

Two-Dimensional Random Walks

MATH 242 Modern Computational Math

Today we will start to understand properties of two-dimensional (2D) random walks.

Each position of a 2D random walk is an ordered pair of integers. Thus, a list of positions of a 2D random walk can be stored as a $N \times 2$ matrix, where N is the number of steps in the random walk.

Before we generate 2D random walks, we should practice using matrices in Sage.

```
In [21]: mat = matrix( [[1,2,3], [4,5,6], [7,8,9]] )  
          print(mat)
```

```
Out[21]: [1 2 3]  
          [4 5 6]  
          [7 8 9]
```

```
In [5]: mat[:2]
```

```
Out[5]: [1 2 3]  
          [4 5 6]
```

```
In [22]: mat[:,0]
```

```
Out[22]: [1]  
          [4]  
          [7]
```

```
In [9]: print( mat.dimensions() )  
          print( mat.nrows() )  
          print( mat.ncols() )
```

```
Out[9]: (3, 3)  
          3  
          3
```

```
In [10]: type( mat )
```

```
Out[10]: <class  
         'sage.matrix.matrix_integer_dense.Matrix_integer_dense'>
```

```
In [11]: zero_matrix(5, 3)
```

```
Out[11]: [0 0 0]  
          [0 0 0]  
          [0 0 0]  
          [0 0 0]  
          [0 0 0]
```

Now we can generate a 2D random walk.

```
In [14]: # define the possible moves at each step of the random walk  
         dirs = matrix( [[0,1],[0,-1],[1,0],[-1,0]] )  
  
         # define the number of steps to take  
         numSteps = 50  
  
         # set up a numpy 2-D array to store the locations visited.  
         locations = zero_matrix(numSteps, 2)  
  
         # take steps  
         for i in range(1, numSteps):  
             r = randrange(4) # random value 0, 1, 2, or 3  
             move = dirs[r]   # direction to move  
             locations[i] = locations[i-1] + move  
  
         # output the random walk  
         print(locations)
```

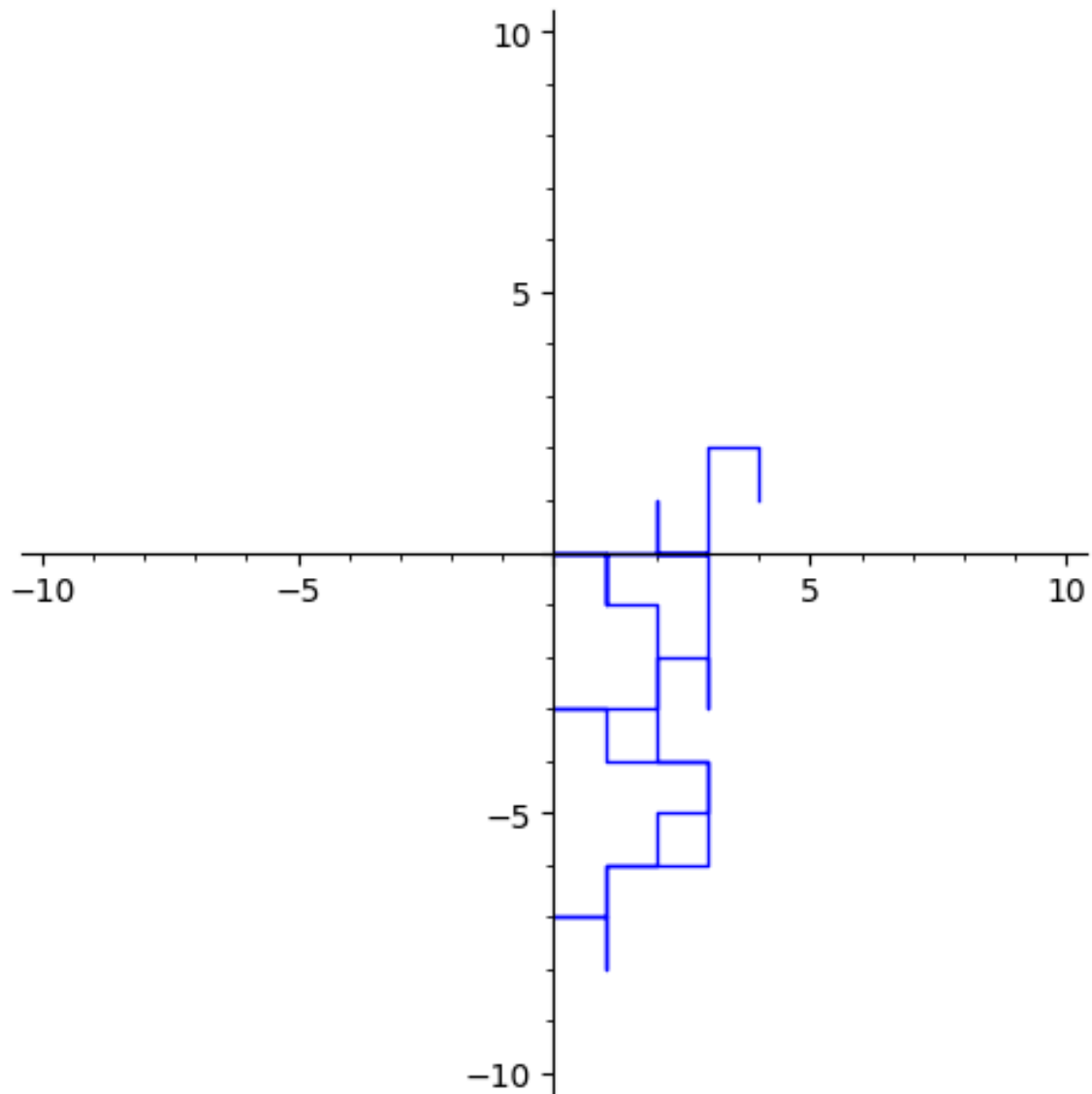
```
Out[14]: [ 0  0]  
          [ 1  0]  
          [ 1 -1]  
          [ 1  0]  
          [ 0  0]  
          [ 1  0]  
          [ 1 -1]  
          [ 2 -1]  
          [ 2 -2]  
          [ 2 -3]  
          [ 2 -4]  
          [ 3 -4]  
          [ 3 -5]  
          [ 2 -5]  
          [ 2 -6]  
          [ 1 -6]  
          [ 1 -7]
```

```
[ 1 -8]
[ 1 -7]
[ 0 -7]
[ 1 -7]
[ 1 -6]
[ 2 -6]
[ 3 -6]
[ 3 -5]
[ 3 -4]
[ 2 -4]
[ 1 -4]
[ 1 -3]
[ 0 -3]
[ 1 -3]
[ 2 -3]
[ 2 -2]
[ 3 -2]
[ 3 -3]
[ 3 -2]
[ 3 -1]
[ 3  0]
[ 2  0]
[ 1  0]
[ 2  0]
[ 3  0]
[ 2  0]
[ 2  1]
[ 2  0]
[ 3  0]
[ 3  1]
[ 3  2]
[ 4  2]
[ 4  1]
```

We can also plot the random walk. Here, we will make a 2D plot.

```
In [15]: list_plot(locations, plotjoined=True, xmin=-10, xmax=10,
                ymin=-10, ymax=10, figsize=[5,5])
```

Out[15]:



Now let's put our 2D random walk code into a function, so we can easily generate lots of 2D random walks.

```
In [1]: def randomWalk2D(numSteps):
# define the possible moves at each step of the random
walk
    dirs = matrix( [[0,1],[0,-1],[1,0],[-1,0]] )

    # set up a numpy 2-D array to store the locations
    visited.
    locations = zero_matrix(numSteps, 2)

    # take steps
    for i in range(1, numSteps):
        r = randrange(4) # random value 0, 1, 2, or 3
        move = dirs[r] # direction to move
        locations[i] = locations[i-1] + move
```

```
# output the random walk
return locations
```

```
In [20]: print(randomWalk2D(50))
```

```
Out[20]: [[ 0  0]
[ 1  0]
[ 0  0]
[ 1  0]
[ 0  0]
[ 0 -1]
[-1 -1]
[-1  0]
[ 0  0]
[ 0 -1]
[ 0 -2]
[ 0 -3]
[ 0 -4]
[-1 -4]
[-1 -5]
[-1 -6]
[-1 -7]
[ 0 -7]
[ 0 -8]
[ 0 -9]
[ 0 -8]
[-1 -8]
[ 0 -8]
[ 1 -8]
[ 1 -9]
[ 2 -9]
[ 2 -10]
[ 1 -10]
[ 0 -10]
[ 0 -9]
[ 1 -9]
[ 1 -8]
[ 1 -9]
[ 0 -9]
[ 0 -10]
[ 1 -10]
[ 1 -9]
[ 0 -9]
[ 1 -9]
[ 0 -9]
[ 0 -8]
[ 0 -9]
[ 1 -9]
[ 2 -9]
[ 1 -9]]
```

```
[ 2 -9]
[ 2 -8]
[ 2 -7]
[ 2 -6]
[ 2 -5]
```

Diameter of a 2D Random Walk

Define the *width* of a 2D random walk to be the maximum difference between any two x -coordinates attained by the walk. Similarly, define the *height* of the walk to be the maximum difference between any two y -coordinates. Define the *diameter* of the random walk to be the maximum of its width and height.

What is the average diameter of a 2D random walk after N steps? How does this depend on N ?

First, a function that takes a 2D random walk (as a list of locations) and returns its diameter.

```
In [2]: # function that takes a 2D random walk (as a list of
         locations) and returns its diameter
         def diameter2D(walk):
             xcoords = walk[:,0]
             width = max(xcoords) - min(xcoords)

             ycoords = walk[:,1]
             height = max(ycoords) - min(ycoords)

             return max(width[0], height[0])
```

Testing:

```
In [3]: aWalk = randomWalk2D(30)
         print(aWalk)
         diameter2D(aWalk)
```

```
Out[3]: [ 0  0]
         [ 1  0]
         [ 0  0]
         [-1  0]
         [ 0  0]
         [ 0 -1]
         [ 0  0]
         [ 0 -1]
```

```
[ 0 -2]
[-1 -2]
[-1 -3]
[-2 -3]
[-2 -2]
[-2 -1]
[-3 -1]
[-3  0]
[-4  0]
[-3  0]
[-3  1]
[-2  1]
[-2  0]
[-2  1]
[-1  1]
[-1  0]
[ 0  0]
[ 0 -1]
[-1 -1]
[-1  0]
[-1 -1]
[ 0 -1]
```

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Function that computes the average diameter of many random walks, all with the same number of steps:

```
In [4]: # function that computes the diameter of a specified number
        # of random walks, all with the same number of steps
        def avgDiameter2D(numSteps, numWalks):
            diams = [diameter2D(randomWalk2D(numSteps)) for _ in
                     range(numWalks)]
            return sum(diams)/numWalks
```

Now compute the average diameter of 2D random walks for a selection of numbers of steps.

```
In [5]: nvals = range(10,500,10)
        avgDiams = [avgDiameter2D(n,100) for n in nvals]
        print(avgDiams)
```

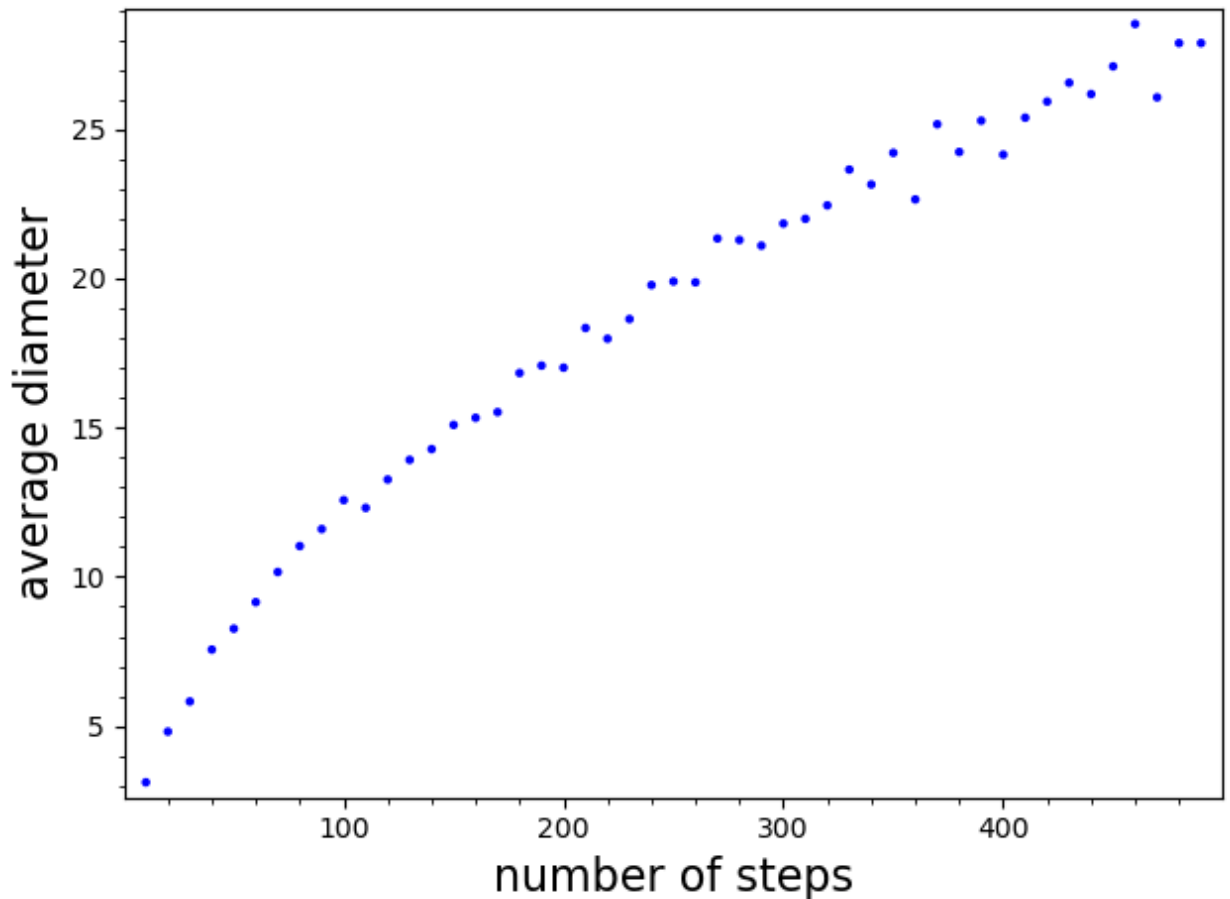
```
Out[5]: [311/100, 481/100, 291/50, 189/25, 413/50, 183/20, 254/25,
1103/100, 58/5, 1257/100, 1231/100, 663/50, 348/25, 357/25,
1509/100, 1533/100, 388/25, 1683/100, 427/25, 1701/100,
917/50, 899/50, 466/25, 989/50, 199/10, 1987/100, 1067/50,
2129/100, 211/10, 546/25, 22, 449/20, 473/20, 463/20,
```

```
2421/100, 453/20, 2517/100, 606/25, 2529/100, 483/20,  
2539/100, 2593/100, 664/25, 1309/50, 2711/100, 2853/100,  
2607/100, 2789/100, 2789/100]
```

Make a plot of the average diameters we just computed:

```
In [7]: list_plot( list(zip(nvals, avgDiams)), axes_labels=["number  
of steps", "average diameter"], frame=True)
```

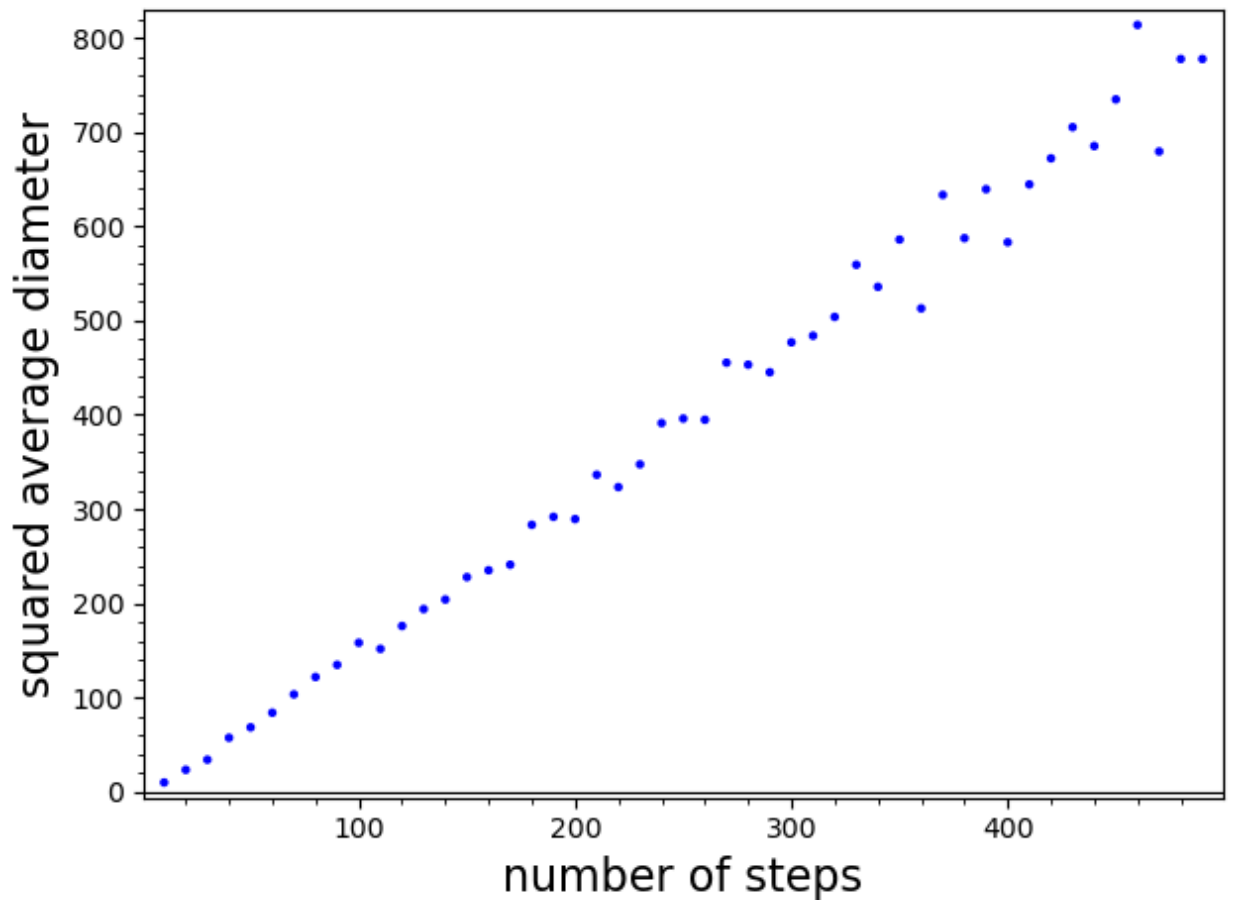
Out[7]:



The previous plot looks like it might have a square-root shape, so let's plot the average diameters squared. Since this plot is nearly linear, the previous plot is proportional to a square root function.

```
In [9]: sqDiams = [d^2 for d in avgDiams]  
list_plot( list(zip(nvals, sqDiams)), axes_labels=["number  
of steps", "squared average diameter"], frame=True)
```

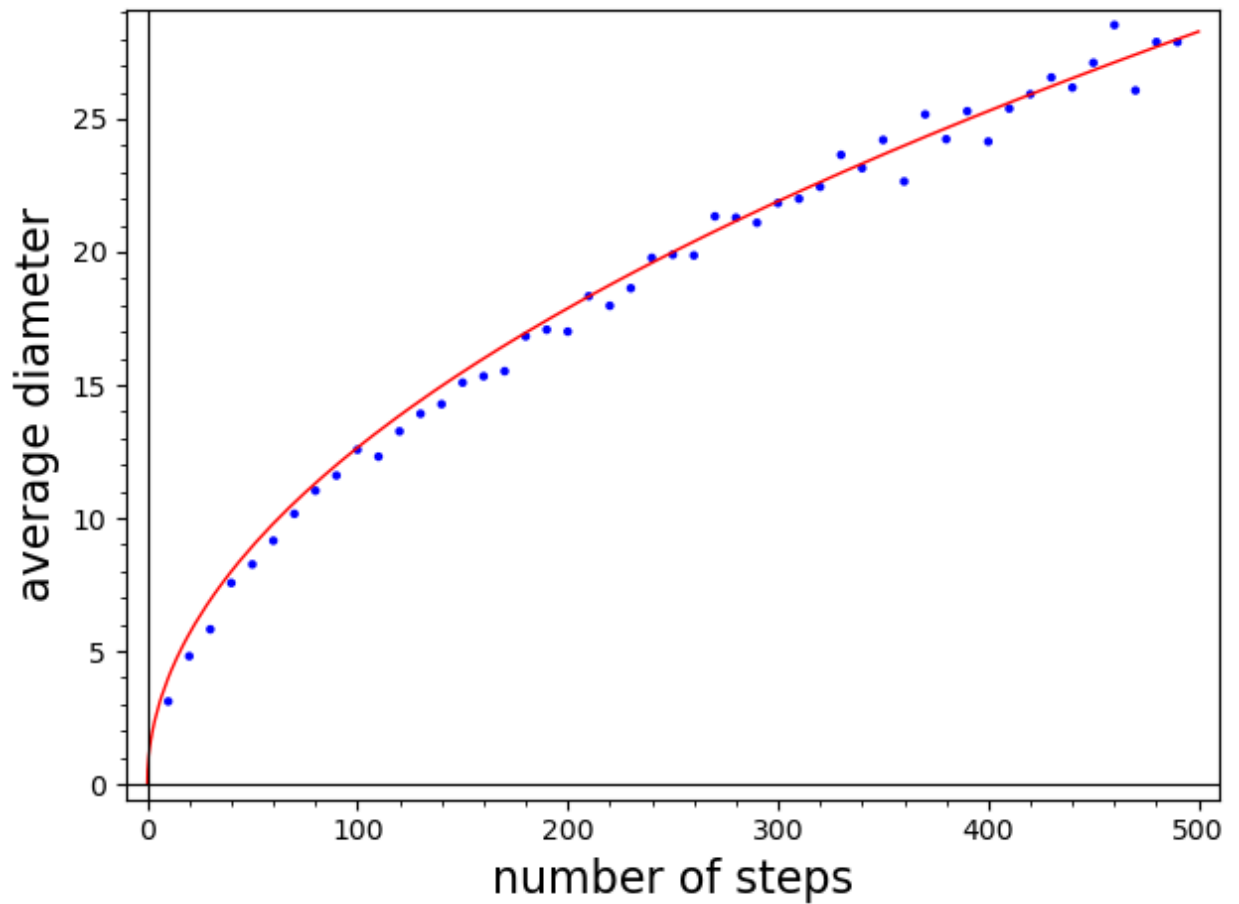

Out[9]:



Since the slope of the linear plot is approximately 1.6, we conclude that the average diameter after n steps is approximately $\sqrt{1.6n}$. The following plot shows the average diameters along with this square root function.

```
In [8]: list_plot( list(zip(nvals, avgDiams)), axes_labels=["number  
of steps", "average diameter"], frame=True) +  
plot(sqrt(1.6*x), (x, 0, 500), color="red")
```

Out[8]:



In [0]:

Four Quadrants

How likely is it that a 2D random walk visits all four quadrants of the plane within N steps? How does this depend on N ?

In [0]:

In [0]:

In [0]:

Distinct Points

How many distinct points does a 2D random walk visit, on average, in N steps? How does this depend on N ?

In [0]:

In [0]:

In [0]: