

Problem 2

Given that the M obstacles are constant and only the N drones are moving, we can construct a kd-tree (2d-tree in this case) to calculate the nearest obstacle to each of the drones.

Tree Construction

Consider O = list of obstacles (length M) and D = list of drones (length N).

```
class KDNode(x,y, x_flag)
{
    x, y,
    x_flag,
    leftChild = NULL,
    rightChild = NULL,
    selection_flag = False
}
```

```
ConstructKDTree(O, x_flag = True):
    pivot = new KDNode(O[0].x, O[0].y, x_flag)
    if len(O) == 1:
        return pivot
    leftTree = [ ], rightTree = [ ]
    if x_flag == True:
        For each point p in O:
            if p.x < pivot.x:
                Add p to leftTree
            else:
                Add p to rightTree
    else:
        For each point p in O:
            if p.y < pivot.y:
                Add p to leftTree
            else:
                Add p to rightTree
    pivot.leftChild = ConstructKDTree(leftTree, !x_flag)
    pivot.rightChild = ConstructKDTree(rightTree, !x_flag)
```

Height of tree (h) = $O(\log M)$ (avg. case)

Thus, search time complexity = $O(h)$ (avg. case)

FindNearestNeighbor(Drone d, KDTree root):

```
    min_dist = inf
    min_point = None
    if root.leftChild == NULL and root.rightChild == NULL:
        if not root.selection_flag and dist(root,d) < min_dist:
            min_point = root
            min_dist = dist(root, d)
        return (min_point, min_dist)
    if not root.selection_flag:
        min_point = root
        min_dist = dist(root, d)
    if root.axis_flag:
        if d.x < root.x:
            min_point1, min_dist1 = Find Nearest Neighbor(d, root.leftChild)
            if min_dist1 < min_dist:
                min_dist = min_dist1
                min_point = min_point1
            if abs(d.x - root.x) < min_dist:
                min_point1, min_dist1 = Find Nearest Neighbor(d, root.rightChild)
                if min_dist1 < min_dist:
                    min_dist = min_dist1
                    min_point = min_point1
        else:
            min_point1, min_dist1 = Find Nearest Neighbor(d, root.rightChild)
            if min_dist1 < min_dist:
                min_dist = min_dist1
                min_point = min_point1
            if abs(d.x - root.x) < min_dist:
                min_point1, min_dist1 = Find Nearest Neighbor(d, root.leftChild)
                if min_dist1 < min_dist:
                    min_dist = min_dist1
                    min_point = min_point1
    else:
        if d.y < root.y:
            min_point1, min_dist1 = Find Nearest Neighbor(d, root.leftChild)
            if min_dist1 < min_dist:
                min_dist = min_dist1
                min_point = min_point1
            if abs(d.y - root.y) < min_dist:
                min_point1, min_dist1 = Find Nearest Neighbor(d, root.rightChild)
                if min_dist1 < min_dist:
                    min_dist = min_dist1
                    min_point = min_point1
```

```

else:
    min_point1, min_dist1 = Find Nearest Neighbor(d, root.rightChild)
    if min_dist1 < min_dist:
        min_dist = min_dist1
        min_point = min_point1
    if abs(d.y - root.y) < min_dist:
        min_point1, min_dist1 = Find Nearest Neighbor(d, root.leftChild)
        if min_dist1 < min_dist:
            min_dist = min_dist1
            min_point = min_point1
return min_dist, min_point

```

Problem 3

In order to find the K-Nearest neighbors, we run the above FindNearestNeighbor algorithm K times. Note that we set the min_point.selection_flag to True once it is selected. This prevents the algorithm from selecting the same node again and again. This FindKNearestNeighbors algorithm can be run for all the drones.

FindKNearestNeighbors(Drone d, KDTree root, int k):

```

knn = [ ]
for i in range(k):
    min_dist, min_point = FindNearestNeighbors(d, root)
    knn.append((min_dist, min_point))
    min_point.selection_flag = True
return knn

```

Time complexity = $O(k.h)$, (avg case) where k = number of neighbors required and h = height of the tree

Problem 4

Similar to OctTree, but with a slight modification. Given a space S we first create a random cuboid C inside the space given to us, and then, we create 8 such non-intersecting subspaces whose union is S - C and the intersection of any two subspaces is NULL.

Let a cuboid be defined by its beginning and ending coordinates

```

class Cuboid
{

```

```
        x1, y1, z1, x2, y2, z2;  
};
```

Let us define an Octree as

```
class Octree  
{  
    Cuboid c,  
    Octree child[8]  
};
```

GenerateNCuboids(int n):

```
list = [ ]  
S = Cuboid(0,0,0,10,10,10)  
subsp = { S }  
for i in range(n):  
    S = extract random subsp from S and remove it  
    Let X1, X2, Y1, Y2, Z1, Z2 be the coordinates of S  
    C = random cuboid inside S  
    Let x1, x2, y1, y2, z1, z2 be the coordinates of C  
    Add C to list  
    subspace s1 = Cuboid(x2, X2, y2, Y2, z2, Z2)  
    subspace s2 = Cuboid(X1, x1, Y1, y1, z1, Z1)  
    subspace s3 = Cuboid(x2, X2, Y1, y2, z2, Z2)  
    subspace s4 = Cuboid(X1,x2, y1, Y2, z2, Z2)  
    subspace s5 = Cuboid(X1, x2, Y1, y1, z1, Z2)  
    subspace s6 = Cuboid(x1, X2, Y1, y2, Z1, z1)  
    subspace s7 = Cuboid(x1, X2, y2, Y2, Z1, z2)  
    subspace s8 = Cuboid(X1, x1, y1, Y2, Z1, z2)  
    Add the new subspaces to subsp
```