# Sample Solutions

Wolfgang Hönig\*, Jiaoyang Li, Sven Koenig — University of Southern California Model AI Assignments 2020: Project on Multi-Agent Path Finding (MAPF)

# 1 Task 1: Implementing Space-Time A\*

### 1.1 Searching in the Space-Time Domain

Changed lines (with respect to the student-provided code) are highlighted below:

```
root = {'loc': start_loc, 'g_val': 0, 'h_val': h_value, 'parent': None, 'timestep':
→ 0}
push_node(open_list, root)
closed_list[(root['loc'], root['timestep'])] = root
while len(open_list) > 0:
    curr = pop_node(open_list)
    if curr['loc'] == goal_loc:
        return get_path(curr)
    for dir in range(5):
        child_loc = move(curr['loc'], dir)
        if my_map[child_loc[0]][child_loc[1]]:
            continue
        child = {'loc': child_loc,
                'g_val': curr['g_val'] + 1,
                'h_val': h_values[child_loc],
                'parent': curr,
                'timestep': curr['timestep'] + 1}
        if (child['loc'], child['timestep']) in closed_list:
            existing_node = closed_list[(child['loc'], child['timestep'])]
            if compare_nodes(child, existing_node):
                closed_list[(child['loc'], child['timestep'])] = child
                push_node(open_list, child)
        else:
            closed_list[(child['loc'], child['timestep'])] = child
            push_node(open_list, child)
def move(loc, dir):
    directions = [(0, -1), (1, 0), (0, 1), (-1, 0), (0, 0)]
    return loc[0] + directions[dir][0], loc[1] + directions[dir][1]
```

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#### 1.2 Handling Vertex Constraints and Adding Edge Constraints

```
In a_star:
constraint_table = build_constraint_table(constraints, agent)
# ...
        if my_map[child_loc[0]][child_loc[1]] or is_constrained(curr['loc'],

    child_loc, curr['timestep'] + 1,
                                                                constraint_table):
            continue
Helper functions:
def build_constraint_table(constraints, agent):
    max_timestep = -1 # the maximum timestep in these constraints
    for constraint in constraints:
        if constraint['agent'] == agent:
            max_timestep = max(max_timestep, constraint['timestep'])
    constraint_table = [[] for _ in range(max_timestep + 1)]
    for constraint in constraints:
        if constraint['agent'] == agent:
            constraint_table[constraint['timestep']].append({'loc':

    constraint['loc']})
    return constraint_table
def is_constrained(curr_loc, next_loc, next_time, constraint_table):
    if len(constraint_table) <= next_time:</pre>
        return False
    for constraint in constraint_table[next_time]:
        if constraint['loc'] == [next_loc] or constraint['loc'] == [curr_loc,
        → next_loc]:
                return True
    return False
```

#### 1.3 Handling Goal Constraints

The new constraint is violated without an additional goal test condition (that is, agent 0 will be at the goal location at timestep 10). One possible goal test condition is the following:

```
if curr['loc'] == goal_loc:
    found = True
    if curr['timestep'] + 1 < len(constraint_table):
        for t in range(curr['timestep'] + 1, len(constraint_table)):
            if is_constrained(goal_loc, goal_loc, t, constraint_table):
            found = False</pre>
```

```
break
if found:
    return get_path(curr)
```

### 1.4 Optional: Designing Constraints

• One possible set of constraints is the following:

- Output paths: [[(1, 1), (1, 2), (1, 3), (1, 4), (1, 5)], [(1, 2), (1, 3), (2, 3), (1, 3), (1, 4)]]
- Sum of costs: 8

# 2 Task 2: Implementing Prioritized Planning

#### 2.1 Adding Vertex Constraints

(See solution below.)

### 2.2 Adding Edge Constraints

#### 2.3 Optional: Adding Additional Constraints

Here, we use a a global constraint set and after finding a path for an agent, we add constraints extracted from that path to all future agents. The following is executed after calling a\_star:

upper\_bound is a very large number or (better) computed as in the next task assignment.

#### 2.4 Optional: Addressing Failures

A good upper bound for the path length is  $height \cdot width \cdot longest$  possible path length:

#### 2.5 Optional: Showing that Prioritized Planning is Incomplete and Suboptimal

• Design a MAPF instance for which prioritized planning does not find an (optimal or suboptimal) collision-free solution for a given ordering of the agents.

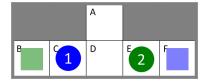
**Example**: Agent 1 moves from A to C, and agent 2 moves from B to D.



If agent 1 has higher priority than agent 2, then prioritized planning does not find an (optimal or suboptimal) collision-free solution because it first finds path [A, C] for agent 1 and then fails to find any collision-free path for agent 2.

• Design a MAPF instance for which prioritized planning does not find an (optimal or suboptimal) collision-free solution, no matter which ordering of the agents it uses.

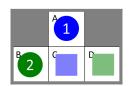
**Example:** Agent 1 moves from C to F, and agent 2 moves from E to B.



If agent 1 has higher priority than agent 2, then prioritized planning first finds path [C, D, E, F] for agent 1 and then fails to find any collision-free path for agent 2. If agent 2 has higher priority than agent 1, then prioritized planning first finds path [E, D, C, B] for agent 2 and then fails to find any collision-free path for agent 1.

• Design a MAPF instance for which prioritized planning does not find an (optimal or suboptimal) collision-free solution for a given ordering of the agents even if an ordering of the agents exists for which prioritized planning finds an optimal collision-free solution.

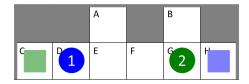
**Example**: Agent 1 moves from A to C, and agent 2 moves from B to D.



If agent 1 has higher priority than agent 2, then prioritized planning does not find an (optimal or suboptimal) collision-free solution because it first finds path [A, C] for agent 1 and then fails to find any collision-free path for agent 2. However, if agent 2 has higher priority than agent 1, then prioritized planning finds the optimal collision-free solution (of cost 4), which consists of path [A, A, C] (of length 2) for agent 1 and path [B, C, D] (of length 2) for agent 2.

• Design a MAPF instance for which prioritized planning finds a suboptimal (but not optimal) collision-free solution for a given ordering of the agents even if an ordering of the agents exists for which prioritized planning finds an optimal collision-free solution.

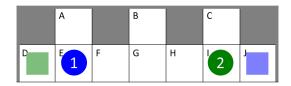
**Example**: Agent 1 moves from D to H, and agent 2 moves from G to C.



If agent 1 has higher priority than agent 2, then prioritized planning finds a suboptimal collision-free solution (of cost 12) that consists of path [D, E, F, G, H] (of length 4) for agent 1 and path [G, B, B, B, G, F, E, D, C] (of length 8) for agent 2. However, if agent 2 has higher priority than agent 1, then prioritized planning is able to find the optimal collision-free solution (of cost 10), which consists of path [D, E, A, E, F, G, H] (of length 6) for agent 1 and path [G, F, E, D, C] (of length 4) for agent 2.

• Design a MAPF instance for which prioritized planning does not find an optimal collision-free solution, no matter which ordering of the agents it uses, even if a collision-free solution exists.

**Example**: Agent 1 moves from E to J, and agent 2 moves from I to D.



If agent 1 has higher priority than agent 2, then prioritized planning finds a suboptimal collision-free solution (of cost 15) that consists of path [E, F, G, H, I, J] (of length 5) for agent 1 and path [I, C, C, C, C, I, H, G, F, E, D] (of length 10) for agent 2. If agent 2 has higher priority than agent 1, then prioritized planning finds a suboptimal collision-free solution (of cost 15) that consists of path [E, A, A, A, E, F, G, H, I, J] (of length 10) for agent 1 and path [I, H, G, F, E, D] (of length 5) for agent 2. However, the optimal collision-free solution is of cost 13 and consists of path [E, F, G, B, G, H, I, J] (of length 7) for agent 1 and path [I, H, G, F, E, D] (of length 6) for agent 2.

# 3 Task 3: Implementing Conflict-Based Search (CBS)

### 3.1 Detecting Collisions

```
def detect_collision(path1, path2):
    for t in range(max(len(path1), len(path2))):
        # extract the current coordinates along the path
        c1 = get_location(path1, t)
        c2 = get_location(path2, t)
        if c1 == c2:
            return [c1, t]
        else:
            # extract the previous coordinates along the path
            # (for edge checking)
            p1 = get_location(path1, t - 1)
            p2 = get_location(path2, t - 1)
            if c1 == p2 and c2 == p1:
                return [p1, c1, t]
    return None
def detect_collisions(paths):
    collisions = []
    for i in range(len(paths) - 1):
        for j in range(i + 1, len(paths)):
            collision = detect_collision(paths[i], paths[j])
            if collision is not None:
                collisions.append({'a1': i,
                                   'a2': j,
                                   'loc': collision[:-1],
                                   'timestep': collision[-1]})
    return collisions
```

#### 3.2 Converting Collisions to Constraints

#### 3.3 Implementing the High-Level Search

```
# Search the tree until the open list is empty
while len(self.open_list) > 0:
    node = self.pop_node() # pop the node with the minimum cost
    if node['collisions'] == []: # the paths of this node is collision-free
        self.print_results(node)
        return node['paths']
    collision = node['collisions'][0] # choose a collision
    print("Choose a collision between {} and {} at location {} at timestep
    → {}".format(
        collision['a1'], collision['a2'], collision['loc'], collision['timestep']))
    # Generate constraints to resolve the chosen collision
    new_constraints = standard_splitting(collision)
    for constraint in new_constraints: # for each constraint, generate a child
    \rightarrow node
        i = constraint['agent']
        print("Negative constraint on agent {} at location {} at timestep
        → {}".format(
                constraint['agent'], constraint['loc'], constraint['timestep']))
        constraints = list(node['constraints']) # copy constraints
        constraints.append(constraint) # add additional constraint
        paths = list(node['paths']) # copy paths
        paths[i] = a_star(self.my_map, self.starts[i], self.goals[i],

    self.heuristics[i],

                          i, constraints) # replan for this agent
        # Generate a child node
        if paths[i] is not None:
            child_node = {'cost': get_sum_of_cost(paths),
                          'constraints':constraints,
                          'paths': paths,
                          'collisions': detect_collisions(paths)}
            self.push_node(child_node)
raise BaseException('No solutions')
```

# 4 Optional Task 4: Implementing CBS with Disjoint Splitting

### 4.1 Supporting Positive Constraints

```
def build_constraint_table(constraints, agent):
    positive = [] # to collect positive constraints
    negative = [] # to collect negative constraints
```

```
max_timestep = -1  # the maximum timestep in these constraints
   # collect constraints that are related to this agent
   for constraint in constraints:
       if constraint['positive']: # positive constraint is effective for everyone
           if constraint['agent'] == agent:
              positive.append(constraint)
           else:
              negative.append(constraint)
           max_timestep = max(max_timestep, constraint['timestep'])
       elif constraint['agent'] == agent: # negative constraint is effective for
       → only one agent
           negative.append(constraint)
           max_timestep = max(max_timestep, constraint['timestep'])
   constraint_table = [[] for _ in range(max_timestep + 1)]
   for constraint in positive:
       if len(constraint['loc']) == 1: # positive vertex constraint
           constraint_table[constraint['timestep']].append({'loc':

    constraint['loc'], 'positive': True})

       else: # positive edge constraint
           constraint_table[constraint['timestep'] - 1].append({'loc':
           constraint_table[constraint['timestep']].append({'loc':
           for constraint in negative:
       if len(constraint['loc']) == 1: # vertex constraint
           constraint_table[constraint['timestep']].append({'loc':

    constraint['loc'], 'positive': False})

       elif constraint['positive']: # positive edge constraint for other agents
           constraint_table[constraint['timestep'] - 1].append({'loc':
           constraint_table[constraint['timestep']].append({'loc':
           constraint_table[constraint['timestep']].append(
              {'loc': [constraint['loc'][1], constraint['loc'][0]], 'positive':
               → False})
       else: # negative edge constraint
           constraint_table[constraint['timestep']].append({'loc':

    constraint['loc'], 'positive': False})
   return constraint_table
def is_constrained(curr_loc, next_loc, next_time, constraint_table):
   if len(constraint_table) <= next_time:</pre>
       return False
   for constraint in constraint_table[next_time]:
       if constraint['positive']: # positive constraint
```

```
i = random.randint(0, 1)
if i == 0:
                   'loc': collision['loc'],
                   'timestep': collision['timestep'],
                   'positive': True}
    constraint2 = {'agent': collision['a1'],
                   'loc': collision['loc'],
                   'timestep': collision['timestep'],
                   'positive': False}
else:
    loc = list(collision['loc'])
    loc.reverse() # if this is an edge collision, we need to use the other
    → direction of the edge in the constraint
    constraint1 = {'agent': collision['a2'],
                   'loc': loc,
                   'timestep': collision['timestep'],
                   'positive': True}
    constraint2 = {'agent': collision['a2'],
                   'loc': loc.
```

'timestep': collision['timestep'],

'positive': False}

#### 4.3 Adjusting the High-Level Search

return [constraint1, constraint2]

```
# Generate constraints to resolve the chosen collision
    new_constraints = disjoint_splitting(collision)
    for constraint in new_constraints: # for each constraint, generate a child
        print("Constraint on agent {} at location {} at timestep {}".format(
            constraint['agent'], constraint['loc'], constraint['timestep']))
        constraints = list(node['constraints']) # copy constraints
        constraints.append(constraint) # add additional constraint
        paths = list(node['paths']) # copy paths
        # Find agents that need to be re-planned
        replan = []
        if constraint['positive']: # positive constraint
            # Any paths that violate this positive constraint needs to be
            \rightarrow re-planned
            replan = paths_violate_constraint(constraint, paths)
        else: # negative constraint
            # Only the constrained path needs to be re-planned
            replan.append(constraint['agent'])
        prune = False
        for i in replan:
            paths[i] = a_star(self.my_map, self.starts[i], self.goals[i],

    self.heuristics[i],

                              i, constraints)
            if paths[i] is None:
                prune = True
                break
        if not prune:
            # Generate a child node
            child_node = {'cost': get_sum_of_cost(paths),
                          'constraints': constraints,
                          'paths': paths,
                          'collisions': detect_collisions(paths)}
            self.push_node(child_node)
raise BaseException('No solutions')
```