Probability & Statistics Project

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Q1: Random Walk

1.1

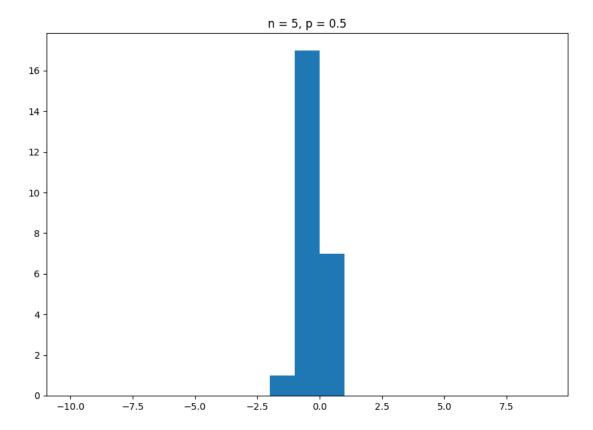
Function implementation in Python:

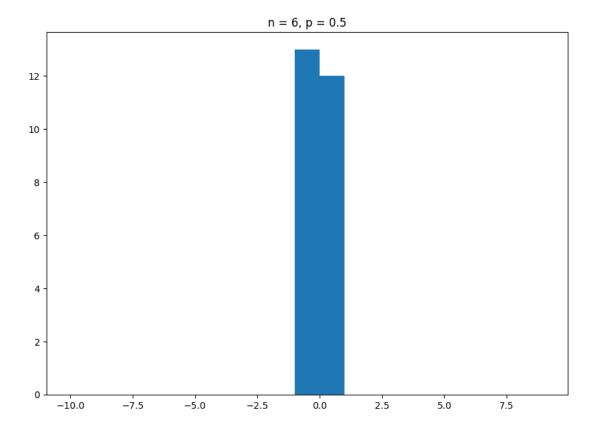
```
def get_updated_position(n,p):
    pos = 0 #position
    for _ in range(n):
        rand = random.randint(1,100) #generating a random number in the range 1 to 100
        if rand < p*100:
            pos += 1 #move one step right
        else:
            pos -= 1 #move one step left
    return pos #return final position</pre>
```

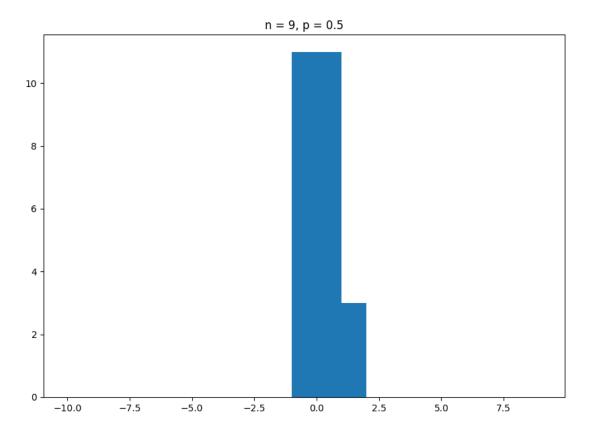
Calling the function for several iterations to get multiple expected values:

```
''', Calculating expected outcomes for various combinations of n and p''',
       expected = []
      outcomes = []
4
      p = 0.5
5
      while p \le 0.9: #for values of p from 0.4 to 0.9
           for n in range(5,11): #for values of n from 5 to 10
               for j in range(25): #expected value for each (n,p) for 25 iterations
9
                   for i in range(25):
                       outcomes.append(get_updated_position(n,p))
11
                   {\tt expected.append(sum(outcomes)/25)} \ {\tt \#appending the expected (average)} \ {\tt value for}
      each(n,p)
                   outcomes = [] #resetting outcomes list
12
13
               #plotting and showing a histogram of calculated expected values
               fig, ax = plt.subplots(figsize =(10, 7))
14
               ax.hist(expected, bins = range(-10,10))
15
               plt.title('n = '+str(n)+', p = '+str(p))
16
               plt.savefig("Q1_histograms/q1"+'n = '+str(n)+', p = '+str(p)+'.png')
17
18
               plt.show()
               expected = [] #reset expected list
19
           p = round((p+0.1),1) #incrementing
20
```

Histograms produced by the above code for various combinations of n and p:







1.2

Function implementation in Python:

```
def get_updated_position_restricted(n,p):
    pos = 0 #position

for _ in range(n):
        rand = random.randint(1,100) #generating a random number in the range 1 to 100

if rand < p*100 or pos <= 0: #move one step right if pos == 0
        pos += 1

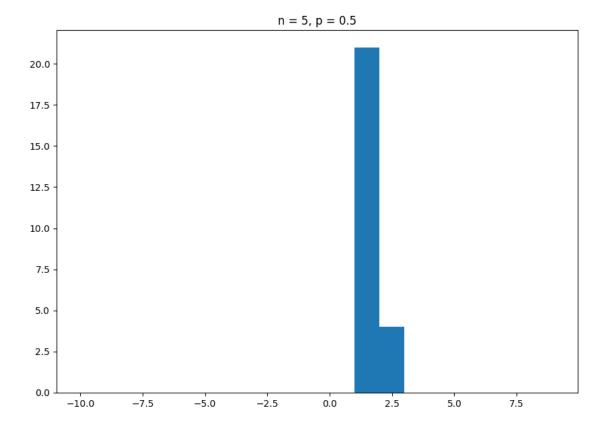
else:
        pos -= 1 #move one step left

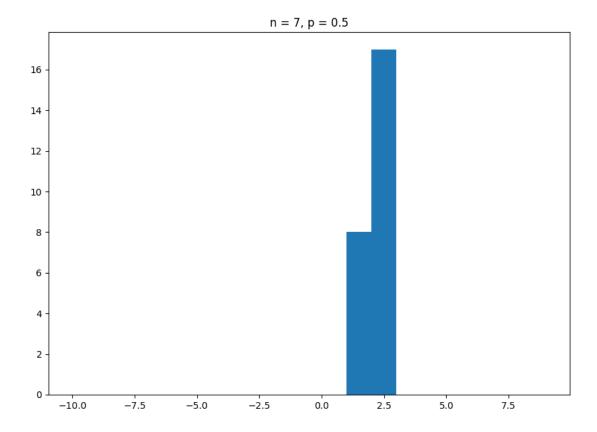
return pos #return final position</pre>
```

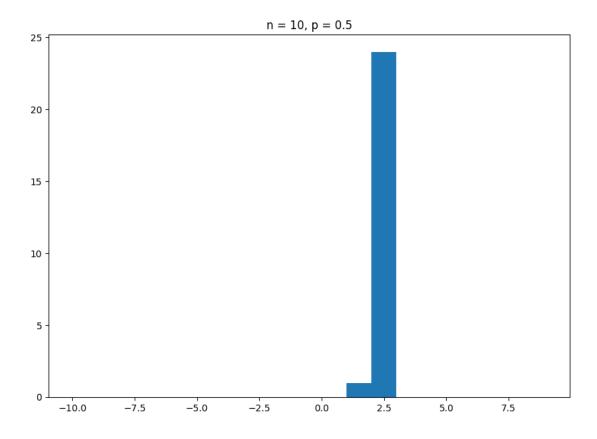
Calling the function for several iterations to get multiple expected values:

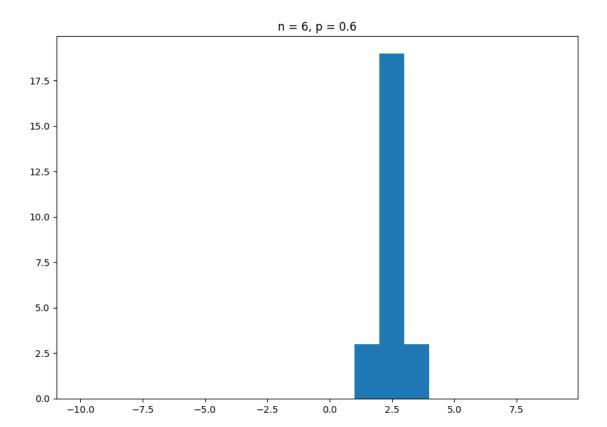
```
''', Calculating expected outcomes for various combinations of n and p''',
       expected = []
4
       outcomes = []
      p = 0.5
5
       while p \le 0.9: #for values of p from 0.4 to 0.9
           for n in range(5,11): #for values of n from 5 to 10
7
               for j in range(25): #expected value for each (n,p) for 25 iterations
8
                   for i in range (25):
                        outcomes.append(get_updated_position_restricted(n,p))
10
                    expected.append(sum(outcomes)/25) #appending the expected (average) value for
11
      each(n,p)
                    outcomes = [] #resetting outcomes list
12
               #plotting and showing a histogram of calculated expected values
13
               fig, ax = plt.subplots(figsize =(10, 7))
14
               ax.hist(expected, bins = range(-10,10))
plt.title('n = '+str(n)+', p = '+str(p))
1.5
16
               plt.savefig("Q1_histograms/Q1.2 "+'n = '+str(n)+', p = '+str(p)+'.png')
17
18
               plt.show()
               expected = [] #reset expected list
19
           p = round((p+0.1),1) #incrementing
20
```

Histograms produced by the above code for various combinations of n and p:









1.3

Function implementation in Python:

```
def stepsToMeet(pos1,pos2,p1,p2):
      count = 0 #keeps count of number of steps taken for objects to meet
3
      while pos1 != pos2:
          rand = random.randint(1,100) #generating a random number in the range 1 to 100 to
4
      determine outcome
          if rand < p1*100:</pre>
              pos1 += 1 #move one step right
6
          else:
              pos1 -= 1 #move one step left
          rand = random.randint(1,100) #generating a random number in the range 1 to 100
9
          if rand < p2*100:</pre>
              pos2 += 1 #move one step right
11
           else:
              pos2 -= 1 #move one step left
          count += 1
14
      return count
```

Calling the function for several iterations to get multiple expected values:

```
def main 13():
       ''', Calculating expected outcomes for various combinations of n & p''',
      expected = []
3
4
      outcomes = []
      p1 = 0.5
      p2 = 0.5
6
      pos1 = -5
      pos2 = 6
      while p1 <= 0.9:</pre>
9
          while p2 <= 0.9:</pre>
              for i in range(25): #calculating the expected value for each (n,p) for 25
      iterations
                   for j in range(25):
                       outcomes.append(stepsToMeet(pos1,pos2,p1,p2))
                   #calculating the average expected value for each(n,p)
14
                   expected.append(sum(outcomes)/25) #appending the expected (average) value for
      each(n,p)
                   outcomes = [] #resetting outcomes list
16
               #plotting a histogram of calculated expected values
17
               fig, ax = plt.subplots(figsize =(10, 7))
18
19
               ax.hist(expected, bins = range(-10,10))
               plt.title('p1 = '+str(p1)+', p2 = '+str(p2)+', pos1 = '+str(pos1)+', pos2 = '+str(
20
      pos2))
21
               expected = [] #reset expected list
               p2 = round((p2+0.1),1) #incrementing
23
           p1 = round((p1+0.1), 1) #incrementing
24
```

Q3. Picking a Random Point Correctly

3.1

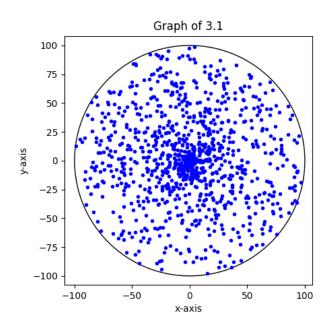
For part 3.1, we followed the method specified in the question. The program takes radius as an input from the user, and passes it to $gen\ points(R)$ to display the graph.

- $polar_to_cart(r,t)$: This function takes the polar coordinates $(rand\theta)$ and converts them to their cartesian equivalent. It is called from $polar_point(R)$, to return the cartesian coordinates.
- $polar_point(R)$: The function is responsible for generating a random point in polar coordinates, and returns the converted cartesian coordinates.
- $gen_points(R)$: This is the main function which runs 1000 iterations, and calls $random_point(R)$, stores the result in the separate lists for X and Y coordinates. Then the lists are passed to the plot functions, to graph the points.

```
import numpy as np
import matplotlib.pyplot as plt
import math as m
```

```
def random_point(radius):
5
      #R=radius, theta = angle
      R = np.random.uniform(0, radius)
                                           #random R in range 0-R
9
      theta = np.random.uniform(0, 360)*(m.pi/180) #random theta between 0-360 degrees
      return(polar_to_cart(R,theta))
10
11
  def polar_to_cart(r,t):
12
      x = r * np.cos(t) #x = rcos(theta)
14
      y = r * np.sin(t) #y = rsin(theta)
15
      return((x,y))
16
17
  def gen_points(l_radius):
18
19
      I = 1000
                    #iterations
20
      X = []
Y = []
                 #list for x axis
21
22
                 #list for y axis
23
      \hbox{\tt\#generating the lists for X and Y}
24
25
      for p in range(I):
           point = random_point(l_radius)
26
27
           X.append(point[0])
28
           Y.append(point[1])
29
30
      print("The variance on the x-axis is " + str(np.var(X)))
                                                                       #variance of x
31
      #Drawing the circle
32
33
      fig, axes = plt.subplots()
      circle = plt.Circle((0,0),l_radius,Fill=False)
34
35
      axes.set_aspect(1)
      axes.add_artist(circle)
36
      plt.title("circle")
37
38
      #Plotting X and Y
39
      plt.plot(X,Y,'.',color="blue")
40
      plt.xlabel("x-axis")
41
      plt.ylabel("y-axis")
42
      plt.title("Graph of 3.1")
43
      plt.show()
45
46 #Testing
47 R = input("Enter the radius: ")
48 gen_points(int(R))
```

Figure 1: radius = 100, variance of x = 1626.2176657086331



In part 3.2, we calculated the cartesian coordinates directly from the specified radius R. If the random point lied farther from the distance from the origin, it will be discarded and the program will find a new point.

- $random_point(R)$: This function finds random points for x and y coordinates between the range -R and R. It returns the tuple with cartesian coordinates.
- dist from origin(x,y): The function find and return the distance of the point from the origin.
- $gen_points(R)$: The main function iterates 1000 times and calls $random_point(R)$. Then it checks the condition that if the distance of the point (x, y) is within the radius R, it would append the points to their respective lists, else the counter will be decremented by 1 and the point will not be added to the list. Then the lists X and Y are passed to the plot functions for display.

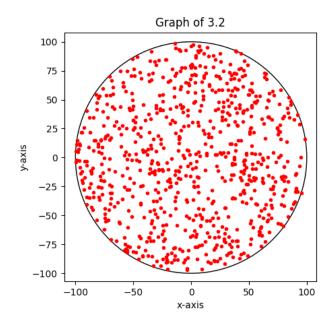
```
import matplotlib.pyplot as plt
2 import numpy as np
3 import math as m
  def random_point(lmt_rad):
      x = np.random.uniform(-lmt_rad,lmt_rad) #random point x
      y = np.random.uniform(-lmt_rad,lmt_rad) #random point y
       return((x,y))
9
  def dist_from_origin(x,y):
11
      return(m.sqrt(x**2+y**2))
12
  def gen_points(l_r):
14
15
16
      X = []
                    #list for x-axis
      Y = []
                    #list for y-axis
17
      I = 1000
                    #iterations
18
19
      for x in range(I):
20
          point = random_point(1_r)
21
          if dist_from_origin(point[0],point[1]) <= 1_r: #if distance is smaller than or equal
22
               X.append(point[0])
23
24
               Y.append(point[1])
                                #if distance is greater than R
           else:
25
26
               x = x-1
27
      print("The variance on the x-axis is " + str(np.var(X)))
                                                                      #variance of X
28
29
      #Drawing the circle
30
      fig, axes = plt.subplots()
31
      circle = plt.Circle((0,0),l_r,Fill=False)
32
      axes.set_aspect(1)
33
34
      axes.add_artist(circle)
      plt.title("circle")
35
36
37
      #Plotting the points
      plt.plot(X,Y,'.',color="red")
38
      plt.xlabel("x-axis")
39
      plt.ylabel("y-axis")
40
      plt.title("Graph of 3.2")
41
42
      plt.show()
43
44 #Testing
45 R = input("Enter the radius: ")
46 gen_points(int(R))
```

3.3

In part 3.3, we had to modify the function for polar coordinates, such that the graph it made was uniformly distributed, similar to part 3.2.

- $random_point(R)$:
- \bullet $polar_to_cart(cord)$:

Figure 2: radius = 100, variance of x = 2533.130556271337



• $gen_points(R)$:

```
import numpy as np
2 import matplotlib.pyplot as plt
3 import math as m
5 def random_point(radius):
     #R=radius, theta=angle
     theta = np.random.uniform(0, 360)*(m.pi/180) #random theta between 0-360 degrees
9
10
     return(polar_to_cart(R,theta))
  def polar_to_cart(r,t):
12
     x = r * np.cos(t) #x = rcos(theta)
14
     y = r * np.sin(t) #y = rsin(theta)
15
     return((x,y))
16
17
  def gen_points(l_radius):
18
19
20
     I = 1000
                 #iterations
              #list for x axis
     X = []
21
22
              #list for y axis
23
     \#generating the lists for X and Y
24
25
     for p in range(I):
         point = random_point(l_radius)
26
         X.append(point[0])
27
28
         Y.append(point[1])
29
     30
31
     #Drawing the circle
32
33
     #Outer Circle
34
     fig, axes = plt.subplots()
     circle1 = plt.Circle((0,0),2,Fill=False,color="red")
35
36
     axes.set_aspect(1)
     axes.add_artist(circle1)
37
38
     #inner circle
39
     circle2 = plt.Circle((0,0),1,Fill=False,color="green")
40
     axes.set_aspect(1)
41
42
     axes.add_artist(circle2)
```

```
fig.legend([circle1,circle2],["outer circle", "inner circle"])
43
44
        #Plotting X and Y
45
        plt.plot(X,Y,'.',color="blue")
plt.xlabel("x-axis")
plt.ylabel("y-axis")
46
47
48
        plt.title("Graph of 3.3")
49
        plt.show()
50
51
52 #Testing
x = gen_points(2)
```

Figure 3: radius = 2, variance of x = 1.017564829759483

