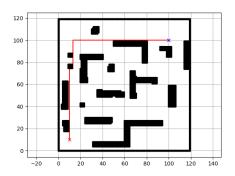
By: WuShuwen (521030910087)

HW#: 1

October 9, 2023

I. TASK 1: BASIC A* ALGORITHM



(a) Output of Basic A* Algorithm

FIG. 1

A. Implementation

I implemented the basic A^* algorithm by defining a class of nodes, named N, and a class of the algorithm itself, named A. Evaluation function, calculated by adding cost function and heuristic function, is included in the class of nodes.

Complete code:

```
import sys
        import os
import numpy as np
import matplotlib.pyplot as plt
        MAP_PATH = os.path.join(os.path.dirname(os.path.abspath(__file__)), '3-map/map.npy')
       ### START CODE HERE ###
# This code block is optional. You can define your utility function and class in this block if necessary.
class N():
    def __init__(self ,x ,y ,parent=None) -> None:
        self .x =x
        self .y =y
        self .parent = parent
        if [x,y] == start.pos:
            self .parent = None
        self .parent = None
        self .g = 0
        else:
11
13
15
17
                              else:
    self.g = self.parent.g + abs(self.x - self.parent.x) + abs(self.y - self.parent.y)
self.h = abs(self.x - goal-pos[0]) + abs(self.y - goal-pos[1])
self.f = self.h+self.g
19
21
23
                             :
__init__(self,map,start,goal) -> None:
self.map =map
self.start = start
self.goal = goal
self.openset = []
self.closeset = []
self.currentN = N(self.start[0],self.start[1])
self.openset.append(self.currentN)
25
27
29
31
                  def in.close(self,x,y):
    for i in self.closeset:
        if x == i.x and y==i.y:
        return True
    return False
33
35
37
                  def in_open(self,x,y):
    for i in self.openset:
        if x == i.x and y==i.y:
        return True
    return False
39
41
43
                             \begin{array}{lll} iterate\,(\,self\,,x\,,y\,,parent\,):\\ if\,\,(x\,<\,0)\,\,or\,\,(x\,>=\,120)\,\,or\,\,(y\,<\,0)\,\,or\,\,(y\,>=\,120):\\ &\,\,return\\ if\,\,self\,.map\,[\,x\,]\,[\,y\,] \ == \ 1: \end{array}
45
47
```

```
49
                                   return
                          return

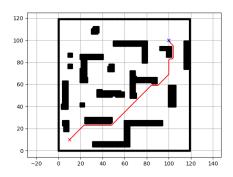
if self.in_close(x,y):
    return

if not self.in_open(x,y):
    n = N(x,y,parent)
    self.openset.append(n)
 51
 53
  55
                 def sel_min(self)
                          min_cost = 1000000
num =0
for n in self.openset:
    if n.f <=min_cost:
 57
 59
                                   min_cost = n.f
sel = num
num +=1
 61
 63
                          return sel
 65
                 def res(self,n):
    path = []
    while True:
 67
                                    n=n.parent
if n:
 69
                                    path.insert(0,[n.x,n.y])
if (n.x == self.currentN.x) and (n.y == self.currentN.y):
return path
 71
 73
 75
                 def algorithm (self):
                          algorithm(self):
while True:
    i = self.sel_min()
    n = self.openset[i]
    self.closeset.append(n)
    del self.openset[i]
    if(n.x == self.goal[0] and n.y == self.goal[1]):
        path = self.res(n)
 77
 79
  81
 83
                                    \begin{array}{c} \textbf{return} & \text{path} \\ \text{self.iterate} \left( \left( n.x-1, n.y, n \right) \\ \text{self.iterate} \left( n.x, n.y-1, n \right) \\ \text{self.iterate} \left( n.x+1, n.y, n \right) \\ \text{self.iterate} \left( n.x+1, n.y, n \right) \\ \text{self.iterate} \left( n.x, n.y+1, n \right) \end{array}
  85
 87
 89
 91
        ### END CODE HERE ###
 93
 95
        def A_star(world_map, start_pos, goal_pos):
        ____Given_map_of_the_world,_start_position_of_the_robot_and_the_position_of_the_goal,___plan_a_path_from_start_position_to_the_goal_using_A*_algorithm.
 99
        ___arguments:
___A_120*120_array_indicating_map,_where_0_indicating_traversable_and_1_indicating_obstacles.
___start_pos___A_2D_vector_indicating_the_start_position_of_the_robot.
___soal_pos___A_2D_vector_indicating_the_position_of_the_goal.
101
103
         ____path____A_N*2_array_representing_the_planned_path_by_A*_algorithm .
105
107
                 ### START CODE HERE ###
a = A(world_map,start_pos,goal_pos)
path = a.algorithm()
### END CODE HERE ###
109
111
                 return path
113
115
117
119
        i f _{-n} n a m e_{-} = '_{-m} a i n_{-}':
                 \# Get the map of the world representing in a 120*120 array, where 0 indicating traversable and 1 indicating
121
                             obstacl
                 map = np.load (MAP_PATH)
123
                 \# Define goal position of the exploration goal-pos = [100\,,\ 100]
125
                 \# Define start position of the robot. start_pos = [10\,,\ 10]
127
129
                 \# Plan a path based on map from start position of the robot to the goal. path = A_star(map, start_pos, goal_pos)
131
                   \# \  \, \text{Visualize the map and path.} \\ \text{obstacles\_x.} \quad \text{obstacles\_y} = [] \;, \; [] \\ \text{for } i \; \text{in range}(120) \colon \\ \text{for } j \; \text{in range}(120) \colon \\ \text{if } \max[i][j] == 1 \colon \\ \text{obstacles\_x.append(i)} \\ \text{obstacles\_y.append(j)} 
133
135
137
139
                 path.x , path.y = [] , []
for path.node in path:
   path.x .append(path.node[0])
   path.y .append(path.node[1])
141
143
145
                 plt.plot(path_x, path_y, "-r")
plt.plot(start.pos[0], start.pos[1], "xr")
plt.plot(goal-pos[0], goal-pos[1], "xb")
plt.plot(obstacles_x, obstacles_y, ".k")
147
149
```

plt.grid(True) plt.axis("equal") plt.show()

151

II. TASK 2: IMPROVED A* ALGORITHM



(a) Output of improved A^* algorithm when lower limit for the distance to obstacle set to 1



(b) Output of improved A^* algorithm when lower limit for the distance to obstacle set to 3

FIG. 2

A. Implementation

The implementation of improved A* algorithm is based on that of basic A* algorithm. Compared to the basic one, the path is enabled to upper left, upper right, bottom left, bottom right. The cost of doing so is set to be $\sqrt{2}$.

To avoid collision, I simply set a lower limit for the distance to obstacle, by creating a new map where the blocks considered dangerous for being too close to obstacles are also abandoned.

```
def gen_obstacle_map(self):
    obstacle_map = self.map.copy()
    for i in range(1, 119):
        for j in range(1, 119):
            if self.map[i][j] + self.map[i - 1][j] + \
```

To avoid steering, I added a steering cost to the evaluation, so that turns of acute angles would be punished. I implemented this with a second time difference of the path, that is, by assessing the difference of the direction from the parent node to the current node and that from the current node to the next node.

```
self.x_change = self.x - self.parent.x
self.y_change = self.y - self.parent.y

if [self.parent.x, self.parent.y] != start_pos and abs(self.x_change-self.parent.x_change)+abs(self.y_change-self.parent.y_change)>1:
self.s=1
```

Complete code:

```
import sys
    import numpy as np
    import matplotlib.pyplot as plt
    MAP_PATH = os.path.join(os.path.dirname(os.path.abspath(__file__)), '3-map/map.npy')
    ### START CODE HERE ###
# This code block is optional. You can define your utility function and class in this block if necessary.
import math
    class N():
    def __init__(self,x,y,parent=None) -> None:
        self.x =x
        self.y =y
11
13
15
                 self.parent = parent
                self.s = 0
if [x,y] == start_pos:
    self.parent = None
    self.g = 0
17
19
21
                       :
self.x-change = self.x - self.parent.x
self.y-change = self.y - self.parent.y
if abs(self.x-change) + abs(self.y-change) == 1:
self.g = self.parent.g + 1
23
25
                       self.g = self.parent.g +math.sqrt(2)
if [self.parent.x,self.parent.y] != start_pos and abs(self.x_change-self.parent.x_change)+abs(self.
    y_change-self.parent.y_change)>1:
27
29
                                    self.s=1
                 self.h = abs(self.x - goal-pos[0]) + abs(self.y - goal-pos[1])
31
                 self.f = self.g + self.h + self.s
33
                -_init__(self,map,start,goal) -> None: self.map =map
35
                37
39
41
43
45
47
49
51
53
55
          def in_close(self,x,y):
    for i in self.closeset:
        if x == i.x and y==i.y:
            return True
57
59
                return False
61
          def in_open(self,x,y):
    for i in self.openset:
        if x == i.x and y=
        return True
    return False
63
65
67
          def iterate(self,x,y,parent):
```

```
\label{eq:control_eq} \mbox{if } (x \ < \ 0) \ \mbox{or} \ (x \ > = \ 120) \ \mbox{or} \ (y \ < \ 0) \ \mbox{or} \ (y \ > = \ 120):
 69
                       return
if self.map[x][y] == 1:
 71
                               return
                       if self.in_close(x,y):
 73
                       return
if not self.in_open(x,y):
n = N(x,y,parent)
 75
 77
                               self.openset.append(n)
 79
               def sel_min(self)
                       sel_min(seir):
min_cost = 1000000
num =0
for n in self.openset:
    if n.f <=min_cost:
        min_cost = n.f
        sel = num
    num +=1
 81
 83
 85
 87
                       return sel
 89
               def res(self,n):
 91
                       path = []
while True:
 93
 95
                                        \begin{array}{lll} \text{path.insert} \; (0 \; , [n \cdot x \; , n \cdot y \; ]) \\ (n \cdot x \; = \; \text{self.currentN} \; .x) \; \; \text{and} \; \; (n \cdot y \; = \; \text{self.currentN} \; .y) \; : \end{array} 
 97
                                       return path
               def algorithm (self):
 99
                       algorithm(self):
while True:
    i = self.sel_min()
    n = self.openset[i]
    self.closeset.append(n)
    del self.openset[i]
    if(n.x == self.goal[0] and n.y == self.goal[1]):
        path = self.res(n)
101
103
105
107
                               return path self.iterate (n, x+1, n, y+1, n) self.iterate (n, x-1, n, y-1, n) self.iterate (n, x-1, n, y-1, n) self.iterate (n, x, n, y, y, n)
109
111
113
115
                               self.iterate (n.x+1,n.y,n]
self.iterate (n.x,n.y+1,n]
117
       ### END CODE HERE ###
119
121
       def Improved_A_star(world_map, start_pos, goal_pos):
       ____Given_map_of_the_world,_start_position_of_the_robot_and_the_position_of_the_goal,___plan_a_path_from_start_position_to_the_goal_using_improved_A*_algorithm.
123
125
        ___Arguments:
127
       ___A_120*120 array_indicating_map,_where_O_indicating_traversable_and_1_indicating_obstacles.
___A_2D_vector_indicating_the_start_position_of_the_robot.
___Soal_pos___A_2D_vector_indicating_the_position_of_the_goal.
129
131
        ____path_____A_N*2_array_representing_the_planned_path_by_improved_A*_algorithm.
133
135
               ### START CODE HERE ###
137
               a = A(world_map, start_pos, goal_pos)
path = a.algorithm()
139
               ### END CODE HERE ###
141
               return path
143
145
147
       if __name__ == '__main__':
149
               \# Get the map of the world representing in a 120*120 array, where 0 indicating traversable and 1 indicating
               map \ = \ np \cdot load \, (MAP\_PATH)
151
153
               \# Define goal position of the exploration goal_pos = [100\,,\ 100]
155
               \# Define start position of the robot. start-pos = [10, 10]
157
               \# Plan a path based on map from start position of the robot to the goal. path = Improved_A_star(map, start_pos, goal_pos)
159
161
               # Visualize the map and path
               # Visualize the map and path.
obstacles_x, obstacles_y = [], []
for i in range(120):
    for j in range(120):
        if map[i][j] == 1:
            obstacles_x.append(i)
            obstacles_y.append(j)
163
165
167
169
```

```
path_x, path_y = [], []

for path_node in path:
    path_x.append(path_node[0])

path_y.append(path_node[1])

175

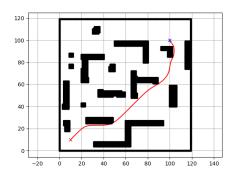
plt.plot(path_x, path_y, "-r")
    plt.plot(start.pos[0], start.pos[1], "xr")
    plt.plot(goal.pos[0], goal.pos[1], "xb")
    plt.plot(obstacles_x, obstacles_y, ".k")
    plt.grid(True)
    plt.axis("equal")

plt.show()
```

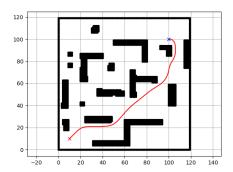
B. Comparison

Compared to the basic one, the improved A^* algorithm keeps a sufficient distance from obstacles, thus guarantees safety. In respect of optimality, the improved algorithm does better in this map, though it is restricted by the anti-collision limits, because it can go upper left, upper right, bottom left, bottom right.

III. TASK 3: PATH PLANNING FOR SELF-DRIVING



(a) Output of improved A^* algorithm with Bezier curve when lower limit of distance to obstacle set to 1



(b) Output of improved A^* algorithm with Bezier curve when lower limit of distance to obstacle set to 3

FIG. 3

A. Implementation

To smooth the path, I simply use the path produced by the improved A^* algorithm as control points of Bezier curve to form a smooth path.

I implemented Bezier curve according to its definition:

$$C_n(u) = \sum_{i=0}^n P_i b_{n,i}(u)$$
 (1)

Where, $b_{n,i}$ represents the Bernstein polynomial:

$$b_{n,i}(u) = C_n^i t^i (1-t)^{(n-i)}$$
(2)

```
def bernstein-poly(self,i, n, t):
    return np.math.comb(n, i) * (1 - t)**(n - i) * t**i

def bezier_curve(self,control-points):
    num.points = 100
    t = np.linspace(0, 1, num.points)
    n = len(control.points) - 1
    curve-points = []
    for i in range(num.points):
        x = sum(self.bernstein-poly(j, n, t[i]) * control-points[j][0] for j in range(n + 1))
        y = sum(self.bernstein-poly(j, n, t[i]) * control-points[j][1] for j in range(n + 1))
        curve-points.append((x, y))

return curve-points
```

As is shown in the pictures, the smoothed path whose lower limit is set to 1 turns out to be too close to obstacles, while the one with lower limit set to 3 still performs well.

Complete code:

```
import sys
import os
import numpy as np
import matplotlib.pyplot as plt
   MAP_PATH = os.path.join(os.path.dirname(os.path.abspath(__file__)), '3-map/map.npy')
   10
12
              _-init_-(self,x,y,parer
self.x =x
self.y =y
self.parent = parent
self.s = 0
if [x,y] == start_pos:
    self.parent = None
    self.g = 0
14
16
18
20
                    :
self.x_change = self.x - self.parent.x
self.y_change = self.y - self.parent.y
if abs(self.x_change) + abs(self.y_change) == 1:
self.g = self.parent.g + 1
22
24
26
                    28
30
              self.h = abs(self.x - goal_pos[0]) + abs(self.y - goal_pos[1])
32
              self.f = self.g + self.h + self.s
34
   class A:
              :
__init__(self,map,start,goal) -> None:
self.map =map
36
38
40
              self.gen_obstacle_map
              self.gen_obstacle_map()
42
              self.start = start
self.goal = goal
self.openset = []
self.closeset = []
self.currentN = N(self.start[0], self.start[1])
self.openset.append(self.currentN)
44
46
48
         def gen_obstacle_map(self):
50
               52
54
56
58
60
              self.map = obstacle\_map
62
         def in_close(self,x,y):
    for i in self.closeset:
        if x == i.x and y==i.y:
            return True
64
66
              return False
68
        def in_open(self,x,y):
    for i in self.openset:
        if x == i.x and y==i.y:
        return True
70
72
```

```
return False
 74
                  \begin{array}{lll} \text{def iterate} \left( \, \text{self} \, , x \, , y \, , \, \text{parent} \, \right) : \\ & \text{if} \quad (x \, < \, 0) \quad \text{or} \quad (x \, > = \, 120) \quad \text{or} \quad (y \, < \, 0) \quad \text{or} \quad (y \, > = \, 120) : \\ & \quad \quad \text{return} \end{array}
  76
                            if \ self .map[x][y] == 1:
  78
                                     return
                            if self.in_close(x,y):
 80
                                     return
                            if not self.in_open(x,y):

n = N(x,y,parent)
 82
 84
                                     self.openset.append(n)
  86
                  def sel_min(self)
                           sel_min(self):
min_cost = 1000000
num =0
for n in self.openset:
    if n.f <=min_cost:</pre>
 88
 90
                                              min_cost = n.f
sel = num
+=1
 92
                            num +=
return sel
 94
 96
                  98
                 def bezier_curve(self,control_points):
    num_points = 100
    t = np.linspace(0, 1, num_points)
    n = len(control_points) - 1
    curve_points = []
    for i in range(num_points):
        x = sum(self.bernstein_poly(j, n, t[i]) * control_points[j][0] for j in range(n + 1))
        y = sum(self.bernstein_poly(j, n, t[i]) * control_points[j][1] for j in range(n + 1))
        curve_points.append((x, y))
100
102
104
106
108
110
                            return curve_points
                  def res(self,n):
    path = []
    while True:
112
114
                                     n=n \cdot p a r e n t

i f n :
116
                                     if n:
    path.insert(0,[n.x,n.y])
if (n.x == self.currentN.x) and (n.y == self.currentN.y):
    return self.bezier_curve(path)
    #return path
118
120
                 def algorithm(self):
    while True:
        i = self.sel.min()
        n = self.openset[i]
        self.closeset.append(n)
        del self.openset[i]
        if(n.x == self.goal[0] and n.y == self.goal[1]):
            path = self.res(n)
122
124
126
128
130
                                     \begin{array}{c} \textbf{return} & \textbf{path} \\ \textbf{self.iterate} \left( n. \, x\!+\!1, n. \, y\!+\!1, n \right) \\ \textbf{self.iterate} \left( n. \, x\!-\!1, n. \, y\!+\!1, n \right) \\ \textbf{self.iterate} \left( n. \, x\!-\!1, n. \, y\!-\!1, n \right) \\ \textbf{self.iterate} \left( n. \, x\!+\!1, n. \, y\!-\!1, n \right) \\ \textbf{self.iterate} \left( n. \, x\!-\!1, n. \, y\!-\!1, n \right) \\ \textbf{self.iterate} \left( n. \, x\!-\!1, n. \, y, n \right) \\ \textbf{self.iterate} \left( n. \, x, n. \, y\!-\!1, n \right) \\ \textbf{self.iterate} \left( n. \, x\!+\!1, n. \, y, n \right) \\ \textbf{self.iterate} \left( n. \, x\!+\!1, n. \, y, n \right) \\ \textbf{self.iterate} \left( n. \, x, n. \, y\!+\!1, n \right) \end{array}
132
134
136
138
140
        ### END CODE HERE ###
142
144
146
        def self_driving_path_planner(world_map, start_pos, goal_pos):
         ____Given_map_of_the_world,_start_position_of_the_robot_and_the_position_of_the_goal,
148
          ___plan_a_path_from_start_position_to_the_goal.
150
         ____A_120*120_array_indicating_map,_where_0_indicating_traversable_and_1_indicating_obstacles.___start_pos___A_2D_vector_indicating_the_start_position_of_the_robot.___A_2D_vector_indicating_the_position_of_the_goal.
152
154
         ___Return:
156
         ____path__
                                  -_A_N*2_array_representing_the_planned_path.
158
                  ### START CODE HERE ###
160
                  a = A(world_map, start_pos, goal_pos)
path = a.algorithm()
162
164
                 ### END CODE HERE ###
return path
166
168
170
172
        if __name__ == '__main__':
         \# Get the map of the world representing in a 120*120 array, where 0 indicating traversable and 1 indicating
```

```
\begin{array}{ll} & \text{obstacles} \; . \\ \text{map} \; = \; \text{np.load} \; (\text{MAP\_PATH}) \end{array}
176
                     178
                     \# Define start position of the robot. start_pos = [10\,,\ 10]
180
182
                     \# Plan a path based on map from start position of the robot to the goal. path = self_driving_path_planner(map, start_pos, goal_pos)
                    path = self-dfiveneer.
# Visualize the map and path.
obstacles.x, obstacles.y = [], []
for i in range(120):
    for j in range(120):
        if map[i][j] == 1:
            obstacles.x.append(i)
            obstacles.y.append(j)
186
188
190
192
                     path_x , path_y = [] , []
for path_node in path:
    path_x .append(path_node[0])
    path_y .append(path_node[1])
194
196
                     plt.plot(path_x, path_y, "-r")
plt.plot(start_pos[0], start_pos[1], "xr")
plt.plot(goal_pos[0], goal_pos[1], "xb")
plt.plot(obstacles_x, obstacles_y, ".k")
plt.grid(True)
plt.axis("equal")
plt.show()
198
200
202
204
```