

# Load Balancing in Distributed Multi-Agent Path Finder (DMAPF)

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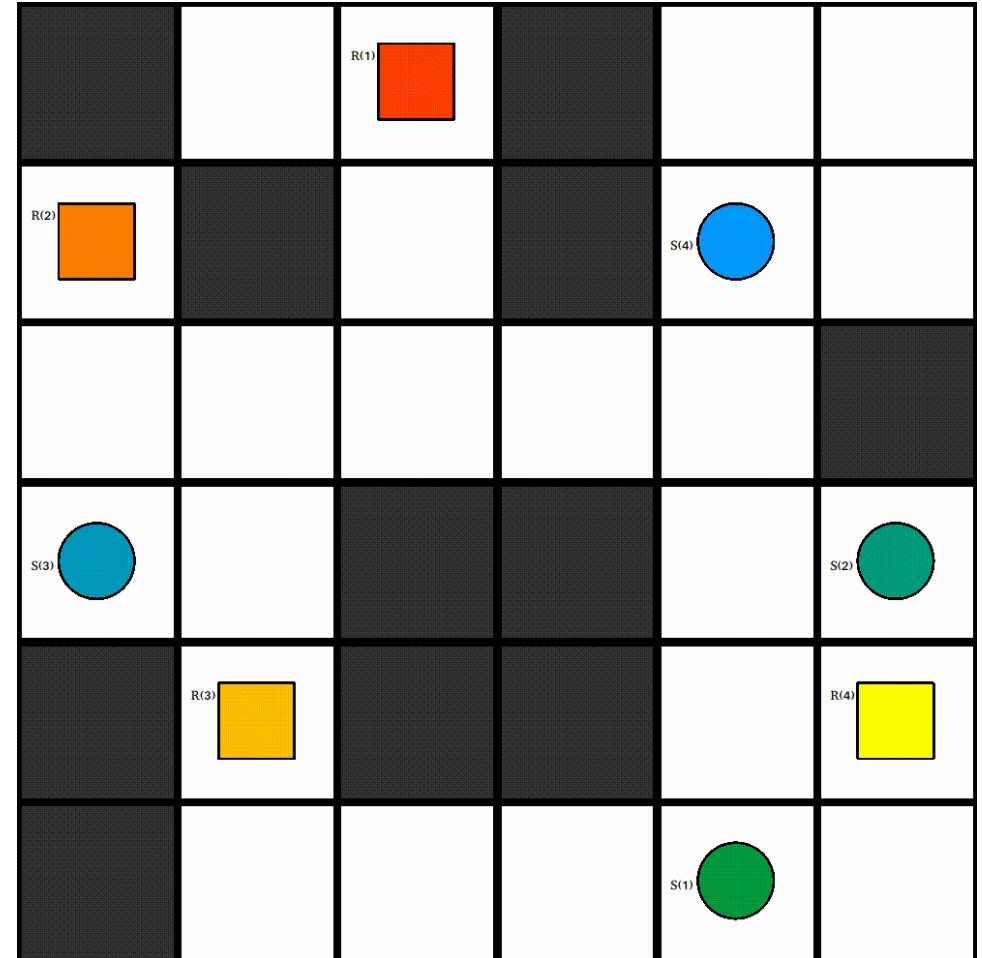
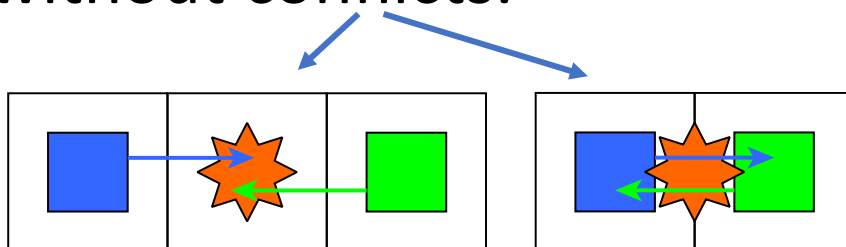
# Multi-Agent Path Finding (MAPF)

## Given

- A map (graph): vertices  & edges
- Agents
  - Start locations 
  - Goal locations 

## Find

- A sequence of actions that brings agents from  to  without conflicts.



# Applications of MAPF



Since 1996



Since 2000s

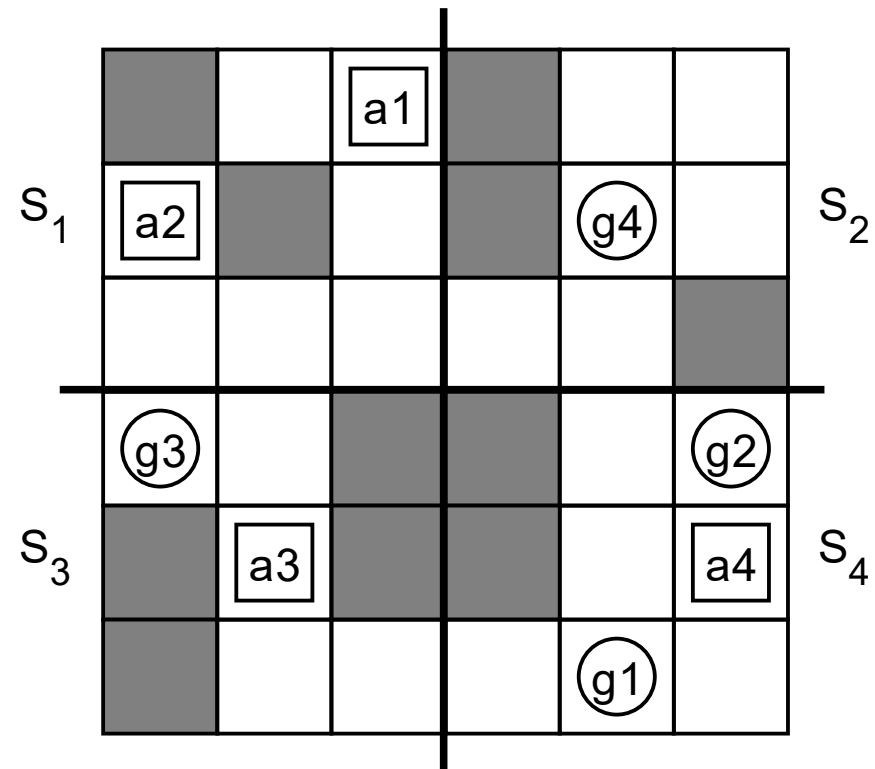
**The industries keep growing!**

# Distributed Multi-Agent Path Finder (DMAPF)

## Idea

1. Partition a given map into subproblems  $S_1, \dots, S_n$ .
2. Assign each set of subproblems to a solving process.
3. Solving processes work together to solve the problem.

For simplicity, we are going to assume that each of the subproblems  $S_1, \dots, S_4$  is assigned to a distinct solving process.



# How Solving Processes Work Together

First

A sequence of areas from start to goal

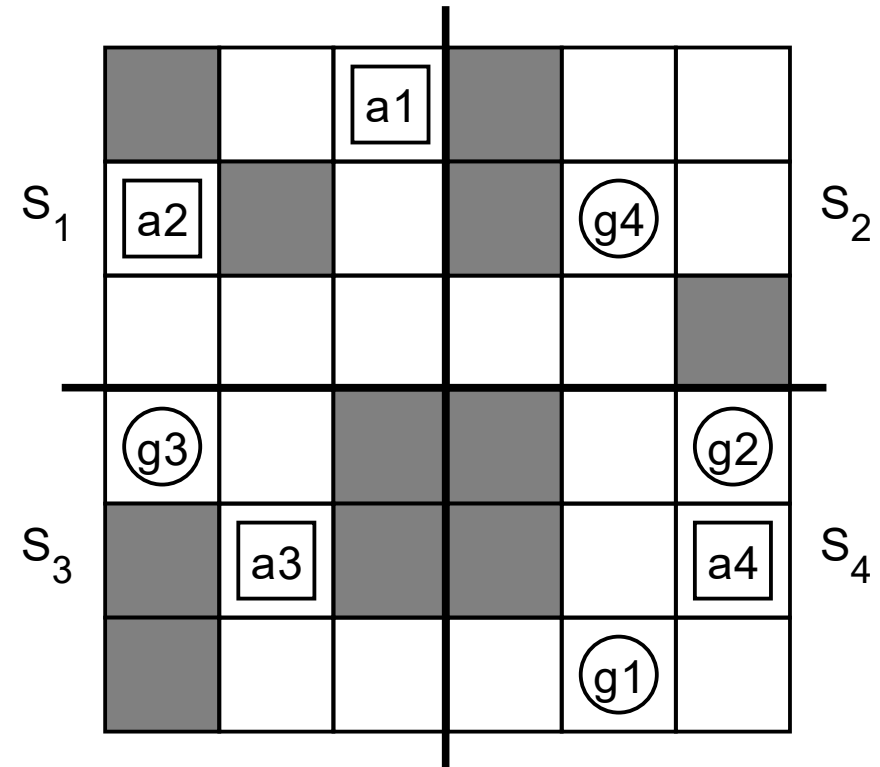
Make an abstract plan for each agent.

e.g.,

- a1:  $\langle S_1, S_2, S_4 \rangle$
- a2:  $\langle S_1, S_3, S_4 \rangle$
- a3:  $\langle S_3 \rangle$
- a4:  $\langle S_4, S_2 \rangle$

Answer Set  
Programming  
(ASP)

```
#external q(i).
1 {r(A2,i) : l(A1,A2)} 1 :- r(A1,i-1).
:- q(i), not r(A,i), g(A).
#show r/2.
```





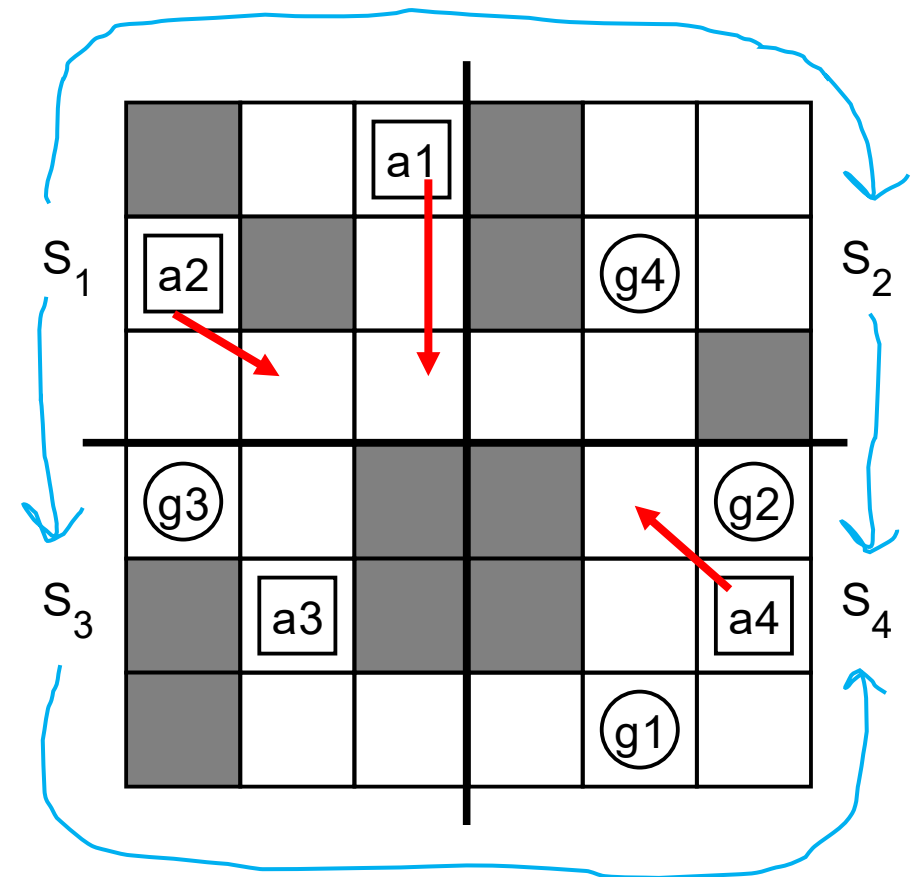
# How Solving Processes Work Together

Then

Follow the 3-step protocol:

## 1. Negotiation

- Decides which agent to migrate, and to which border location.
- Prioritize agents with higher number of remaining steps in the abstract plans.

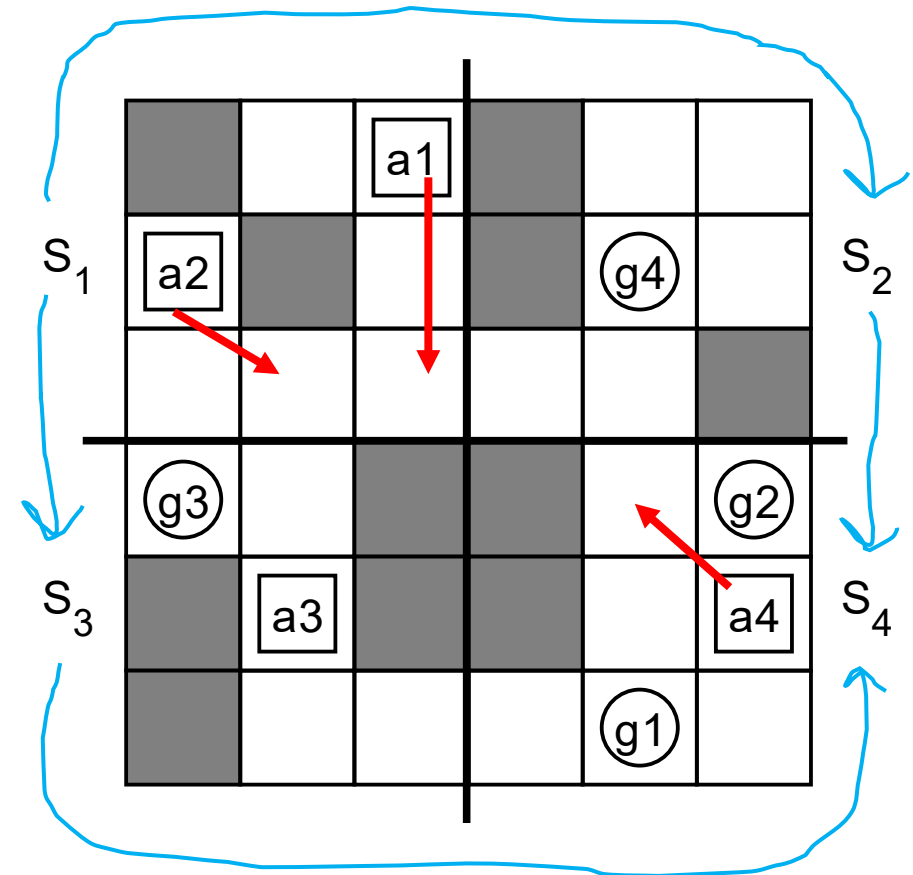
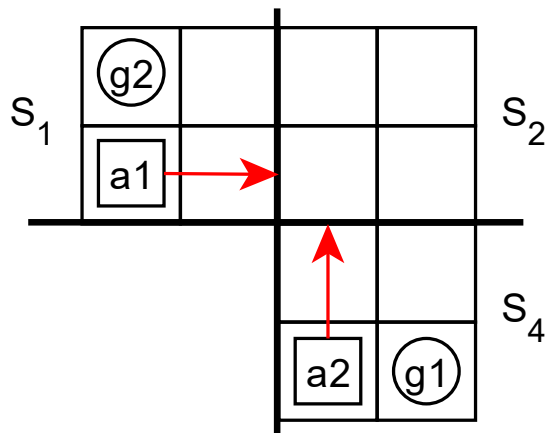


# How Solving Processes Work Together

Then

Follow the 3-step protocol:

1. Negotiation
2. Rejection
  - Prevents possible conflicts from the negotiation.
  - Regulate the area.



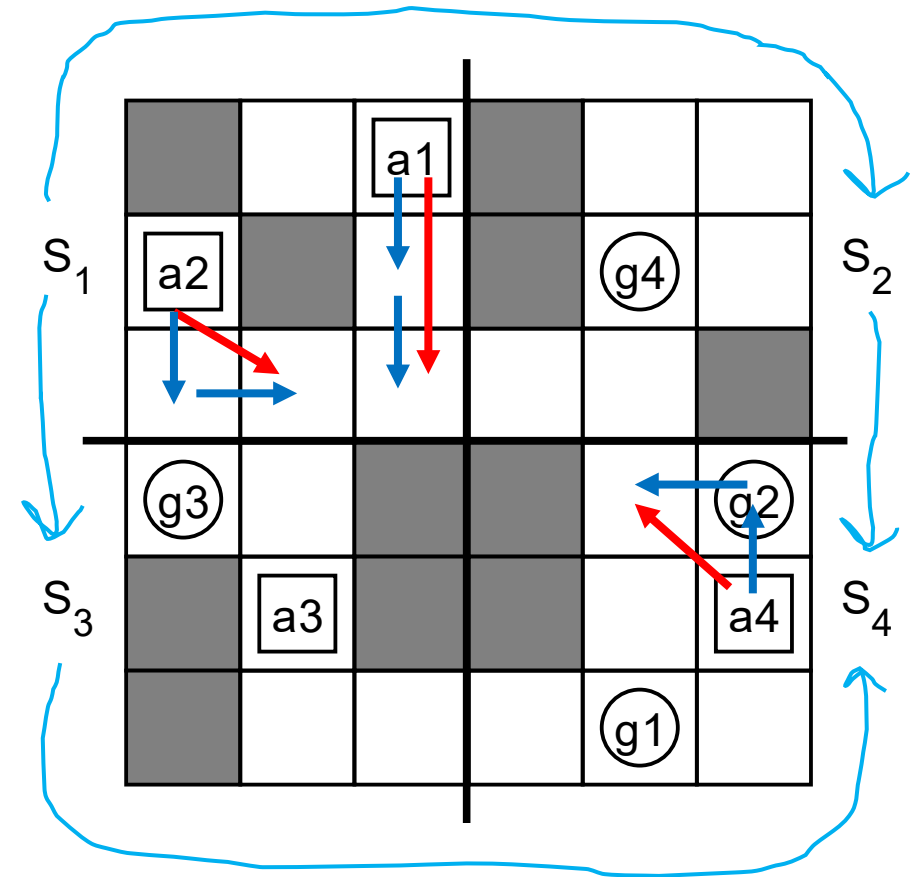
# How Solving Processes Work Together

Then

Follow the 3-step protocol:

1. Negotiation
2. Rejection
- Solving MAPF + Relaxation.

Canceling some migration





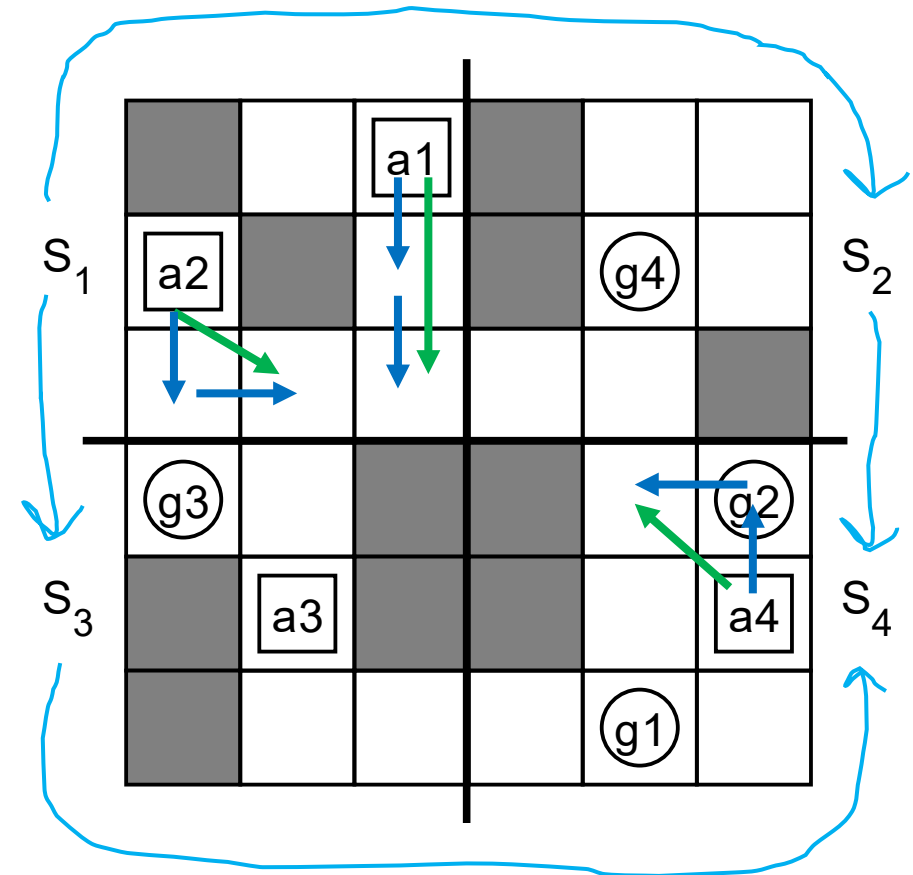
# How Solving Processes Work Together

Then

Follow the 3-step protocol:

1. Negotiation
2. Rejection
  - Solving MAPF + Relaxation.
3. Confirmation
  - Confirms agents that can actually migrate.

Loop with new start locations.



# Reducing Congestion in Abstract Level

## Observation

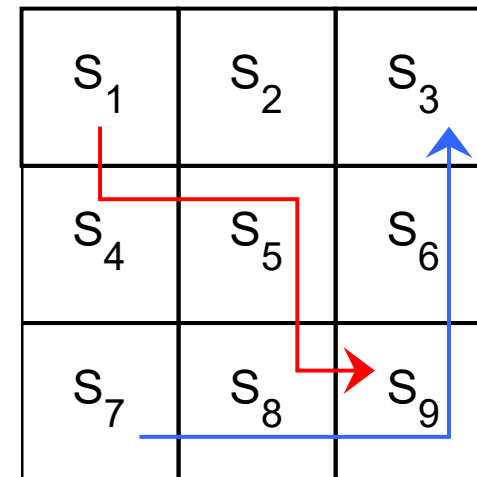
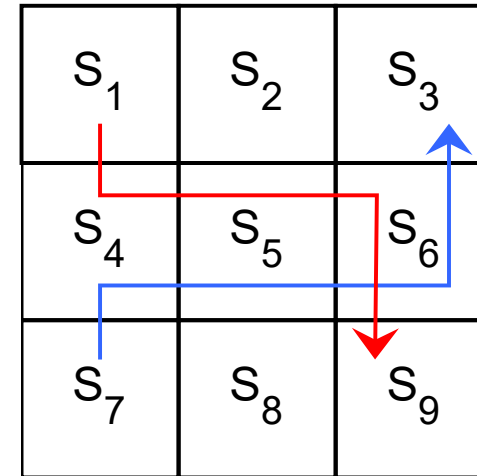
- Using blind search tends to cause congestion in some area.

↓

#agents/#vertices

## Idea

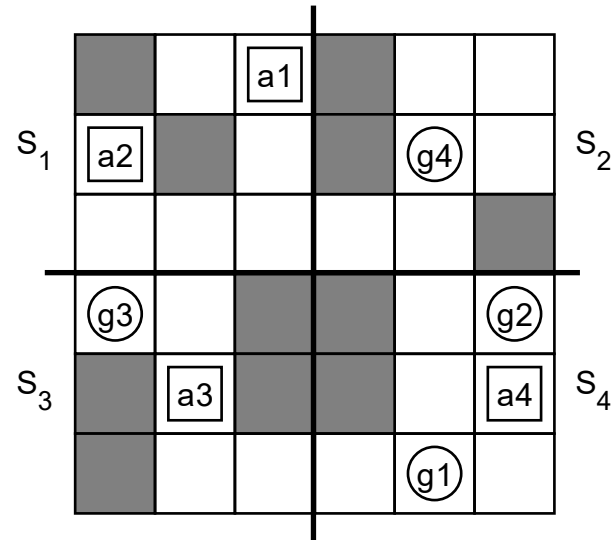
- Search while taking congestion into account.



# Uniform-Cost Search on Congestion Matrix

Abstract plans:

- N/A



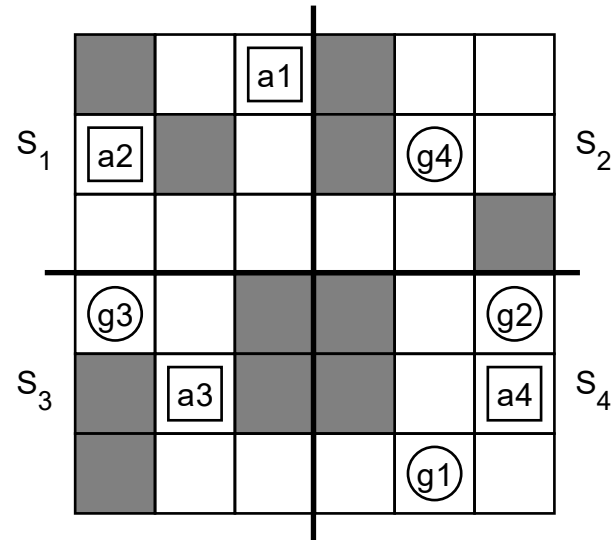
t = 0

$S_1$	0/7
$S_2$	0/6
$S_3$	0/5
$S_4$	0/7

# Uniform-Cost Search on Congestion Matrix

Abstract plans:

- $a1: \langle S_1, S_2, S_4 \rangle$



t = 0

$S_1$	0/7
$S_2$	0/6
$S_3$	0/5
$S_4$	0/7

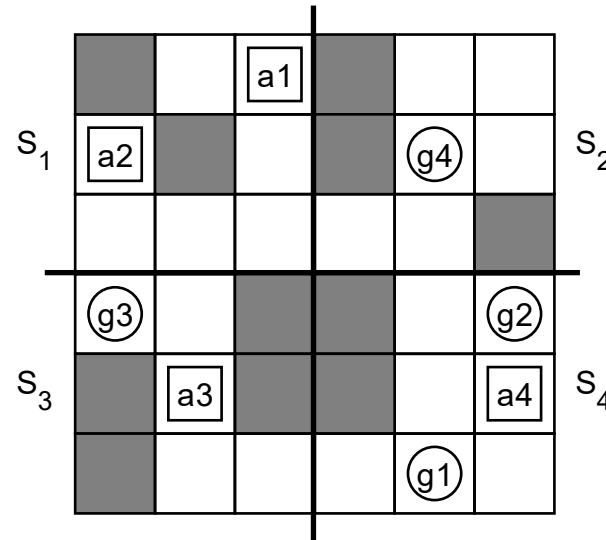
t = 0    1    2

$S_1$	1/7	0/7	0/7
$S_2$	0/6	1/6	0/6
$S_3$	0/5	0/5	0/5
$S_4$	0/7	0/7	1/7

# Uniform-Cost Search on Congestion Matrix

Abstract plans:

- a1:  $\langle S_1, S_2, S_4 \rangle$
- a2:  $\langle S_1, S_3, S_4 \rangle$

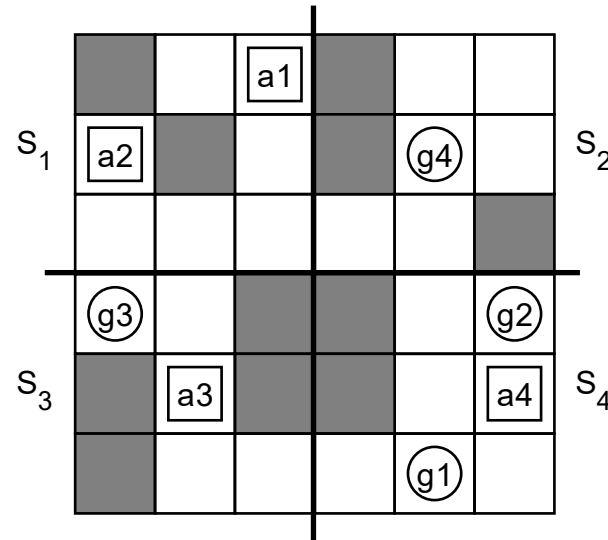


	t = 0	t = 0	1	2	t = 0	1	2
S <sub>1</sub>	0/7	1/7	0/7	0/7	2/7	0/7	0/7
S <sub>2</sub>	0/6	0/6	1/6	0/6	0/6	1/6	0/6
S <sub>3</sub>	0/5	0/5	0/5	0/5	0/5	1/5	0/5
S <sub>4</sub>	0/7	0/7	0/7	1/7	0/7	0/7	2/7

# Uniform-Cost Search on Congestion Matrix

Abstract plans:

- a1:  $\langle S_1, S_2, S_4 \rangle$
- a2:  $\langle S_1, S_3, S_4 \rangle$
- a3:  $\langle S_3 \rangle$



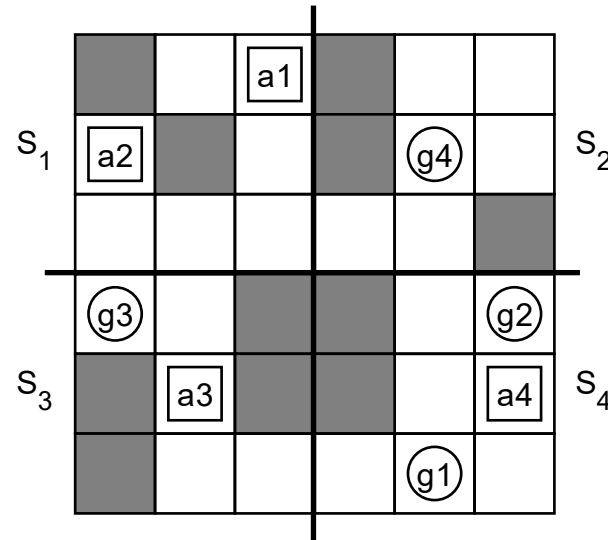
	t = 0	t = 0	1	2	t = 0	1	2	t = 0	1	2
S <sub>1</sub>	0/7	1/7	0/7	0/7	2/7	0/7	0/7	2/7	0/7	0/7
S <sub>2</sub>	0/6	0/6	1/6	0/6	0/6	1/6	0/6	0/6	1/6	0/6
S <sub>3</sub>	0/5	0/5	0/5	0/5	0/5	1/5	0/5	1/5	2/5	1/5
S <sub>4</sub>	0/7	0/7	0/7	1/7	0/7	0/7	2/7	0/7	0/7	2/7



# Uniform-Cost Search on Congestion Matrix

Abstract plans:

- a1:  $\langle S_1, S_2, S_4 \rangle$
- a2:  $\langle S_1, S_3, S_4 \rangle$
- a3:  $\langle S_3 \rangle$
- a4:  $\langle S_4, S_2 \rangle$



	t = 0	t = 0	1	2	t = 0	1	2	t = 0	1	2	t = 0	1	2
$S_1$	0/7	1/7	0/7	0/7	2/7	0/7	0/7	2/7	0/7	0/7	2/7	0/7	0/7
$S_2$	0/6	0/6	1/6	0/6	0/6	1/6	0/6	0/6	1/6	0/6	0/6	2/6	1/6
$S_3$	0/5	0/5	0/5	0/5	0/5	1/5	0/5	1/5	2/5	1/5	1/5	2/5	1/5
$S_4$	0/7	0/7	0/7	1/7	0/7	0/7	2/7	0/7	0/7	2/7	1/7	0/7	2/7

# Timeout Estimation Mechanism

Keep the time the problem is solved per agent; call this  $t_a$ .

- If the problem is solved within  $n \cdot t_a \cdot \varepsilon$ 
  - If some agent reach its assigned location;
    - update  $t_a \leftarrow t_s / n$
  - Otherwise;
    - update  $t_a \leftarrow f \cdot t_a$
- Otherwise
  - If some migrating agent has a location assigned;
    - remove its assigned location.
  - Otherwise;
    - terminate.

$n$ – the number of agents $\varepsilon$ – timeout tolerance factor $f$ – timeout penalty factor $t_s$ – solving time
--

# Exp 1. The Numbers of Solving Processes



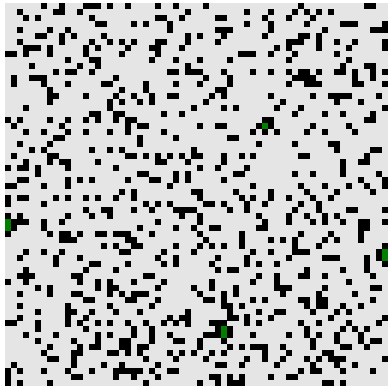
lak303d  
194x194,  
 $|V|=14,784$

$n$	Runtime (s)							
	$p = 4$	$p = 8$	$p = 12$	$p = 16$	$p = 20$	$p = 24$	$p = 28$	$p = 32$
200	32.1	25.1	21.8	20.0	19.4	<b>18.9</b>	20.5	19.6
400	97.0	75.1	63.1	66.2	<b>52.8</b>	56.2	53.8	63.5
600	214.2	158.7	129.3	127.5	<b>110.7</b>	113.3	116.0	120.0

Moving AI's Benchmark:

<https://movingai.com/benchmarks/mapf/index.html>

# Exp 2. The Size of Subproblems



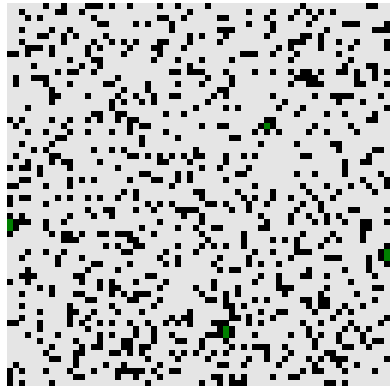
random-64-64-20  
64x64,  $|V|=3,270$

$n = 1000$

Makespan = The last time step where every agent reach its goal.  
Sum-of-Cost (SoC) = The total arrival time.

$v$	Runtime (s)	Makespan	SoC ( $\times 1k$ )	Success Rate
30	<b>39.9</b>	<b>864</b>	<b>558.4</b>	0.4
40	46.6	906	598.5	<b>1.0</b>
50	88.2	1041	627.9	0.8
60	80.0	1107	676.6	<b>1.0</b>
70	86.8	1070	633.6	<b>1.0</b>

# Exp 3. Timeout Sensitivity



random-64-64-20  
64x64,  $|V|=3,270$

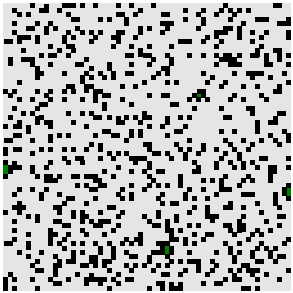
$n = 1000$

timeout tolerance factor

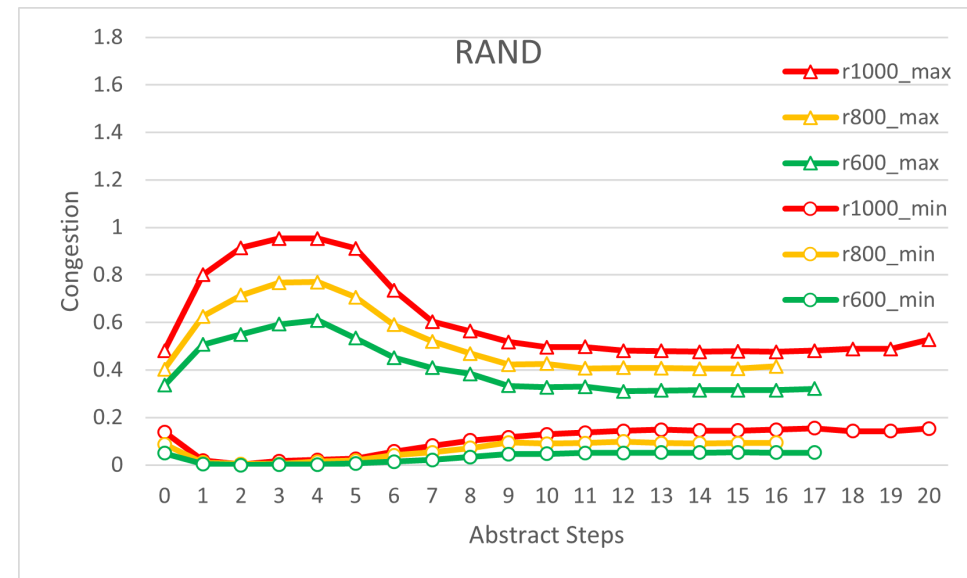
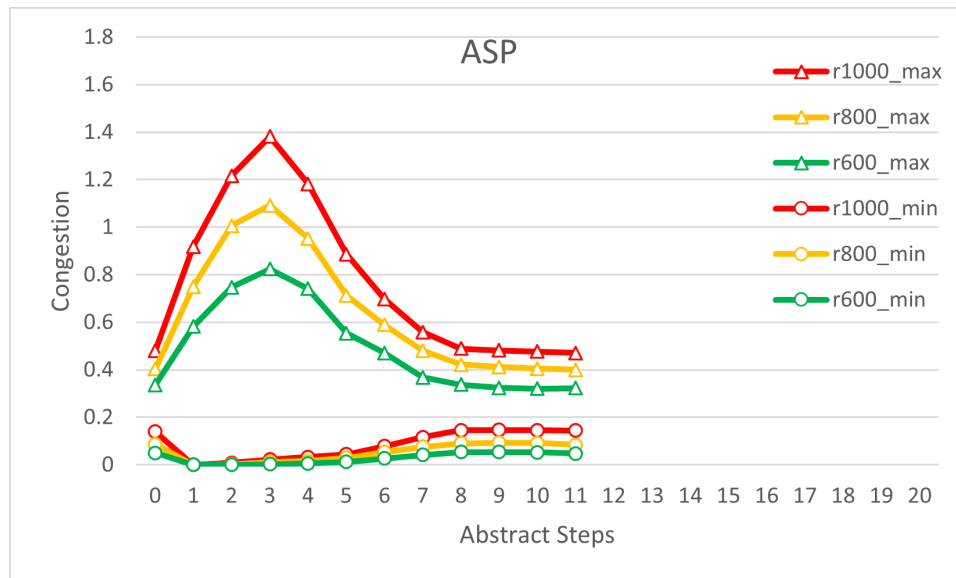
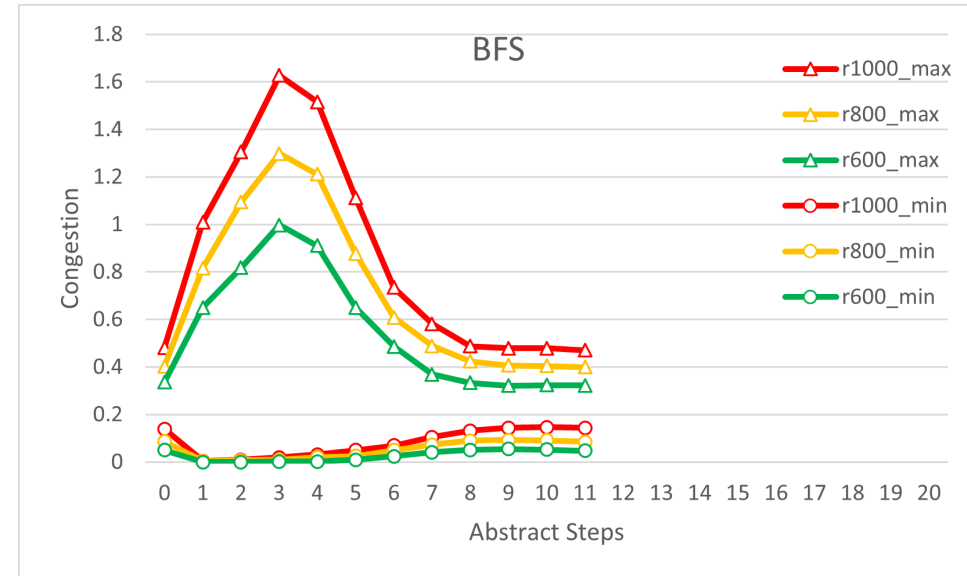
$\epsilon$	Runtime (s)	Makespan	SoC ( $\times 1k$ )	#Stops
4	51.9	906	611.9	25
6	49.2	906	600.1	13
8	48.6	<b>903</b>	607.3	7
10	46.6	906	<b>598.5</b>	3
12	<b>46.3</b>	906	603.2	<b>1</b>
14	49.1	908	606.8	2

# Exp 4. Congestion

Blind searches...



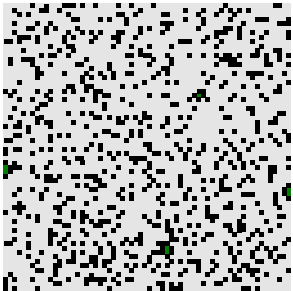
random-64-64-20  
64x64,  $|V|=3,270$



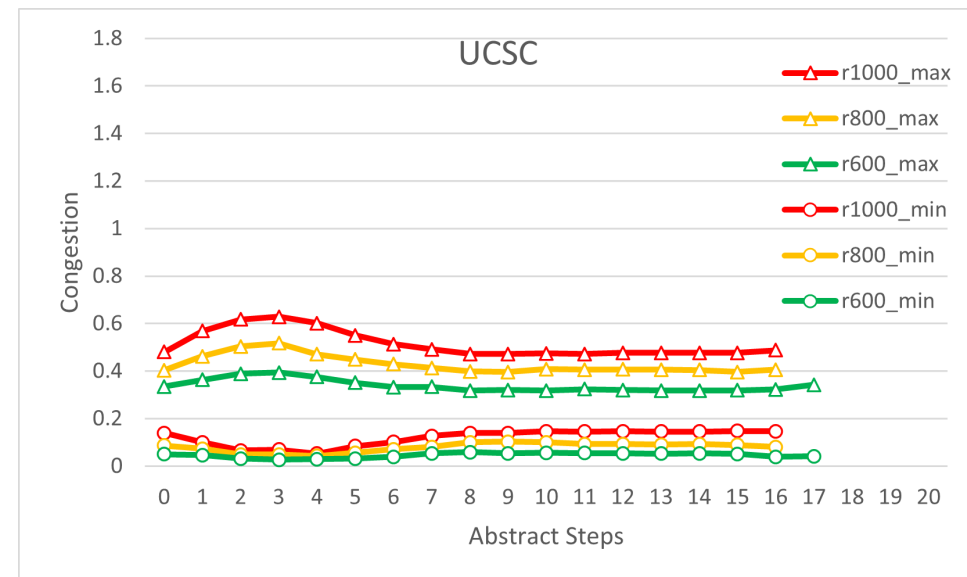
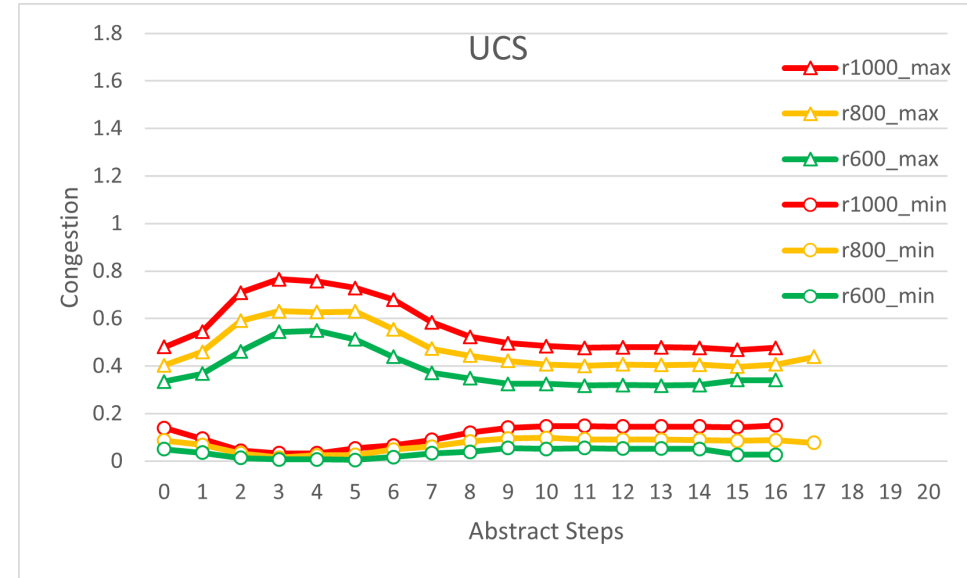


# Exp 4. Congestion

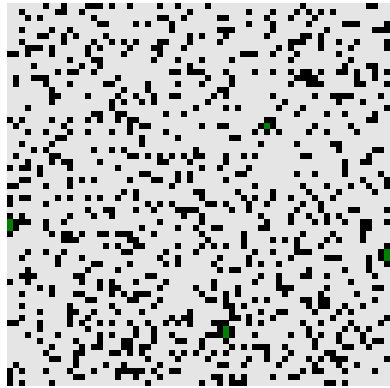
Not-so-blind searches...



random-64-64-20  
64x64,  $|V|=3,270$



# Exp 4. Congestion



random-64-64-20  
64x64,  $|V|=3,270$

$n$	Method	Runtime (s)		Makespan	SoC ( $\times 1k$ )	Success Rate
		Abs.	Total			
600	ASP	0.3	48.4	914	247.3	<b>1.0</b>
	BFS	<b>0.0</b>	110.2	1185	265.6	0.8
	RAND	<b>0.0</b>	19.8	673	236.8	<b>1.0</b>
	UCS	<b>0.0</b>	28.2	779	258.4	<b>1.0</b>
	UCSC	<b>0.0</b>	<b>16.1</b>	<b>564</b>	<b>207.2</b>	<b>1.0</b>
800	ASP	0.4	106.6	1241	511.7	0.6
	BFS	<b>0.0</b>	162.1	1367	525.3	0.1
	RAND	<b>0.0</b>	74.2	893	424.8	0.8
	UCS	<b>0.0</b>	56.1	934	451.0	0.4
	UCSC	<b>0.0</b>	<b>26.1</b>	<b>757</b>	<b>379.1</b>	<b>1.0</b>
1000	ASP	0.5	-	-	-	0.0
	BFS	<b>0.0</b>	-	-	-	0.0
	RAND	<b>0.0</b>	204.6	1058	703.9	0.1
	UCS	<b>0.0</b>	103.8	1109	740.2	0.1
	UCSC	<b>0.0</b>	<b>46.6</b>	<b>906</b>	<b>598.5</b>	<b>1.0</b>

# Exp 5. Comparisons

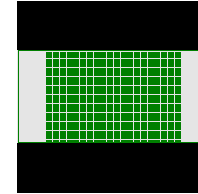
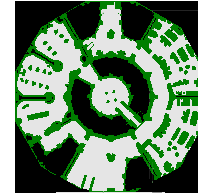
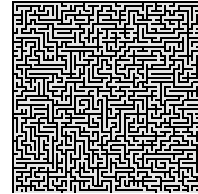
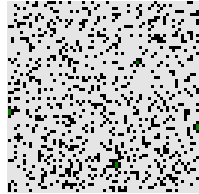
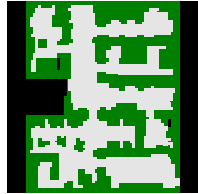
DMAPF

-A = with ASP;

-C = with CBSH2-RTC

-E = with EECBS;

-P = with PBS



Solver	<i>den312d</i> (2445)			<i>random</i> (3270)			<i>maze</i> (10858)			<i>lak303d</i> (14784)			<i>warehouse</i> (38756)		
	200	300	400	600	800	1000	100	200	300	200	400	600	600	800	1000
	— Runtime (seconds) —														
DMAPF-A	6.9	18.4	39.3	16.1	26.0	<b>46.6</b>	20.7	<b>40.7</b>	-	15.6	37.9	75.9	32.5	42.4	52.3
DMAPF-C	-	-	-	-	-	-	8.5	-	-	35.7	-	-	14.2	19.8	35.2
DMAPF-E	170.6	-	-	-	-	-	9.8	-	-	10.7	-	-	14.8	20.4	62.9
DMAPF-P	3.1	-	-	-	-	-	8.1	-	-	7.5	44.2	-	13.1	17.0	<b>20.8</b>
EECBS	<b>0.4</b>	<b>1.4</b>	<b>6.4</b>	<b>2.5</b>	<b>21.4</b>	141.7	<b>2.6</b>	135.8	<b>279.6</b>	<b>1.2</b>	<b>6.5</b>	<b>42.6</b>	<b>5.8</b>	<b>13.1</b>	22.0
PBS	15.1	217.5	-	-	-	-	50.5	-	-	17.2	266.9	-	9.4	26.1	57.2
	— Success Rate —														
DMAPF-A	<b>1.0</b>	<b>1.0</b>	0.7	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	0.8	0.0	<b>1.0</b>	<b>1.0</b>	0.6	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>
DMAPF-C	0.0	0.0	0.0	0.0	0.0	0.0	0.8	0.0	0.0	<b>1.0</b>	0.0	0.0	<b>1.0</b>	<b>1.0</b>	0.7
DMAPF-E	0.4	0.0	0.0	0.0	0.0	0.0	0.9	0.0	0.0	<b>1.0</b>	0.0	0.0	<b>1.0</b>	0.9	0.6
DMAPF-P	0.6	0.0	0.0	0.0	0.0	0.0	0.9	0.0	0.0	<b>1.0</b>	0.4	0.0	0.6	0.5	0.3
EECBS	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	0.9	<b>1.0</b>	0.9	<b>0.9</b>	<b>0.1</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>
PBS	<b>1.0</b>	0.8	0.0	0.0	0.0	0.0	<b>1.0</b>	0.0	0.0	<b>1.0</b>	0.3	0.0	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>

# Exp 5. Comparisons

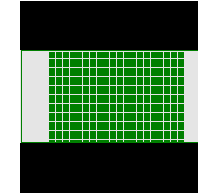
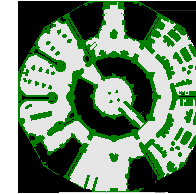
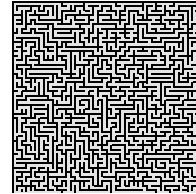
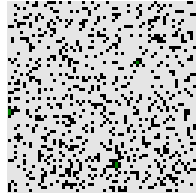
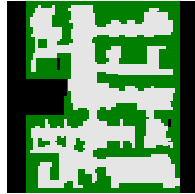
DMAPF

-A = with ASP;

-C = with CBSH2-RTC

-E = with EECBS;

-P = with PBS



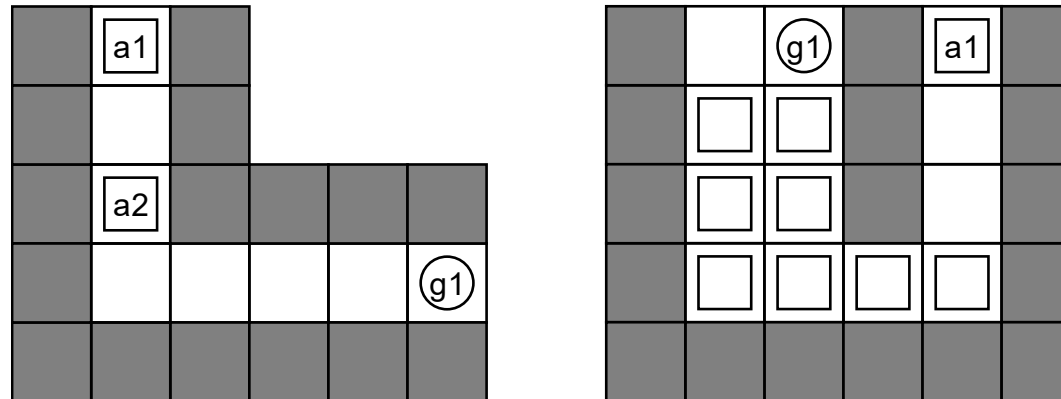
Solver	<i>den312d</i> (2445)			<i>random</i> (3270)			<i>maze</i> (10858)			<i>lak303d</i> (14784)			<i>warehouse</i> (38756)		
	200	300	400	600	800	1000	100	200	300	200	400	600	600	800	1000
	— Makespan —														
DMAPF-A	475	722	980	564	757	906	3075	3704	-	1014	1794	2774	748	774	795
DMAPF-C	-	-	-	-	-	-	3091	-	-	1017	-	-	753	781	804
DMAPF-E	643	-	-	-	-	-	3072	-	-	1033	-	-	780	828	893
DMAPF-P	477	-	-	-	-	-	3069	-	-	1016	1690	-	761	779	803
EECBS	180	288	<b>377</b>	<b>145</b>	<b>218</b>	<b>302</b>	<b>1474</b>	<b>1571</b>	<b>1702</b>	483	511	<b>583</b>	<b>451</b>	<b>455</b>	<b>457</b>
PBS	<b>132</b>	<b>158</b>	-	-	-	-	1475	-	-	<b>482</b>	<b>479</b>	-	<b>451</b>	<b>455</b>	<b>457</b>
	— Sum-of-Cost ( $\times 1000$ ) —														
DMAPF-A	51.8	117.3	248.9	207.2	379.1	598.5	181.4	511.8	-	112.9	362.2	794.9	233.1	331.3	448.3
DMAPF-C	-	-	-	-	-	-	187.2	-	-	110.7	-	-	230.7	335.8	443.2
DMAPF-E	56.4	-	-	-	-	-	175.4	-	-	110.4	-	-	236.7	356.6	498.8
DMAPF-P	54.1	-	-	-	-	-	176.2	-	-	111.8	358.7	-	231.8	334.8	449.1
EECBS	13.8	28.0	<b>46.9</b>	<b>34.8</b>	<b>60.6</b>	<b>101.6</b>	<b>56.1</b>	<b>119.7</b>	<b>191.4</b>	38.2	78.6	<b>131.1</b>	109.6	146.4	181.1
PBS	<b>11.6</b>	<b>19.1</b>	-	-	-	-	56.4	-	-	<b>37.9</b>	<b>74.1</b>	-	<b>109.5</b>	<b>146.2</b>	<b>180.9</b>

# Remarks

Comparing with EECBS and PBS...

- DMAPF can find a solution faster in problems with a lot of agents.

- DMAPF has a trouble when the map is not decomposed nicely.



- The solution quality from DMAPF is about 2-6 times worse.

