Q1

ca)
$$P(\theta|Y_{1}...Y_{m}) = \frac{P(Y_{1}...Y_{m}|\theta)P(\theta)}{P(Y_{1}...Y_{m})}$$
The estimated error implies that all Y are independent

Thus above $\alpha P(Y_{1}|\theta)P(Y_{2}|\theta) ...P(Y_{n}|\theta)P(\theta)$

(b) (1) $\Rightarrow \epsilon_{1} = Y_{1} - A_{1}\theta - b_{1} \quad \text{plug in } (3)$

$$P(Y_{1}|\theta) = \frac{1}{\int \det(2\pi X_{1})} \exp\left(-\frac{1}{2}(Y_{1} - A_{1}\theta - b_{1} - u_{1})\right)$$

$$\sum_{1}^{1} (Y_{1} - A_{1}\theta - b_{1} - u_{1})$$

(c) We know that $P(\theta|Y_{1}...Y_{m}) \propto P(\theta) \prod_{i=1}^{m} P(Y_{i}|\theta)$

while $P(\theta)$ can plug into (3)
$$P(\theta|Y_{1}...Y_{m}) \propto \exp\left(-\frac{1}{2}(\theta - u_{0})^{T} \sum_{0}^{1}(\theta - u_{0})\right) \times \prod_{1}^{m} \exp\left(-\frac{1}{2}(Y_{1} - A_{1}\theta - b_{1} - u_{1})\right)$$

 $\sum_{i}^{-1} \left(\sum_{i}^{-1} - A_{i} \theta - b_{i} - u_{i} \right)$

(a)

(i)

```
def recoverPath(start, goal, pred):
    length = list()
    trace_back = [goal]
    while goal != start:
        length.append(d(goal, pred[goal]))
        goal = pred[goal]
        trace_back.append(goal)
    trace_back = list(reversed(trace_back))
    return trace_back, sum(length)
```

(ii)

```
def A_star(Vertex, start, goal, Neighbor, weight, heuristic):
    # Initialization
    pred = dict()
    CostTo = dict()
    EstTotalCost = dict()
    q = list()

for v in Vertex:
    CostTo[v] = float('inf')
    EstTotalCost[v] = float('inf')

CostTo[start] = 0
    EstTotalCost[start] = heuristic(start, goal)

heapq.heappush(Q, (heuristic(start, goal), start))

while Q:
    priority, v = heapq.heappop(Q)
    # If reached the goal, end the loop and start the trace back algorithm
    if v == goal:
        return recoverPath(start, goal, pred)

for index in Neighbor(v):
    # print(f'{floate}: {list(Neighbor(v))}')
    pvi = CostTo[v] + weight(v, index)
    if pvi < CostTo[index]:
    # Update based on heuristic
        pred[index] = v
        CostTo[index] = pvi
        EstTotalCost[index] = pvi + heuristic(index, goal)
        # insert here
        # heapq.has(index) ?
        # if any(index == b for a, b in Q):
        # heapq.heappush(Q, (heuristic(index, goal), index))

return None</pre>
```

(i)

```
col, row = v
return (
          (col + g, row + h)
          for g in (-1, 0, 1)
          for h in (-1, 0, 1)
          if g != 0 or h != 0
          if 0 <= col + g & col + g < len(occupancy_map)
          if 0 <= row + h & row + h < len(occupancy_map[0])
          if occupancy_map[col + g][row + h] == 1)</pre>
```

This ensures that g and h is the 8 surrounding neighbor of the chosen v and it's not out of bound, and lastly it's a valid space. This pastes it all into a list and returns it.

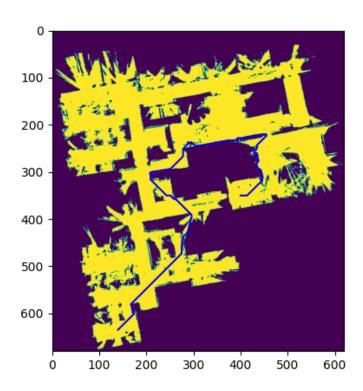
(ii)

```
def d(v1, v2):
    return math.dist([v1[0], v1[1]], [v2[0], v2[1]])
```

It can be done with math, dist in one line

(iii)

And the graph generated for A* is as follows



The total length of the path taken:

```
Sum of length taken is: 938.2834046956888
```

938.2834

(c)

(i)

```
def sample(G):
    row, col = G.shape
    sample1 = random.uniform(0, row)
    sample2 = random.uniform(0, col)
    while G.item(int(sample1), int(sample2)) == 0:
        sample1 = random.uniform(0, row)
        sample2 = random.uniform(0, col)
    out = (int(sample1), int(sample2))
    return out
```

```
def valid_Path(G, start, goal):
    for v in list(bresenham(start[0], start[1], goal[0], goal[1])):
        if G.item(v) == 0:
            return False
        return True
```

Bresenham returns all the vertices on a straight line and this algorithm checks if they are occupied or not

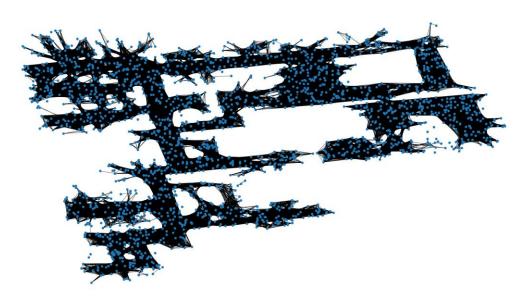
(iii)

```
def add_Vertex(G, vnew, dmax):
   V.append(vnew)
   PRM.add_node(PRM.number_of_nodes() + 1, pos=vnew)
        if i != vnew and d(i, vnew) < dmax and valid_Path(G, i, vnew):</pre>
           PRM.add_edge((i, vnew), weight=d(i, vnew))
           E.append((i, vnew))
   return PRM
def construct_PRM(G, N, dmax):
       vnew = sample(G)
       add_Vertex(G, vnew, dmax)
   return PRM
V = list()
PRM = networkx.Graph()
```

With the provided Algorithm, networkX provided the graph and the nodes

and edges are all added to the PRM constructed by it.

(iv)



```
V = list()
E = list()
N = 2500
dmax = 75
start = (635, 140)
goal = (350, 400)

PRM = nx.Graph()
PRM_complete = construct_PRM(occupancy_grid, N, dmax)

# Extract pos attr prom the PRM constructed
pos_list = nx.get_node_attributes(PRM_complete, 'pos')

# print(pos_list.get(1))

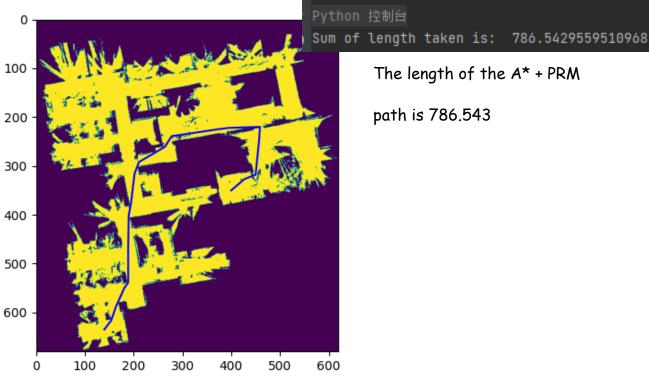
repos_list = {}

# Inverse the graph's coordinate to fit the original graph

ofor i in range(1, PRM_complete.number_of_nodes() + 1):
    inv_X = -pos_list.get(i)[0]
    inv_Y = pos_list.get(i)[1]
    repos_list[i] = (inv_Y, inv_X)

output_graph = matplotlib.pyplot.figure(1, figsize=(150, 150), dpi=60)
nx.draw_networkx(PRM_complete, repos_list, node_size=100, linewidths=0.1, with_labels=False)
```

```
add_Vertex(occupancy_grid, goal, dmax)
startIndex = PRM_complete.number_of_nodes() - 1
goalIndex = PRM_complete.number_of_nodes()
for t in range(PRM_complete.number_of_nodes()):
    if V[t] == start:
        startIndex = t + 1
A_star = nx.astar_path(PRM_complete, startIndex, goalIndex)
length = nx.astar_path_length(PRM_complete, startIndex, goalIndex)
print("Sum of length taken is: ", length)
    path.append(PRM_complete.nodes[e]['pos'])
matplotlib.pyplot.imshow(matplotlib.pyplot.imread(
matplotlib.pyplot.plot(numpy.array(path)[:, 1], numpy.array(path)[:, 0], 'b')
matplotlib.pyplot.show()
```



The length of the $A^* + PRM$ path is 786.543