL plans collision-free paths for all agents from their current locations to their assigned parking locations.

CENTRAL plans paths to delivery locations only when agents reach pickup locations.



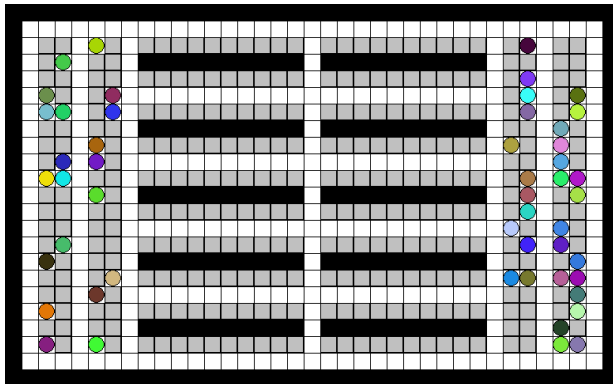


# Comparisons of Three Algorithms





# Small Simulated Warehouse Environment



**Figure:**  $21 \times 35$  4-neighbor grid with 50 agents. Gray cells are inventory stations and storage locations. Colored circles are the initial locations of agents.



# Experimental Results: 500 Random Tasks, 10 to 50 Agents

- ▶ Effectiveness

1. **Service Time:**

$CENTRAL < TPTS < TP$

2. **Throughput** – # tasks executed per 100 timesteps:

$TP < TPTS < CENTRAL$

3. **Makespan** – timestep when all tasks are finished:

$CENTRAL < TPTS < TP$

- ▶ **Runtime per Timestep:**

$TP < 10$  milliseconds

$TPTS < 200$  milliseconds

$CENTRAL < 4,000$  milliseconds





# Large Simulated Warehouse Environment

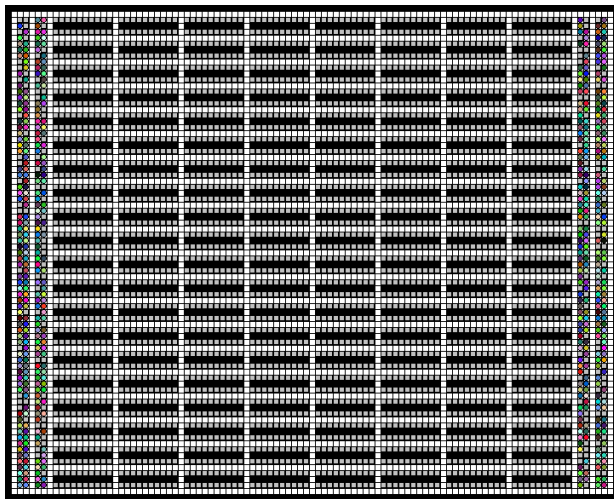


Figure:  $81 \times 81$  4-neighbor grid with 500 agents.



# Results for TP: 1000 Random Tasks, 100 to 500 Agents

100 agents:  $\sim 0.09$  seconds per timestep

500 agents:  $\sim 6$  seconds per timestep

agents	100	200	300	400	500
service time	463.25	330.19	301.97	289.08	284.24
runtime (milliseconds)	90.83	538.22	1,854.44	3,881.11	6,121.06





# Takeaways

MAPD: A “lifelong” version of multi-agent path finding.

Three Algorithms:

- ▶ Decoupled and complete for well-formed MAPD instances: TP, TPTS.
- ▶ Centralized: CENTRAL.

Task Assignment Effort:  $TP < TPTS < CENTRAL$

Effectiveness:  $TP < TPTS < CENTRAL$

Efficiency:  $CENTRAL < TPTS < TP$