



AAE2004 Introduction to Aviation Systems AAE

Design of Path Planning Algorithm for Aircraft Operation

Week 9 (Introduction to Path Planning)

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Necessary Information

- Course Repository (project download) link:
- https://github.com/IPNL-POLYU/PolyU AAE2004 Github Project

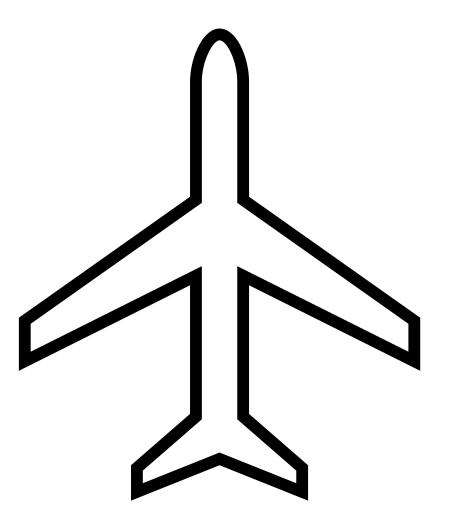
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Week 2 Content

- 1. Introduction to A* Path Planning Algorithm
- 2. Cost Intensive Areas
- 3. Path Planning Programming Guide
- 4. Project Compulsory Tasks





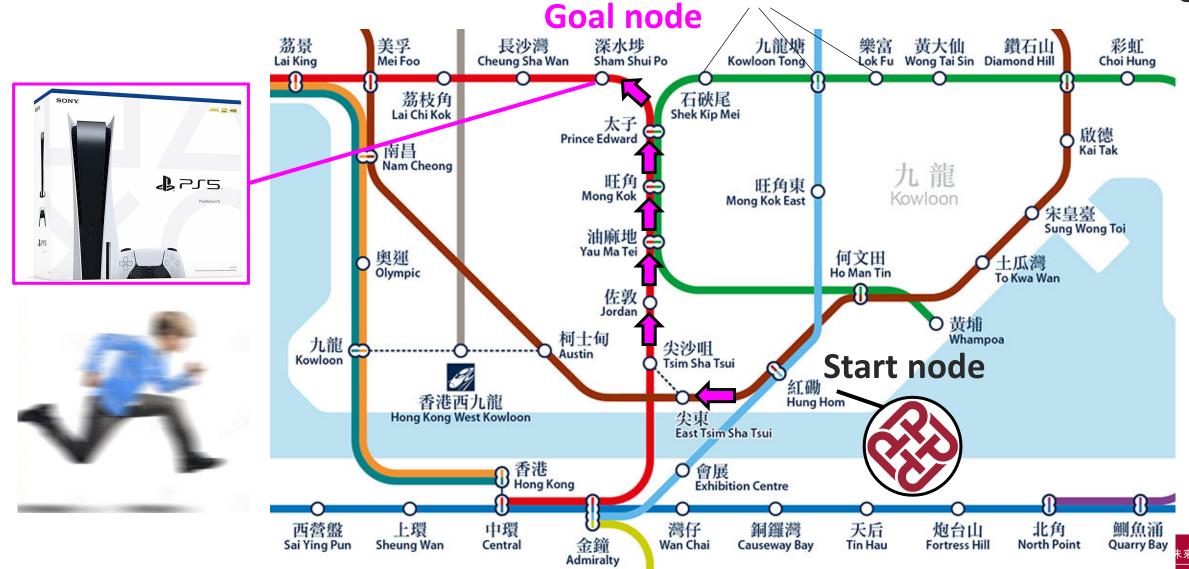
Introduction to A* Path Planning Algorithm





Daily Path Planning

MTR station – available node to go

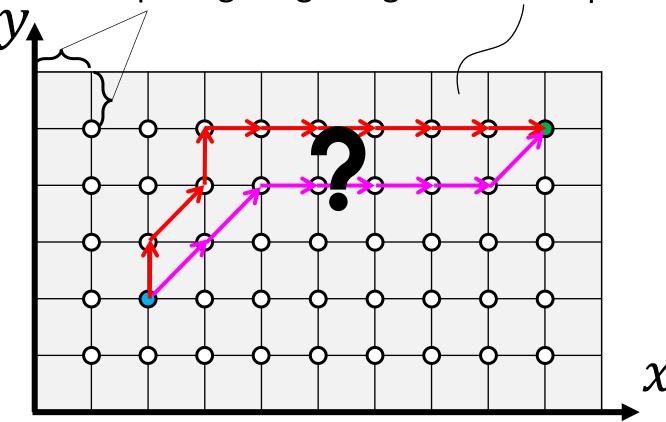






Definition of Path Planning

1 m spacing for griding 2D free space



(0,0) • Start node • Goal node

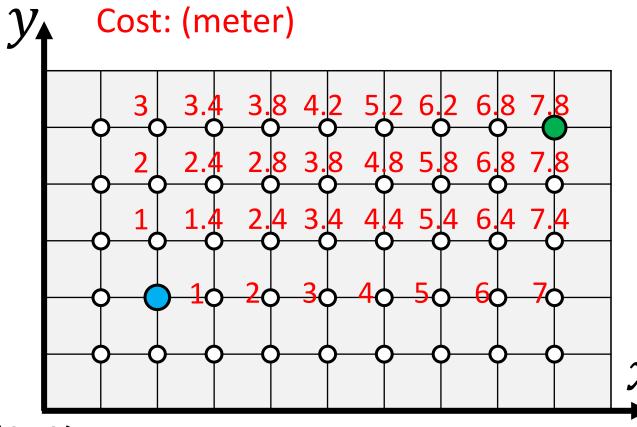
- •Node All potential position you can go across with a unique position (x, y)
- •Search space A collection of nodes, like all board positions of a board game.
- •Objective of path planning— Find the <u>shortest</u> routes with <u>smallest cost</u> from start node to goal node.

Which one is better?





Path Planning by Checking All Available Nodes



(0,0) • Start node • Goal node 1 m spacing for griding Test each possible nodes one-by-one from Start node



Record its shortest path between Start node

Reach Goal

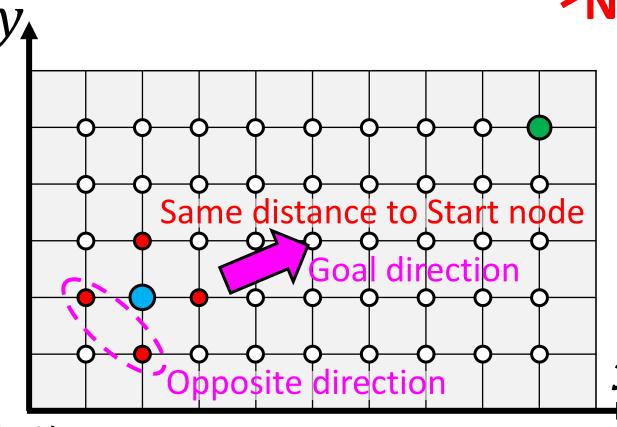
Retrieve the shortest path (Dijkstra's algorithm)

Need higher efficiency!





Path Planning by Checking All Available Nodes



(,0) • Start node • Goal node 1 m spacing for griding Test each possible nodes one-by-one from Start node



Record its shortest path

between Start node



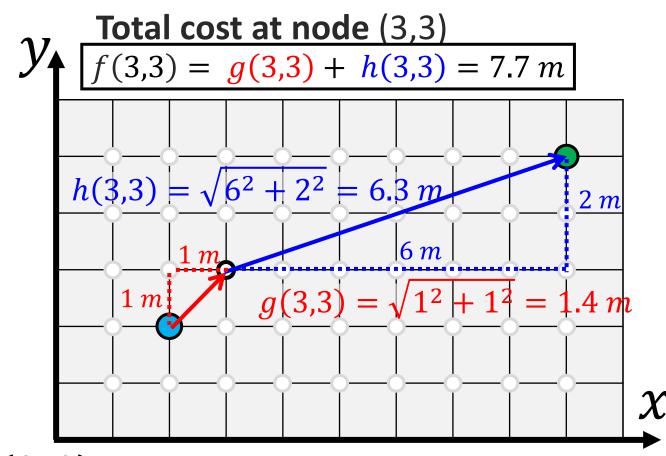
We also know a rough direction to Goal!







Path Planning by A-star (A*) Search Algorithm



- (0,0) \circ Start node
 - Goal node 1 m spacing for griding

Definition of cost:

g(x,y) — this represents the **exact cost** of the path **from** the **Start** node to node (x,y)

h(x, y) — this represents the heuristic <u>estimated cost</u> from node (x, y) to the Goal node

$$f(x,y) = g(x,y) + h(x,y)$$

 $- \underline{\text{total}} \text{ cost of a neighboring}$
 $\text{node } (x,y)$





A-star (A*) Path Planning – Cost at Node (3,2)

$$y = f(3,2) = g(3,2) + h(3,2) = 7.7 m$$

$$h(3,2) = \sqrt{6^2 + 3^2} = 6.7 m$$

$$g(3,2) = 1 m$$

(0,0) • Start node • Goal node 1 m spacing for griding

Definition of cost:

g(x, y) — this represents the **exact cost** of the path **from** the **Start** node to node (x, y)

h(x, y) — this represents the heuristic <u>estimated cost</u> from node (x, y) to the Goal node

$$f(x,y) = g(x,y) + h(x,y)$$

 $- \underline{\text{total}} \text{ cost of a neighboring}$
 $\text{node } (x,y)$





A-star (A*) Path Planning – Cost at Node (2,1)

$$y = f(2,1) = g(2,1) + h(2,1) = 8.2 m$$

$$h(2,1) = \sqrt{6^2 + 4^2} = 7.2 m$$

$$g(2,1) = 1 m$$

(0,0) • Start node • Goal node 1 m spacing for griding

Definition of cost:

g(x,y) — this represents the **exact cost** of the path **from** the **Start** node to node (x,y)

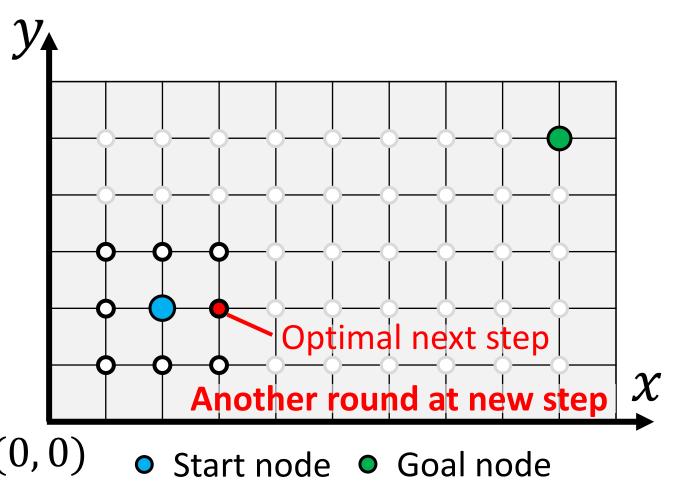
h(x, y) — this represents the heuristic <u>estimated cost</u> from node (x, y) to the Goal node

f(x,y) = g(x,y) + h(x,y) $- \underline{\text{total}} \text{ cost of a neighboring}$ node (x,y)





Total Costs at Neighbouring Nodes



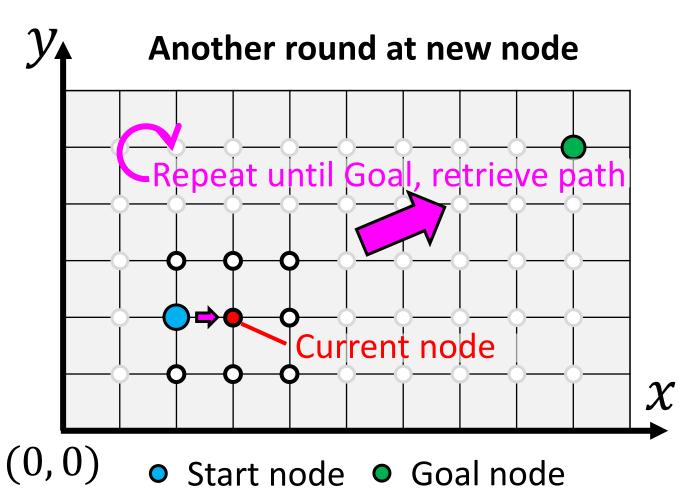
Exact cost Estimated Total cost from Start cost to Goal

Node ID	g(x,y)	h(x,y)	f(x,y)
(1,1)	1.4	8.9	10.3
(1,2)	1	8.5	9.5
(1,3)	1.4	8.2	9.6
(2,1)	1	8.1	9.1
(2,3)	1	7.3	8.3
(3,1)	1.4	7.2	8.6
(3,2)	1	6.7	7.7
(3,3)	1.4	6.3	7.7





Total Costs at Neighbouring Nodes – Round 2



Exact cost Estimated from Start cost to Goal

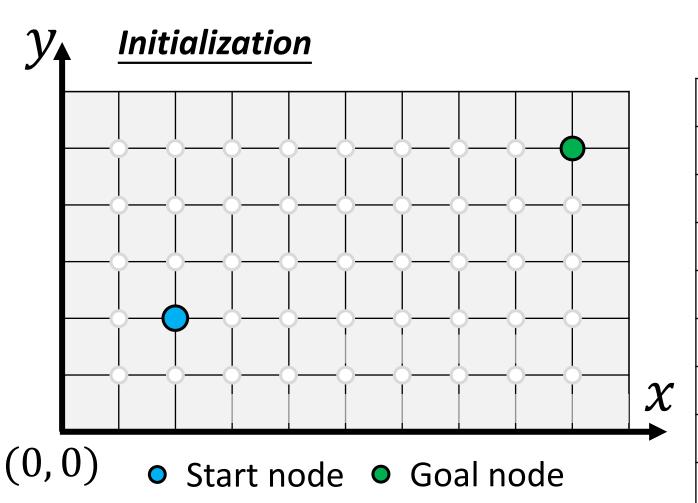
Total cost

Node ID	g(x,y)	h(x,y)	f(x,y)
(2,1)			
(2,2)			
(2,3)			
(3,1)			
(3,3)			
(4,1)			
(4,2)			
(4,3)			

How does computer search step by step and got path?







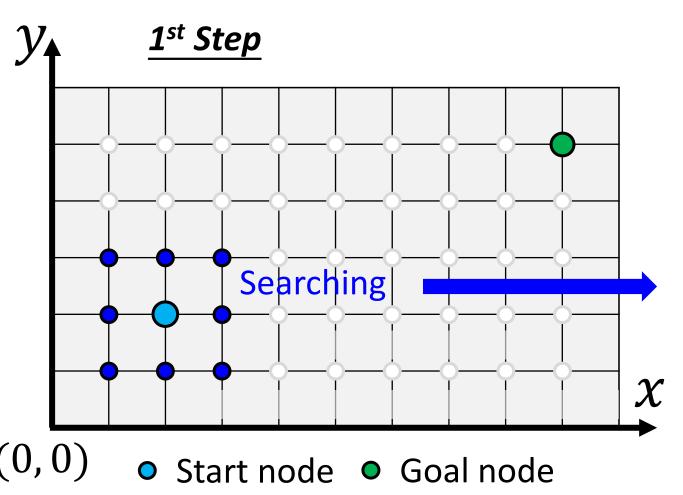
Open List (searched nodes) (arrived nodes)

Close List

Node	f	Source	Node	f	Source
			Start	ı	_







Open List (searched nodes) (arrived nodes)

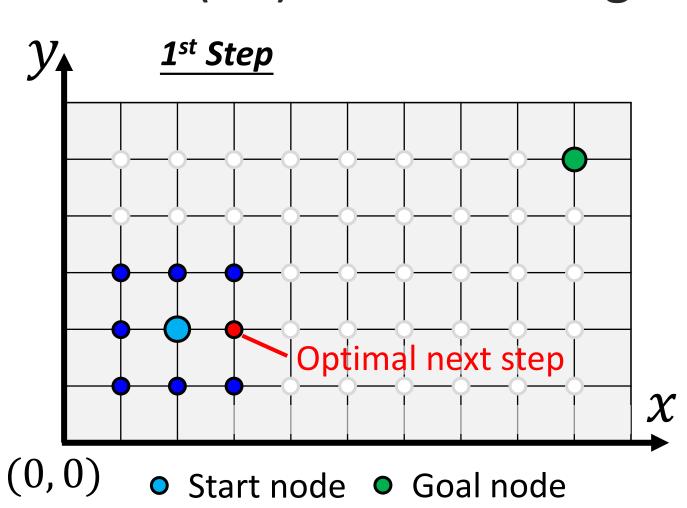
Close List

Node	f	Source
(1,1)	8.9	(2,2)
(1,2)	8.5	(2,2)
(1,3)	8.2	(2,2)
(2,1)	8.1	(2,2)
(2,3)	7.3	(2,2)
(3,1)	7.2	(2,2)
(3,2)	6.7	(2,2)
(3,3)	6.3	(2,2)

Node	f	Source
Start	ı	ı







Open List (searched nodes)

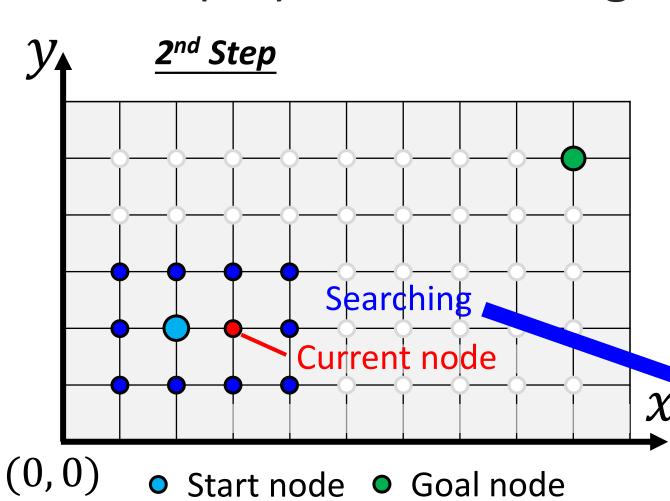
Close List			
(arrived	nodes)		

Node	f	Source
(1,1)	8.9	(2,2)
(1,2)	8.5	(2,2)
(1,3)	8.2	(2,2)
(2,1)	8.1	(2,2)
(2,3)	7.3	(2,2)
(3,1)	7.2	(2,2)
(3.2)	6.7	(2.2)
(3,2)	0.7	(2,2)
(3,3)	6.3	(2,2)

Node	f	Source
Start	ı	-
(3,2)	6.7	(2,2)
	Start	Start -







Open List (searched nodes) (arrived nodes)

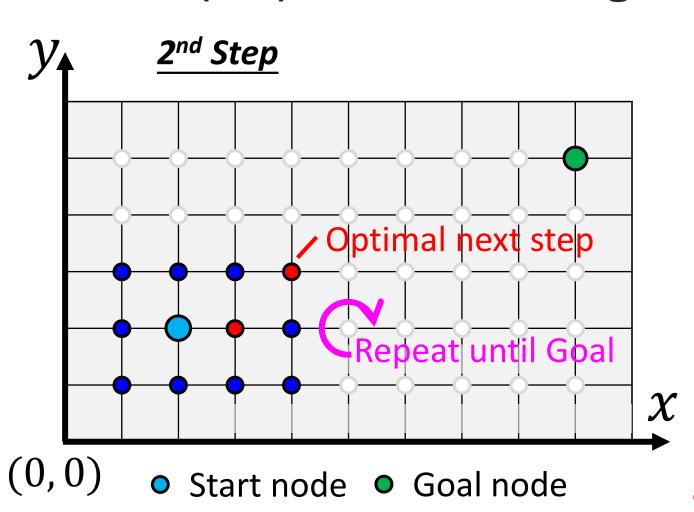
Close List

Node	f	Source
(1,1)	8.9	(2,2)
•	•••	•
(3,1)	7.2	(2,2)
(3,3)	6.3	(2,2)
	• • •	•
(4,1)	7.8	(2,2)
(4,2)	7.8	(2,2)
(4,3)	8.8	(2,2)

f	Source
1	-
6.7	(2,2)
	<i>f</i> - 6.7







Open List (searched nodes) (arrived nodes)

Close List

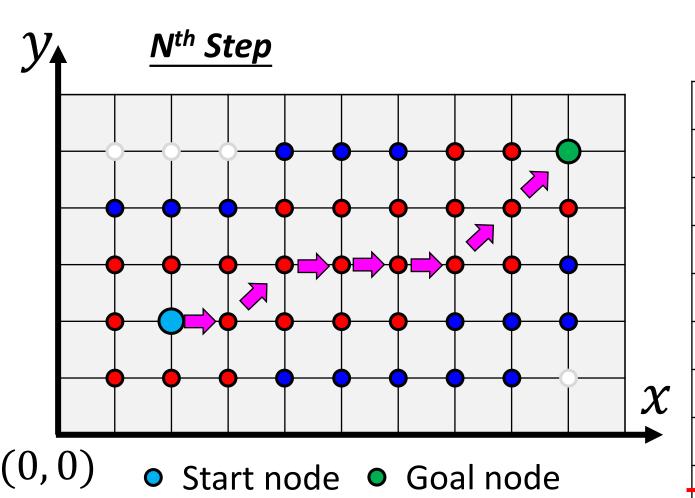
Node	f	Source
(1,1)	8.9	(2,2)
•	•••	•
(3,1)	7.2	(2,2)
(3,3)	6.3	(2,2)
	•••	•
(4,1)	8.8	(3,2)
(4,2)	7.8	(3,2)
(4,3)	7.8	(3,2)

Node	f	Source
Start	1	1
(3,2)	6.7	(2,2)
(4,3)	7.8	(3,2)





A-star (A*) Path Planning – Retrieve Best Path



Open List Close List (searched nodes) (arrived nodes)

Node	f	Source
(1,1)	8.9	(2,2)
•	:	:
(3,1)	7.2	(2,2)
(3,3)	6.3	(2,2)
:	:	:
:	:	:
:	:	i
Goal	X	(x,y)

		1
Node	f	Source
Start	-	-
(3,2)	6.7	(2,2)
(4,3)	7.8	(3,2)
•	i	:
•	i	:
Goal	Х	(x,y)
-	race	back
ar	rived	nodes





A-star Example

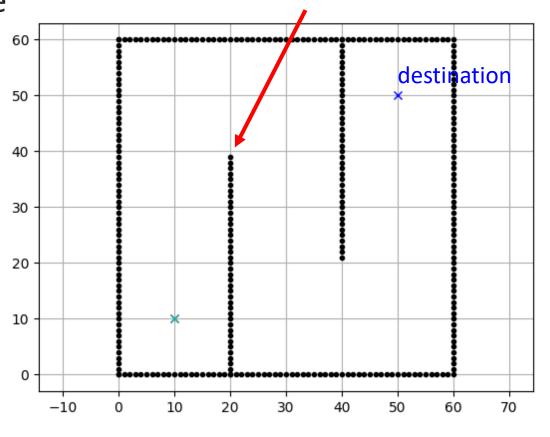
Each time A* enters a node, it calculates the cost: f(n) - n being the neighboring node

It travel to all of the neighboring nodes, and then enters the node with the lowest value of f(n)

These values we calculate using the following formula:

$$f(x,y) = g(x,y) + h(x,y)$$

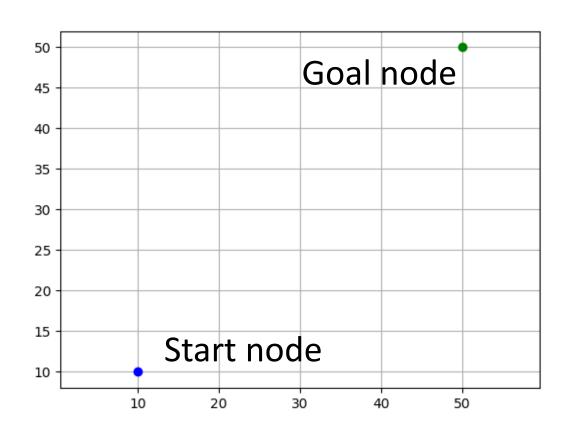
Wall (obstacles) cannot go through!







Code Example (1): Set Up Start and Goal Node



- Start node
- Goal node

```
# start and goal position
sx = 10.0 # [m]
sy = 10.0 # [m]
gx = 50.0 # [m]
gy = 50.0 # [m]
grid_size = 2 # [m]
```

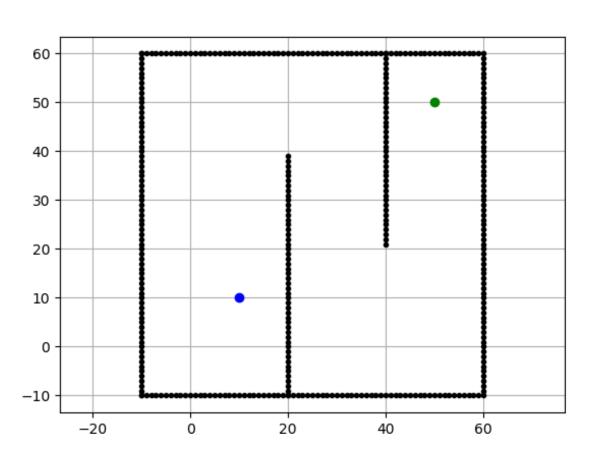
Base code tutorial:

https://www.youtube.com/watch?v=P RKLhcG2kB0&ab channel=POLYUIPNL





Code Example (2): Set Up Obstacle



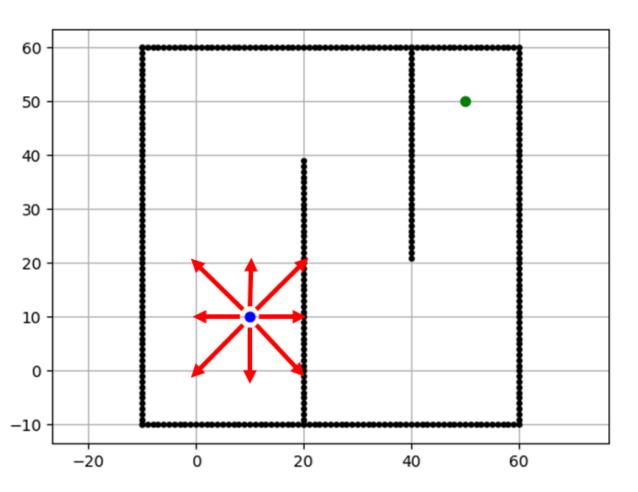
```
# set obstacle positions
ox, oy = [], []
for i in range(-10, 60): # draw the button border
    ox.append(i)
    oy.append(-10.0)
for i in range(-10, 60):
    ox.append(60.0)
    oy.append(i)
for i in range(-10, 61):
    ox.append(i)
    oy.append(60.0)
for i in range(-10, 61):
    ox.append(-10.0)
    oy.append(i)
for i in range(-10, 40):
    ox.append(20.0)
    oy.append(i)
for i in range(0, 40):
    ox.append(40.0)
    oy.append(60.0 - i)
```

- Start node
- Goal node





Code Example (3): Neighboring Node Search



- Start node
- Goal node





Code Example (4): Cost Calculation

Exact cost g(x, y) calculation

Heuristic cost h(x, y) calculation

```
def calc_heuristic(n1, n2):
    w = 1.0  # weight of heuristic
    d = w * math.hypot(n1.x - n2.x, n1.y - n2.y)
    return d
```

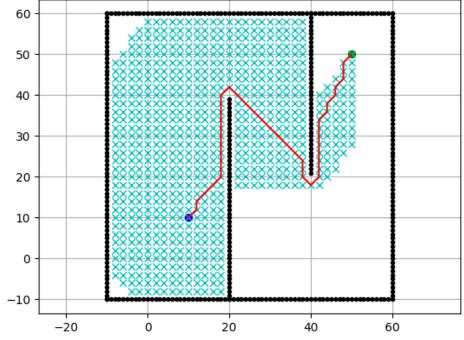




Code Example (5): Calculation of Final Path

```
def calc_final_path(self, goal_node, closed_set):
    # generate final course
    rx, ry = [self.calc_grid_position(goal_node.x, self.min_x)], [
        self.calc_grid_position(goal_node.y, self.min_y)] # save the goal node as the first point
    parent_index = goal_node.parent_index
    while parent_index != -1:
        n = closed_set[parent_index]
        rx.append(self.calc_grid_position(n.x, self.min_x))
        ry.append(self.calc_grid_position(n.y, self.min_y))
        parent_index = n.parent_index
    return rx, ry
```

Retrieve the optimal path from all passing through nodes (once being the current node)





Play with Example Codes (**)







Trip Cost of Flight





Trip Cost of Flight

The fundamental rationale of the cost index concept is to achieve minimum **trip cost** by means of a trade-off between operating costs per hour and incremental fuel burn.

$$C = C_F \cdot \Delta F \cdot T + C_T \cdot T + C_C$$

* Related to travelling time

With

- C_F =cost of fuel per kg
- C_T =time related cost per minute of flight
- C_c =fixed cost independent of time
- ΔF =trip fuel (e.g., 3000kg/h)
- T=trip Time (e.g., 8 hours from Hong Kong to Paris)

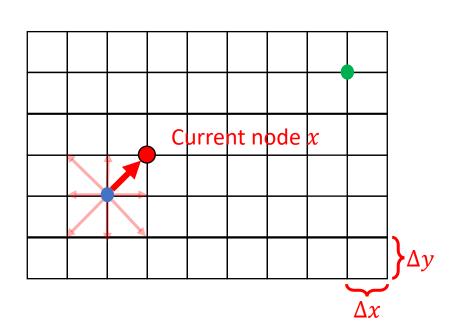
Consider this relationship to imitate the path planning for flights?







Conventional Path Planning Task



Total cost:
$$f(x,y) = g(x,y) + h(x,y)$$

Distance information (conventional)

Example for one step:

$$g(x,y) = \sqrt{\Delta x^2 + \Delta y^2}$$

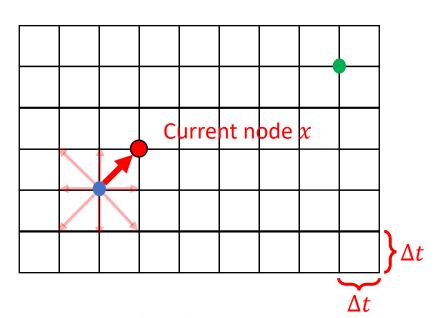
Objective: Find the path with the lowest traveling <u>distance</u>. (in the unit of meter)

- Goal node
- Start node





Path Planning Task for Out Flight



Total cost:
$$f(n) = g(n) + h(n)$$

Traveling time (our case for flight)

Time cost on each step:

$$s(n) = \begin{cases} \Delta t, & vertical/horizontal\ motion \\ \sqrt{2}\Delta t, & diagonal\ motion \end{cases}$$

Example for first step:

$$g(n) = s(n) = \sqrt{2}\Delta t$$

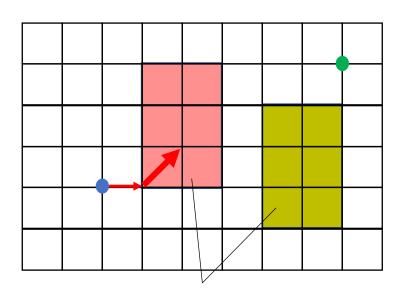
Objective: Find the path with the lowest traveling <u>time</u>. (in the unit of min)

• Goal node • Start node n – index of nodes





Flight Planning Considering Cost Intensive Areas



Cost intensive area (addition time)

Cost Intensive Areas: The cost for flying through such area is increased due to airflow, legal restrictions and other reasons.

Total cost for one node:

$$f(n) = g(n) + \alpha \cdot s(n) + h(n)$$

Previous cost from start

Additional time at current step

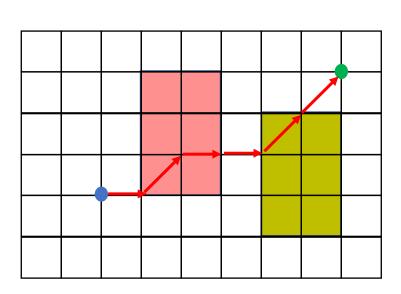
 α – additional time factor (equal zero for normal area)

- Goal node
- Start node





Flight Path Planning Trip Cost





Cost Intensive Areas: The cost for flying

through such area is increased due to

airflow, legal restrictions and other reasons.

Total cost for one node:

$$f(n) = g(n) + \alpha \cdot s(n) + h(n)$$

$$T_{best} = \min[f(goal)]$$

$$T_{best} = \min[f(goal)]$$
 Additional cost on each node
$$= \min[g(goal) + \alpha_1 s(n_1) + \alpha_2 s(n_2) + \dots + \alpha_{goal} s(n_{goal})]$$

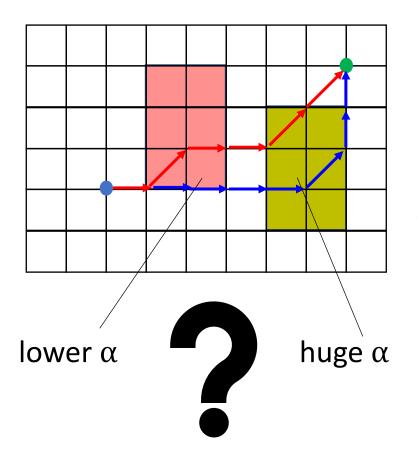
Trip cost for planned path:
$$C = C_F \cdot \Delta F \cdot T_{best} + C_T \cdot T_{best} + C_c$$

- Goal node
- Start node α additional time factor





How it choose different routes?



Cost Intensive Areas: The cost for flying through such area is increased due to airflow, legal restrictions and other reasons.

Time for the planned path:

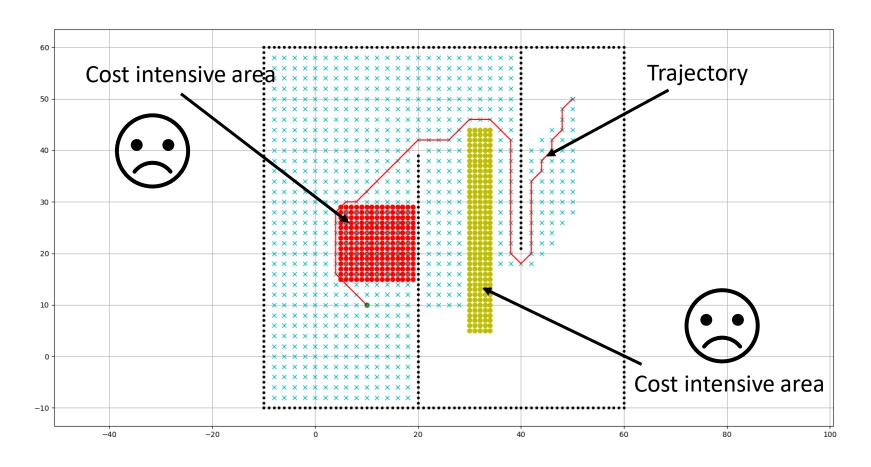
$$T_{best} = \min \left[g(goal) + \alpha_1 s(n_1) + \alpha_2 s(n_2) + \dots + \alpha_{goal} s(n_{goal}) \right]$$

Depending on the extra time accumulated more specifically, the additional time factor





Example of Flight Path Planning

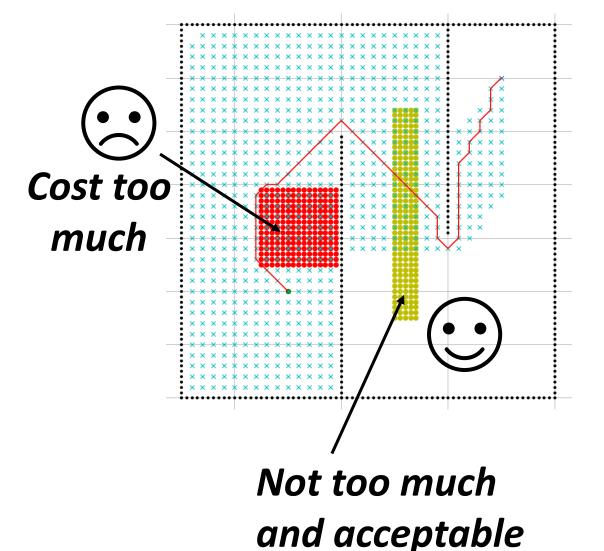


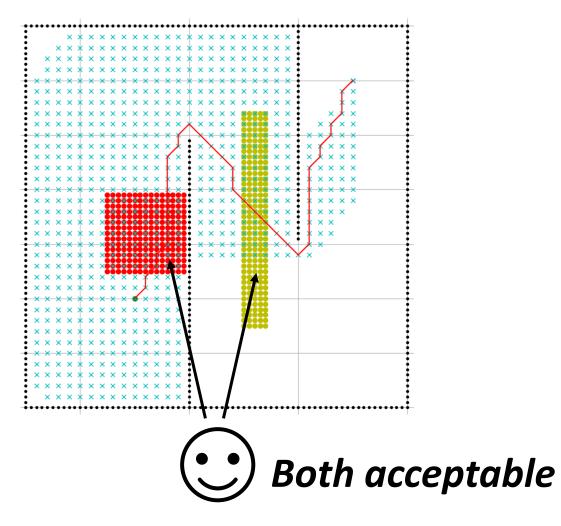
Cost is way too high for going through, better to avoid!





Example of Flight Path Planning









Flight Path Planning Project



You will be creating and completing your own path planning program based on groups



You can find the project tasks / requirements in this slide after the code tutorial



Additional resources could be found inside the course GitHub repository

- Video tutorial
- Tutorial slides



Path Planning Programming Guide





The Path Planning Code

- You can find the path planning code inside the course GitHub repository
- There are 2 set of codes:
 - A default one
 - A noted one
- The default one is a basic A* path planning code without any extra information and features
- The noted one provides an example of what your code should look like after modifications (Remember each group should complete a different set of obstacles and requirements)
- Repository link: https://github.com/IPNL-POLYU/PolyU AAE2004 Github Project





Where you can find the code

PolyU_AAE2004_Github_Project / Sample Codes /		Go to file Add file ▼ ···
This branch is up to date with qmohsu:main.		\$↑ Contribute ▼ 🌣 Fetch upstream ▼
amohsu Merge branch 'main' into LT2		de19c32 on Oct 15, 2021 🖰 History
☐ Tutorial 1 Sample.py	Update Tutorial 1 Sample.py	5 months ago
a_star_noted.py	Add files via upload	5 months ago
a_star_original.py	Merge branch 'main' into LT2	4 months ago
animation.gif	update sample code	5 months ago
readme.md	Update readme.md	5 months ago





- Line 50,51: Declaration of cost intensive area cost modifier
- Line 53: Declare cost per grid

```
self.resolution = resolution # get resolution of the grid
             self.rr = rr # robot radis
             self.min_x, self.min_y = 0, 0
              self.max x, self.max y = 0, 0
              self.obstacle map = None
              self.x_width, self.y_width = 0, 0
              self.motion = self.get motion model() # motion model for grid search expansion
              self.calc obstacle map(ox, oy)
             self.fc x = fc x
             self.fc y = fc y
             self.tc_x = tc_x
             self.tc y = tc y
             self.Delta C1 = 0.2 # cost intensive area 1 modifier
              self.Delta_C2 = 0.4 # cost intensive area 2 modifier
53
             self.costPerGrid = 1
```





- Line 115: Showing the final calculation of total trip time
- Line 135-144: Adding additional cost during cost intensive area

```
if show_animation: # pragma: no cover
                      plt.plot(self.calc grid position(current.x, self.min x),
                               self.calc grid position(current.y, self.min y), "xc")
                      # for stopping simulation with the esc key.
                      plt.gcf().canvas.mpl_connect('key_release_event',
108 🗸
                                                   lambda event: [exit(
                                                       0) if event.key == 'escape' else None])
110 🗸
                      if len(closed_set.keys()) % 10 == 0:
                          plt.pause(0.001)
                  # reaching goal
                  if current.x == goal node.x and current.y == goal node.y:
                      print("Total Trip time required -> ",current.cost )
                      goal node.parent index = current.parent index
                      goal node.cost = current.cost
118
                      break
                  # Remove the item from the open set
                  del open set[c id]
                  # Add it to the closed set
                  closed set[c id] = current
                  # expand grid search grid based on motion model
129 🗸
                  for i, in enumerate(self.motion): # tranverse the motion matrix
130 🗸
                      node = self.Node(current.x + self.motion[i][0],
                                       current.y + self.motion[i][1],
                                       current.cost + self.motion[i][2] * self.costPerGrid, c_id)
                      ## add more cost in cost intensive area 1
                      if self.calc grid position(node.x, self.min x) in self.tc x:
136 🗸
                          if self.calc_grid_position(node.y, self.min_y) in self.tc_y:
                              node.cost = node.cost + self.Delta C1 * self.motion[i][2]
                      # add more cost in cost intensive area 2
                      if self.calc_grid_position(node.x, self.min_x) in self.fc_x:
                          if self.calc grid position(node.y, self.min y) in self.fc y:
                              # print("cost intensive area!!")
                              node.cost = node.cost + self.Delta_C2 * self.motion[i][2]
```





- Line 263-270: Declaring motions for the aircraft
- Line 279-284: Declaring starting point and end point

```
@staticmethod
          def get motion model(): # the cost of the surrounding 8 points
              motion = [[1, 0, 1],
                        [0, 1, 1],
                        [-1, 0, 1],
                        [0, -1, 1],
                        [-1, -1, math.sqrt(2)],
                        [-1, 1, math.sqrt(2)],
                        [1, -1, math.sqrt(2)],
269
                        [1, 1, math.sqrt(2)]]
270
271
              return motion
      def main():
275
          print(__file__ + " start the A star algorithm demo !!") # print simple notes
276
          # start and goal position
278
279
          SX = 0.0 \# [m]
          sy = 0.0 \# [m]
          gx = 50.0 \# [m]
          gy = 0.0 \# [m]
          grid size = 1 # [m]
283
          robot radius = 1.0 # [m]
```





- Line 309-329: Adding obstacles
- Line 337-348, Adding cost intensive areas (Hint: Refer to this part for your task 2!)

```
# set obstacle positions for group 9
          ox, oy = [], []
          for i in range(-10, 60): # draw the button border
              ox.append(i)
              oy.append(-10.0)
          for i in range(-10, 60): # draw the right border
              ox.append(60.0)
              oy.append(i)
          for i in range(-10, 60): # draw the top border
              ox.append(i)
              oy.append(60.0)
          for i in range(-10, 60): # draw the left border
              ox.append(-10.0)
              oy.append(i)
          for i in range(-10, 30): # draw the free border
              ox.append(20.0)
              oy.append(i)
          for i in range(0, 20):
              ox.append(i)
              oy.append(-1 * i + 10)
          # for i in range(40, 45): # draw the button border
                oy.append(30.0)
          # set cost intesive area 1
          fc_x, fc_y = [], []
          for i in range(30, 40):
              for j in range(0, 40):
                  fc_x.append(i)
                  fc_y.append(j)
          # set cost intesive area 1
344
          tc x, tc y = [], []
          for i in range(10, 20):
              for j in range(20, 50):
                  tc x.append(i)
                  tc_y.append(j)
```





Bonus!

• If you wish to do the <u>trip cost</u> (in terms of expense) calculation using the program, you should add the calculation function under line 117, inside the reaching goal condition

• It would be even better if the program could distinguish viable and non-viable aircraft types!

• Use the noted version as your example to modify your own code!





Bonus Showcase

When you add in a cost calculation function, the output should look something like this, it should be able to:

- 1. Calculate and show each aircraft types' operating costs
- 2. Mention which type might not be viable for certain scenarios

```
min_x: -10
min_y: -10
max_x: 60
max_y: 60
x_width: 70
y_width: 70
Total travelling time -> 93.35575746753788
A321 not viable!
Total cost of operating A330 in this scenario: 27360.167918740684
Total cost of operating A350 in this scenario: 30752.648960130347
```