Plan merging project

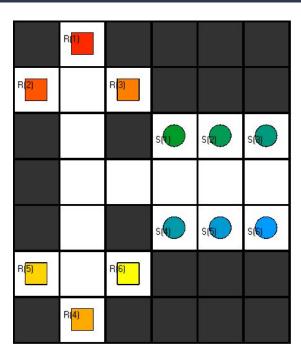
by Anton Rabe, Aaron Bishop and Louis Donath

Overview

- The Problem
- Plan merging
- Approaches
- Evaluation
- Conclusion

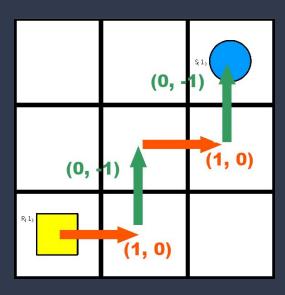
The Problem

- Given:
 - set of robots
 - set of shelfs
 - o plans between robot and shelf for each robot
- Wanted:
 - new plans without collisions



Plans

```
occurs (object(robot,1), action(move,(1,0)),0).
occurs (object(robot,1), action(move,(0,-1)),1).
occurs (object(robot,1), action(move,(1,0)),2).
occurs (object(robot,1), action(move,(0,-1)),3).
...
```



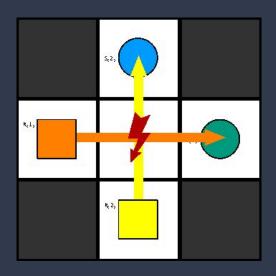
From plans to positions

```
occurs(object(robot,1),action(move,(1,0)),0).
occurs(object(robot,2),action(move,(0,-1)),0).
occurs(object(robot,1),action(move,(1,0)),1).
occurs(object(robot,2),action(move,(0,-1)),1).

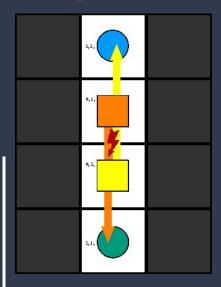
move(robot(R),(DX,DY),T):-
    occurs(object(robot,R),action(move,(DX,DY)),T).

position(R,(X+DX,Y+DY),T+1):-
    move(R,(DX,DY),T), position(R,(X,Y),T).
```

Vertex Collision



Edge Collision



The idea behind plan merging

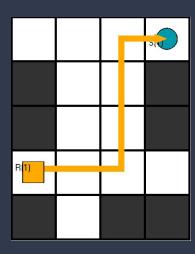
- Why not using an algorithm that explores the whole search space?
 - It could guarantee optimality
 - It would not require initial plans
 - o disadvantage: exponential runtime
- Plan merging
 - Only collisions need to be considered
 - Our approach:
 - no leaving of original plans
 - plan switching allowed
 - switching and waiting multiple times
 - Runtime stays low

Approaches

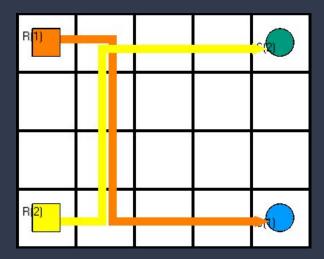
Waiting → Positions

```
\begin{split} & position\_(R,C,T+1,N+1) := position\_(R,C,T,N) \,, \; wait(R,\_,T) \,. \\ & position\_(R,C',T+1,N) := position(R,C',T-N+1) \,, \; not \; wait(R,\_,T) \,, position\_(R,C,T,N) \,. \end{split}
```

```
\{wait(R,C,T)\} :- position(R,C,T).
```



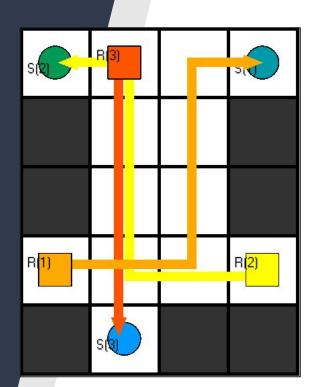
Switching → **Positions**



The incremental approach

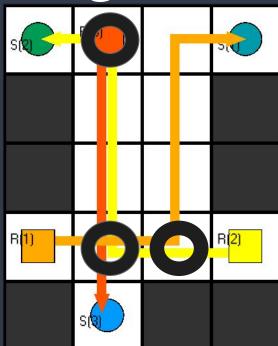
- 1. limit possibilities for assigning switching/ waiting positions
- 2. define a heuristic for optimization
- 3. Recursively execute the logic program on previous output until a fixed point is reached
- 4. Only select the optimal model at each step (greedy approach)

Plan-switching



Incremental plan switching

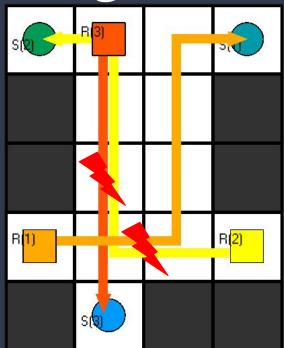
Limit switching positions to earliest intersections of plans



Incremental plan switching

 Only allow pairwise switching if two robots have a collision

```
{switch(R,R')} :- collision(R,R',T).
switch(R',R) :- switch(R,R').
:- switch(R,R'), switch(R,R''), R'!=R''.
```



incremental plan switching

Only remove edge collision-like collisions

 To prevent generating new edge collisions during waiting remove all edge-collision like intersections.

```
edge collision possible(R,R',T) :- intersect(R,R',C,T,T'), intersect(R,R',C',T+1,T'-1).
```

incremental plan switching

Minimize horizon and number of positions in final plan

```
horizon(T_MAX) :- T_MAX == #max{T : position_(R,C,T,X,Y)}.
#minimize{T@5 : horizon(T)}.

#minimize {1@4,T,R : position_(R,C,T,R',D)}.

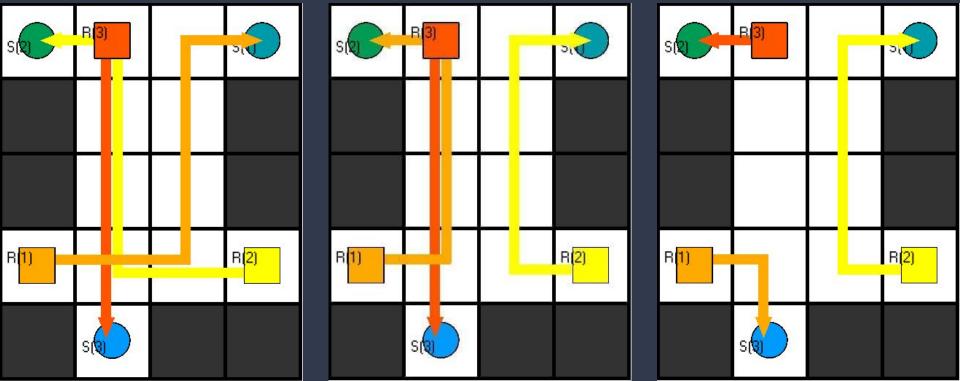
#minimize {1@3,R,R' : final_collision(R,R',T)}.

#minimize {1@1,R,R' : final_overtake_edge_collision(R,R',T,T')}.

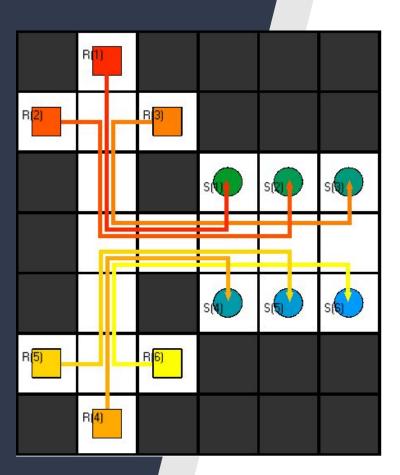
#minimize {1@2,R,R' : final_overtake_edge_collision(R,R',T,T')}.

overtake_edge_collision(R,R',T,T')}.
```

incremental plan switching



waiting



incremental waiting

Only allow one wait per robot one step before the earliest intersection with another plan.

incremental waiting

Minimize the squared distance between collision and last timestep.



One model approach

Calculate waiting positions deterministically robots with longer paths ahead have higher priority.

One model approach

Wait A times at the time of a collision for A robots with higher priority. Only count robots that meet for the first time

```
do_wait(R,C,T-1,A) :- A == #count{1,R' : earliest_collision(R,R',C,T), higher(R,R')},
earliest_collision(R,R'',C,T).
```

Combining waiting and switching

1. apply plan switching

2. apply preventive plan switching

3. choose a waiting approach/ can be done automatically

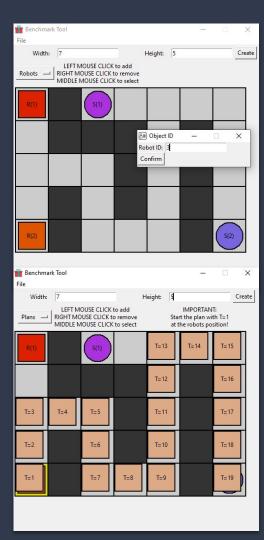
Evaluation

Benchmarks

- 1. for testing and verifying quality
- 2. to compare approaches

Benchmark Tool

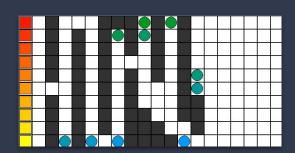
- custom benchmark creation tool for the groups
- 2. in Python with intuitive GUI
- exports/imports correct format
- 4. saving time

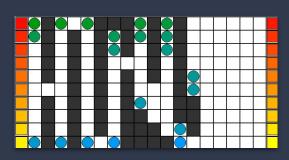


Metrics

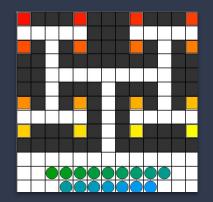
- 1. Satisfiability
 - no model
- 2. Validity
 - validity_checker.lp
- 3. Performance
 - total computation time
- 4. Optimality
 - not similarity
 - sum of all paths (SOAP)
 - or \triangle SOAP = new SOAP orig. SOAP

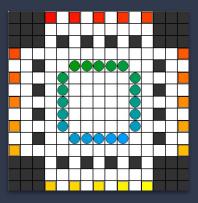
Benchmark Selection

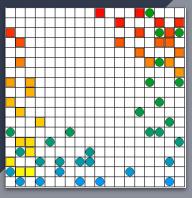


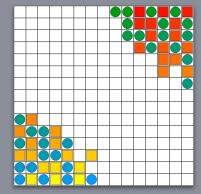












Approach Selection

- 1. Naive Layer & Choice rule approaches too slow
- 2. Sequential Approaches yielded highest performance



- 1) Preventive Plan Switching + Incremental Waiting
- 2) Preventive Plan Switching + Deterministic Waiting

Internal Comparison

- 1. deterministic vs. non-deterministic incremental waiting
- 2. look for drastic differences between them
- 3. what is causing that difference?



Attributes of a Benchmark Instance

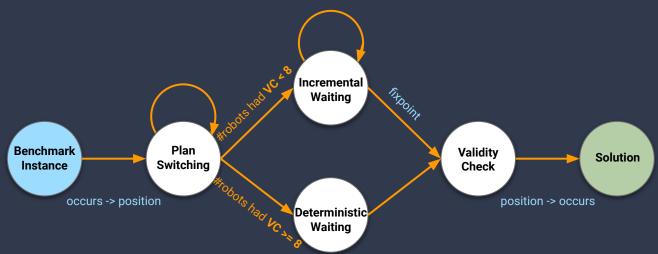
- Benchmark instances have different attributes
- 2. look at preprocessed benchmark instances
- 3. find attributes that are correlated or causal for the difference



BM1 1 BM2 2		5	iongest patri	II VCS					
									n robots had fake ecs
		12	5 6	0	0	0	0	0	0
BM3 2		10	5	1	2	0			
BM4 4		8	2	1	4	0	0	0	0
BM5 2		10	5	2	2	0	0	0	0
BM6 4		12	3	2	4	0	0	0	0
BM7 3		15	5	3	3	0	0	0	0
BM8 6		48	9	9	6	0	0	0	0
BM9 3		19	7	3	2	0	0	0	0
BM10 3		11	4	0	0	0	0	0	0
BM11 3		17	6	4	2	0	0	0	0
BM12 2		13	7	0	0	0	0	0	0
BM13 1		341	47	82	8	0	0	0	0
BM14 2		326	32	32	6				0
BM15 2	2	4	2						0
BM16 3	3	11							0
BM17 7		32							0
BM18 4	1	20		6	4				0
BM19 8	3	56		12	8				0
BM20 2	20	124	7	8	8				0
BM21 1	12	56		4	8				0
BM22 3	32	983	36	122	32	0	0	0	0
BM23 1	16	327	22	56	16	0	0	0	0
BM24 8	3	116	16	26	8	0	0	0	0
BM25 3	30	538	26	18	9				0
BM26 0)								0

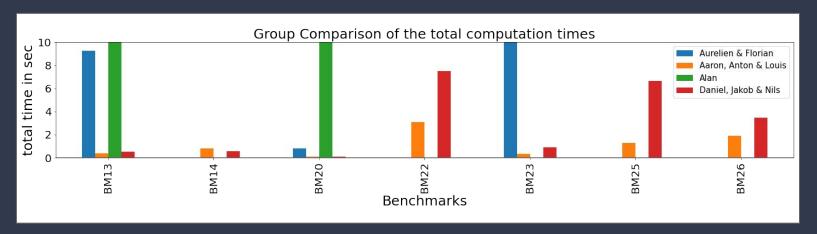
Final Approach

1. Merger class implements automatic merge method



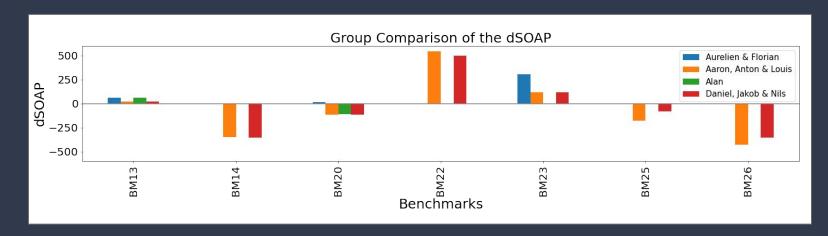
Group Comparison (1)

1. total computation time on large benchmarks



Group Comparison (2)

1. difference in sum of all positions (dSOAP) on large benchmarks



Conclusion

- Sequential approach
 - plan switching (preprocessing)
 - waiting

- Greedy Incremental Approach
 - repeatedly optimizing heuristics
 - o trade-off: speed vs. optimality

- Combined Approach
 - Why incremental waiting?
 - deterministic waiting might have risks
 - not tested on random BMs
- Outlook
 - improve deterministic waiting
 - test on random benchmarks

We'll gladly answer all of your questions:)

Questions?

References

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