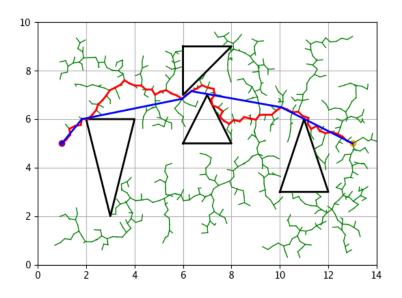
CS 133b Problem Set 4

Problem 1 (Rapidly-exploring Random Tree Planner) - 48 points

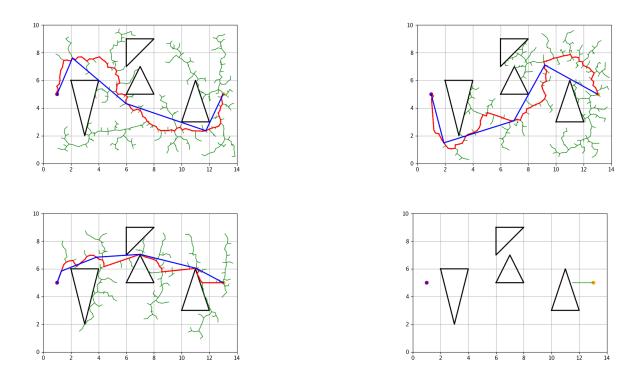


part (a)
The code is shown on the next page. An example of output is shown above.

Code for problem 1a

```
def rrt(startnode, goalnode, visual):
    startnode.parent = None
    tree = [startnode]
    # Function to attach a new node to an existing node: attach the
    def addtotree(oldnode, newnode):
        newnode.parent = oldnode
        tree.append(newnode)
        visual.drawEdge(oldnode, newnode, color='g', linewidth=1)
        visual.show()
    # Loop - keep growing the tree.
    steps = 0
        #FIXME:
        x_coord = random.uniform(xmin, xmax)
        y_coord = random.uniform(ymin, ymax)
        targetnode = Node(x_coord, y_coord)
        distances = np.array([node.distance(targetnode) for node in tree])
        index = np.argmin(distances)
        nearnode = tree[index] #nearest node to target
                 = distances[index] #distance between nearest node and target
        #FIXME:
        if d <= DSTEP:</pre>
           nextnode = targetnode
            dir_vec = np.array([targetnode.x - nearnode.x, targetnode.y - nearnode.y])
            dir_vec = dir_vec/np.linalg.norm(dir_vec)
           new_x = nearnode.x + DSTEP * dir_vec[0]
            new_y = nearnode.y + DSTEP * dir_vec[1]
           nextnode = Node(new_x, new_y)
        if nextnode.inFreespace() and nearnode.connectsTo(nextnode):
           addtotree(nearnode, nextnode)
           # If within DSTEP, also try connecting to the goal. If
           # the connection is made, break the loop to stop growing.
            dist_togoal = nextnode.distance(goalnode)
            if dist_togoal <= DSTEP and nextnode.connectsTo(goalnode):</pre>
                addtotree(nextnode, goalnode)
        steps += 1
        if (steps >= SMAX) or (len(tree) >= NMAX):
            print("Aborted after %d steps and the tree having %d nodes" %
                 (steps, len(tree)))
            return None
    path = [goalnode]
    while path[0].parent is not None:
```

For the first row: 5% probability on the left, 50% probability on the right For the second row: 99% probability on the left, 100% probability on the right

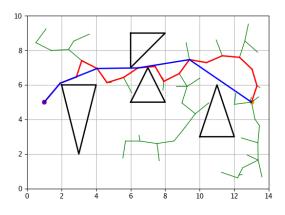


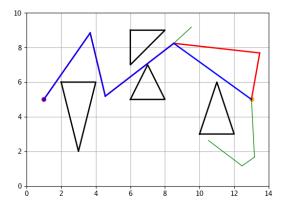
part (b)

As you increase the probability that the goal node is selected as the target node, then the tree becomes smaller (less nodes are added to the tree). In the case that the probability is 100 percent, no path is ever made between the start and goal because there is an obstacle. With probability of 99 percents, there are instances when a path cannot be found given the step size and the max number of steps (attempts), though in the output provided above a path is found. The code is on the next page, the probability of that the goal node is selected as the target node is adjusted accordingly (variable goal_p).

Code for problem 1b

```
def rrt(startnode, goalnode, visual):
   startnode.parent = None
   tree = [startnode]
   def addtotree(oldnode, newnode):
       newnode.parent = oldnode
       tree.append(newnode)
       visual.drawEdge(oldnode, newnode, color='g', linewidth=1)
       visual.show()
   # Loop - keep growing the tree.
   steps = 0
       goal_p = 0.05
       val = np.random.choice(a = np.array([0,1]), p = np.array([goal_p, 1-goal_p]))
       if val == 0:
           targetnode = goalnode
           x_coord = random.uniform(xmin, xmax)
           y_coord = random.uniform(ymin, ymax)
           targetnode = Node(x_coord, y_coord)
       distances = np.array([node.distance(targetnode) for node in tree])
               = np.argmin(distances)
       nearnode = tree[index] #nearest node to target
                = distances[index] #distance between nearest node and target
       #FIXME:
       if d <= DSTEP:</pre>
          nextnode = targetnode
           dir_vec = np.array([targetnode.x - nearnode.x, targetnode.y - nearnode.y])
           dir_vec = dir_vec/np.linalg.norm(dir_vec)
           new_x = nearnode.x + DSTEP * dir_vec[0]
           new_y = nearnode.y + DSTEP * dir_vec[1]
           nextnode = Node(new_x, new_y)
       if nextnode.inFreespace() and nearnode.connectsTo(nextnode):
           addtotree(nearnode, nextnode)
           # the connection is made, break the loop to stop growing.
           dist_togoal = nextnode.distance(goalnode)
           if dist_togoal <= DSTEP and nextnode.connectsTo(goalnode):</pre>
               addtotree(nextnode, goalnode)
               break
       # Check whether we should abort - too many steps or nodes.
       steps += 1
       if (steps >= SMAX) or (len(tree) >= NMAX):
           print("Aborted after %d steps and the tree having %d nodes" %
               (steps, len(tree)))
           return None
   path = [goalnode]
   while path[0].parent is not None:
       path.insert(0, path[0].parent)
```

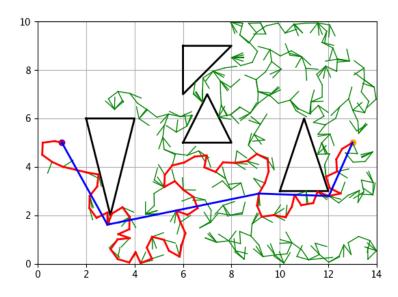




part (c)

The top graph corresponds to an output given when the step size was 1, the bottom graph corresponds to a step size of 5. As shown above we can see that the number of nodes in the tree significantly decrease from when the step size was 0.25 (part b), especially when the step size is set to 5. Additionally the number of steps is also decreased. The code is the same as in 1b (shown in the previous page), the step size (DSTEP) is adjusted accordingly.

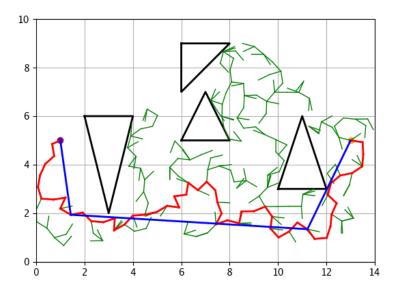
Problem 2 (Expansive-Spaces Tree Planner) - 48 points

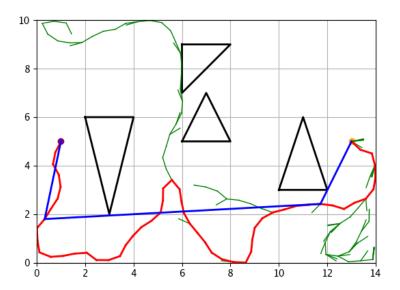


part (a) A function and attribute were added to the Node class (shown below). The rest of the code changed is shown on the next page (part of the est function). DSTEP was set to 0.50. An example of output is shown above.

Code for problem 2a (within est function)

```
while True:
   # Determine the local density by the number of nodes nearby.
   # KDTree uses the coordinates to compute the Euclidean distance.
   # It returns a NumPy array, same length as nodes in the tree.
   X = np.array([node.coordinates() for node in tree])
   kdtree = KDTree(X)
   numnear = kdtree.query_ball_point(X, r=1.5*DSTEP, return_length=True)
   # Directly determine the distances to the goal node.
   distances = np.array([node.distance(goalnode) for node in tree])
   num_neighbors = np.array([node.num_neighbors(tree, DSTEP*1.50) for node in tree])
   min_neighbors = np.min(num_neighbors)
   # get the indices that correspond to nodes with min_neighbors amount of neighbors
   indices = np.where(num_neighbors == min_neighbors)[0]
   i = np.random.choice(indices)
   grownode = tree[i]
   # Check the incoming heading, potentially to bias the next node.
    if grownode.parent is None:
       heading = 0
       heading = atan2(grownode.y - grownode.parent.y,
                       grownode.x - grownode.parent.x)
    # Find something nearby: keep looping until the tree grows.
    while True:
       #FIXME:
       heading = random.uniform(-pi, pi)
       dx = DSTEP * cos(heading)
       dy = DSTEP * sin(heading)
       x\_coord = grownode.x + dx
       y_coord = grownode.y + dy
       nextnode = Node(x_coord, y_coord)
       #FIXME...
        if nextnode.inFreespace() and grownode.connectsTo(nextnode):
            addtotree(grownode, nextnode)
           break
   dist_togoal = nextnode.distance(goalnode)
    if dist_togoal <= DSTEP and nextnode.connectsTo(goalnode):</pre>
       addtotree(nextnode, goalnode)
       break
```



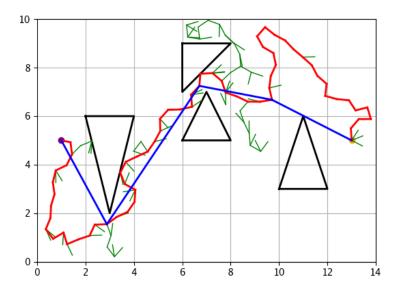


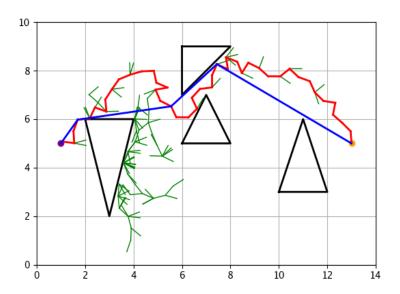
part (b)

A function and attribute were added to the Node class (same as shown in part a). The rest of the code changed is shown on the next page (part of the est function). DSTEP was set to 0.50. Example of outputs are shown above. The top graph is with a standard deviation of pi/2 while the bottom is with a standard deviation of pi/8. As you can see, in both cases, the branches are straighter than they were in part a. Though this is more noticeable with the smaller standard deviation of pi/8.

Code for problem 2b (within est function)

```
while True:
    X = np.array([node.coordinates() for node in tree])
    kdtree = KDTree(X)
    numnear = kdtree.query_ball_point(X, r=1.5*DSTEP, return_length=True)
    # Directly determine the distances to the goal node.
    distances = np.array([node.distance(goalnode) for node in tree])
    # Select the node from which to grow, which minimizes some metric.
    # get the number of neighbors within distance DSTEP*1.50 of every node
    num_neighbors = np.array([node.num_neighbors(tree, DSTEP*1.50) for node in tree])
    min_neighbors = np.min(num_neighbors)
    # get the indices that correspond to nodes with min_neighbors amount of neighbors
    indices = np.where(num_neighbors == min_neighbors)[0]
    i = np.random.choice(indices)
    grownode = tree[i]
    if grownode.parent is None:
       heading = 0
    else:
        heading = atan2(grownode.y - grownode.parent.y,
                        grownode.x - grownode.parent.x)
    # Find something nearby: keep looping until the tree grows.
    while True:
        #FIXME:
        std = pi/2 #standard deviation
        theta = np.random.normal(loc = heading, scale = std)
        dx = DSTEP * cos(theta)
        dy = DSTEP * sin(theta)
        x\_coord = grownode.x + dx
        y_coord = grownode.y + dy
        nextnode = Node(x_coord, y_coord)
        # Try to connect.
        #FIXME...
        if nextnode.inFreespace() and grownode.connectsTo(nextnode):
            addtotree(grownode, nextnode)
            break
    dist_togoal = nextnode.distance(goalnode)
    if dist_togoal <= DSTEP and nextnode.connectsTo(goalnode):</pre>
        addtotree(nextnode, goalnode)
        break
```





part (c)

A function and attribute were added to the Node class (same as shown in part a). The rest of the code changed is shown on the next page (part of the est function). DSTEP was set to 0.50. Examples of outputs are shown above. The top graph is with a scale of 1, while the bottom is with a scale of 5. As you can see, in both cases, the tree now tries to go towards the goal.

Code for problem 2c (within est function)

```
while True:
   X = np.array([node.coordinates() for node in tree])
   kdtree = KDTree(X)
   numnear = kdtree.query_ball_point(X, r=1.5*DSTEP, return_length=True)
   # Directly determine the distances to the goal node.
   scale = 1.0 #scale factor used in metric
   distances = np.array([node.distance(goalnode) for node in tree])
   # Select the node from which to grow, which minimizes some metric.
   num_neighbors = np.array([node.num_neighbors(tree, DSTEP*1.50) for node in tree])
   metric = num_neighbors + scale * distances
   min_val = np.min(metric)
   indices = np.where(metric == min_val)[0]
   i = np.random.choice(indices)
   grownode = tree[i]
   # Check the incoming heading, potentially to bias the next node.
   if grownode.parent is None:
       heading = 0
       heading = atan2(grownode.y - grownode.parent.y,
                        grownode.x - grownode.parent.x)
   while True:
       #FIXME:
       std = pi/2 #standard deviation
       theta = np.random.normal(loc = heading, scale = std)
       dx = DSTEP * cos(theta)
       dy = DSTEP * sin(theta)
       x\_coord = grownode.x + dx
       y_coord = grownode.y + dy
       nextnode = Node(x_coord, y_coord)
       #FIXME...
       if nextnode.inFreespace() and grownode.connectsTo(nextnode):
           addtotree(grownode, nextnode)
           break
   dist_togoal = nextnode.distance(goalnode)
   if dist_togoal <= DSTEP and nextnode.connectsTo(goalnode):</pre>
       addtotree(nextnode, goalnode)
       break
```

Problem 3 (Time Spent) - 4 points

I spent about 2.5 hours on this set. No bottlenecks for me.