# CMPT 417 - Individual Project

Multi-Agent Path Finding

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## 1 Space-Time A\*

### 1.1 Searching in Space-Time Domain

These are the results after modifying single\_agent\_planner.py to support temporal constraints, pertaining to Section 1.1 of the lab.

Figure 1 presents similar animation results.<sup>1</sup>

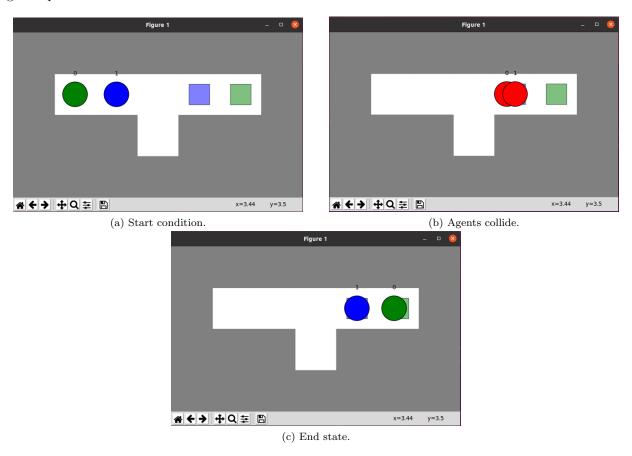


Figure 1: Animation after modifying single\_agent\_planner.py.

 $<sup>^{1} \</sup>rm https://tex.stackexchange.com/questions/148438/putting-two-images-beside-each-other and the stacked properties of the stacked properties of$ 

Figure 2 presents the terminal output:

Figure 2: Terminal output after modifying single\_agent\_planner.py.

### 1.2 Handling Vertex Constraints

These are the results after modifying single\_agent\_planner.py to handle vertex constraints, pertaining to Section 1.2 of the lab.

Using the example constraint provided in the lab, Agent 0 stops before (1,5) at time step 4. Figure 3 presents the animation where Agent 0 stops.

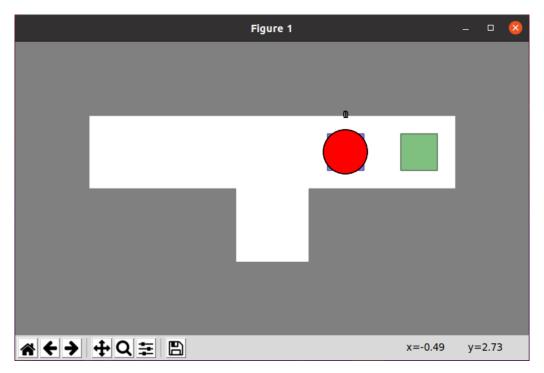


Figure 3: Agent 0 at time step 4 after adding vertex constraints.

Figure 4 shows the terminal output:

```
***Import an insta<u>nce</u>***
Start locations
0000000
 0 1 . . .
            @
   @
        @
          @
     0 0 0 0
   @
Goal locations
0000000
        1 0 @
 @
   @
     . @ @ @
 000000
***Run Prioritized***
Found a solution!
CPU time (s):
Sum of costs:
[[(1, 1), (1, 2), (1, 3), (1, 4), (1, 4), (1, 5)], [(1, 2), (1, 3), (1, 4)]]
***Test paths on a simulation***
COLLISION! (agent-agent) (0, 1) at time 3.4
           (agent-agent) (0, 1) at time 3.5
COLLISION!
           (agent-agent) (0, 1) at time 3.6
(agent-agent) (0, 1) at time 3.7
COLLISION!
COLLISION!
                         (0,
COLLISION!
           (agent-agent) (0, 1) at time 3.8
COLLISION!
           (agent-agent) (0, 1) at time 3.9
COLLISION!
                          (0, 1)
           (agent-agent)
                                 at time
                          (0, 1) at time
COLLISION!
           (agent-agent)
COLLISION!
           (agent-agent) (0,
                              1)
                                 at time 4.2
           (agent-agent) (0,
(agent-agent) (0,
COLLISION!
                              1)
                                  at
                                     time
COLLISION!
                              1)
                                 at time 4.4
COLLISION!
           (agent-agent)
                          (0, 1)
                                 at time 4.5
COLLISION!
           (agent-agent) (0, 1)
                                 at time 4.6
COLLISION!
           (agent-agent)
                          (0,
                              1)
                                 at time
           (agent-agent)
                          (0, 1)
COLLISION!
                                 at time
COLLISION!
           (agent-agent) (0,
                              1)
                                 at time 4.9
           (agent-agent)
COLLISION!
                          (0,
                               1)
                                 at
                                     time
           (agent-agent) (0, 1) at time 5.1
COLLISION!
COLLISION!
           (agent-agent)
                         (0, 1)
                                 at time 5.2
COLLISION!
           (agent-agent)
                          (0,
                              1)
                                  at
                                     time
                              1)
COLLISION!
           (agent-agent)
                                          5.4
                          (0,
                                 at time
COLLISION!
           (agent-agent)
                          (0,
                              1)
                                 at time 5.5
COLLISION! (agent-agent) (0,
                              1) at time 5.6
```

Figure 4: Terminal output after modifying single\_agent\_planner.py to handle vertex constraints.

#### 1.3 Handing Goal Constraints

These are the results after modifying single\_agent\_planner.py to handle goal constraints, pertaining to Section 1.4 of the lab.

With the original goal condition and the constraint where Agent 0 is prohibited from being at its goal location at time step 10, the animation ends prior to time step 10. Therefore, I have modified the goal conditions such that:

- First check if there are constraints.
- If there are constraints, check if a solution is found. If found, only return the path once the current time step matches the time step of the constraint with the latest time step.
- $\bullet$  If there are no constraints, check if a solution is found. If found, return the path.

The following shows the pseudo-code of the modifications:

```
if (number of constraints) > 0:
    if (current location = goal location & UNLOCK): return path
    if (current time step + 1 = latest constraint time step): UNLOCK
else:
    if (current location = goal location): return path
```

In the code implementation, variables lock and n\_constraints have been created to keep track of the locking mechanism and number of constraints (respectively) as shown in the pseudo-code. Note that the constraints table have been ordered<sup>2</sup> starting from the constraints with the earliest time step to the latest time step; therefore, variable latest constraint time step is easily specified because it is at the end of the table.

In the condition where constraints exist, the second if-statement indicates current time step + 1. The + 1 stems from the fact that the path should be returned as soon as the constraint with the latest time step is met. Without it, an extra iteration of the overarching for-loop is conducted because the lock is only set to UNLOCK once the time step matches the time step of the constraint with the latest time step. Increasing the value would cause the animation to terminate prior to the goal constraint being met (at least visually), and having negative values would cause the animation to run longer than it needs to.

### 1.4 Designing Constraints

These are the results after modifying single\_agent\_planner.py to find collision-free paths with a minimal sum of path lengths, pertaining to Section 1.5 of the lab.

The set of constraints used to create a collision-free path between the two agents is as follows:

- 1. Agent 1 cannot be at (1,2) at time step 2.
- 2. Agent 1 cannot be at (1,3) at time step 2.
- 3. Agent 1 cannot be at (1,4) at time step 2.

This forces Agent 1 to navigate towards (2,3) at time step 2, letting Agent 0 pass through and avoiding any collisions. Figure 5 shows the animated solution where Agent 1 navigates towards (2,3):

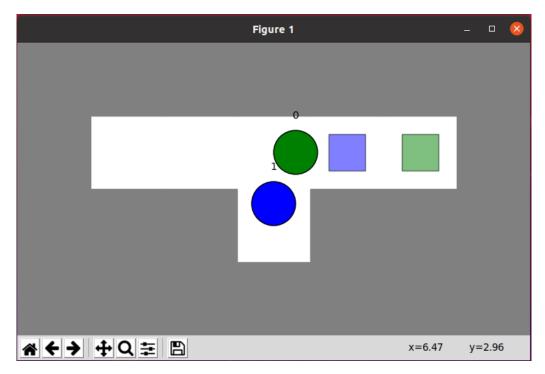


Figure 5: Agent 1 navigating towards (2,3).

<sup>&</sup>lt;sup>2</sup>https://stackoverflow.com/questions/72899/how-do-i-sort-a-list-of-dictionaries-by-a-value-of-the-dictionary

Figure 6 shows the terminal output:

Figure 6: Terminal output after designing collision-free constraints.

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From the terminal output, the (minimal) sum of path lengths is 8.

## 2 Prioritized Planning

#### 2.1 Addressing Failures

These are the results after modifying prioritized.py to handle edge, vertex and goal constraints, pertaining to Section 2.4 of the lab. Note that the constraints are implemented according to the given instructions in Section 2.1 to Section 2.3.

Unlike in exp2\_1.txt and exp2\_2.txt, my solver does not terminate properly when running the experiment on instance exp2\_3.txt. In fact, it reports that a solution is found when a collision between two agents occurs. When running the experiment, Agent 1 just waits on (1,3) as shown in Figure 7.

To fix this issue, I modified the code written (for Section 2.3) to handle the constraints to have an upper-bound. This upper-bound is strictly as follows:

```
if maximum time step < counter: return 'No Solution'
```

where counter is defined as the number of constraints created for other agents once a higher priority agent is at its goal location.

The upper-bound prevents the situation where a lower priority agent i will wait out the constraints where a higher priority agent k is at its goal location, and the only path for i to its goal location is through k's goal location. Note however, that the implementation is simplified. The upper-bound condition is set to return no solution despite a non-collision-free path being found.

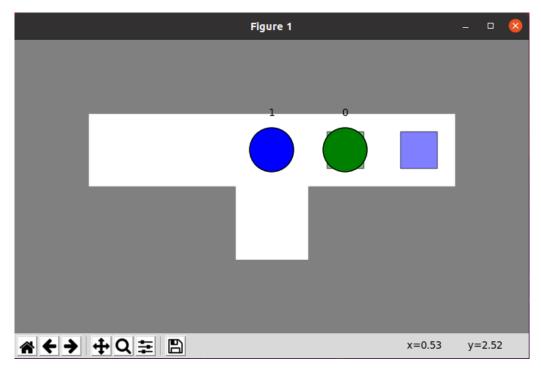


Figure 7: Agent 1 waits on (1,3).

#### 2.2 Incompleteness and Suboptimality

This section discusses instances that fulfills the requirements as mentioned in Section 2.5 of the lab.

File exp2\_4.txt is an instance for which prioritized planning does not find an optimal or non-optimal collision-free solution for a given ordering of agents. Without the upper-bound condition mentioned in Section 2.1 of the report, Agent 1 will simply wait out goal constraints because Agent 0 has a higher priority than Agent 1. This is seen in Figure 8a. In Figure 8b, we see the collision between the two agents when Agent 1 attempts to reach its goal location.

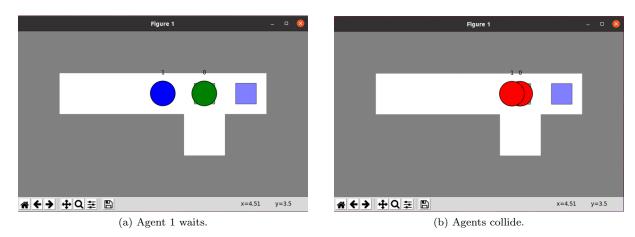


Figure 8: Animations for instance exp2\_4.txt.

Building on-top of exp2\_4.txt, file exp2\_5.txt is an instance where the ordering of the agents have been reversed. In this situation, both agents reach their goal as seen in Figure 9. This shows that there is an ordering of agents in which a collision-free solution is found despite having an ordering of agents in which no collision-free solution (based on exp2\_4.txt) is found.

In conclusion, both instances provide the same environment; however, exp2\_4.txt shows that there is no collision-free solution given some ordering. This satisfies the first requirement. exp2\_5.txt reverses the priorities between the agents presented in exp2\_4.txt, showing that there is a solution for a different ordering. This

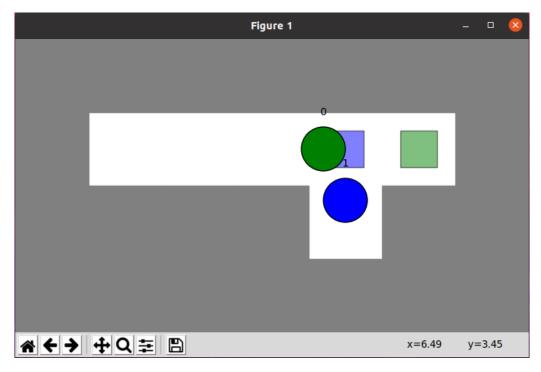


Figure 9: A solution is found in exp2\_5.txt.

satisfies the bonus requirement. Both of these instances prove that Prioritized Planning is incomplete and suboptimal.

## 3 Conflict-Based Search

## 3.1 Implementing High Level Search

These are the results after modifying cbs.py to detect collisions, convert collisions to constraints, and to conduct a high-level search. This section pertains to Section 3.3 of the lab. Note that collision detection is implemented according to the instructions in Section 3.1, and the conversion of collisions to constraints is implemented according to the instructions in Section 3.2.

The minimum sum of costs that I receive (based on results.csv) compared to min-sum-of-cost.csv shows that my Conflict-Based search algorithm works as intended. The costs match for every instance.

Figure 10 is the (truncated) transcript of running cbs.py on exp2\_1.txt with print statements on the nodes expanded.

#### 3.2 Custom Instances

This section discusses the custom instances made for cbs.py, pertaining to Section 3 of the lab.

File exp3\_1.txt is made to assess whether the correct information (for both Section 3.1 and Section 3.2) is returned when an edge collision occurs between two agents.

```
***Import an instance***

**Import an instance***

**Start locations

0 0 0 0 1 . . . 0

0 0 0 0 0 0 0

0 0 0 0 0 0

0 0 0 0 0

0 0 0 0 0

0 0 0 0 0

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```

Figure 10: Transcript of running Conflict-Based Search on exp2\_1.txt.

## 4 Conflict-Based Search with Disjoint Splitting

## 4.1 Adjusting the High-Level Search

These are the results after modifying cbs.py to perform disjoint splitting, pertaining to Section 4.3 of the lab.

Based on my results, normal splitting expands 15 nodes while disjoint splitting consistently expands 9 nodes on exp4.txt (which nearly matches the results indicated in the lab). The run-time and number of expanded nodes on the test\_\* instances have also decreased (by observation).

## 5 Additional Notes

I have added libraries in cbs.py, namely: import copy. This was used to deep copy some objects.