

Simple models of diffusion - Random walk

Exercise Report: 2D Random Walk

(Whole code at the end of the report)

Introduction

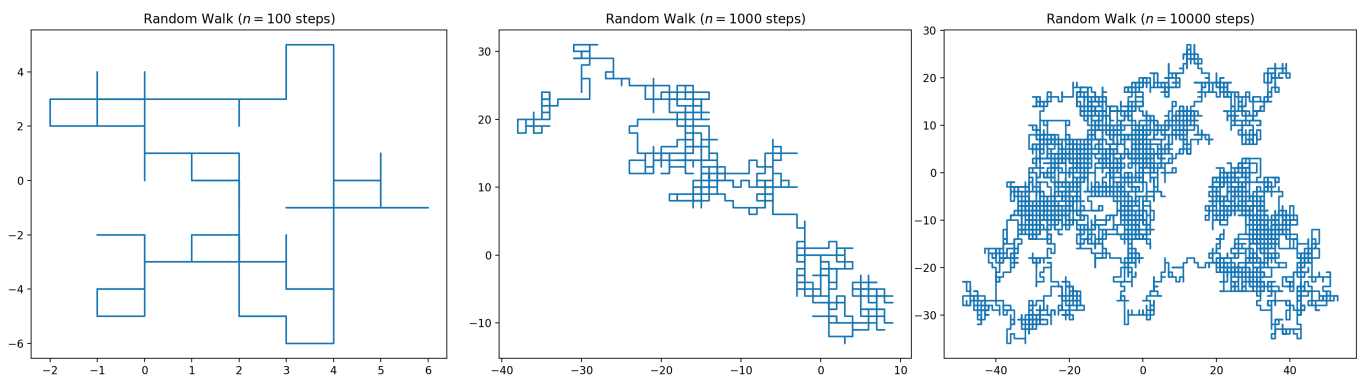
We have drunk sailer that starts on $x = (0,0)$ position and randomly choose to what 2D direction he is going to go, can take steps like $(0,1)$, $(0,-1)$, $(1,0)$, $(-1,0)$. He take steps of length = 1 and he loses all memory between any single steps.

We can simulate this situation by provided code:

```
def random_walk(N, K):  
    directions = np.array([[1, 0], [-1, 0], [0, 1], [0, -1]])  
    walks = np.zeros((K, N, 2))  
    for k in range(K):  
        steps = directions[np.random.randint(0, 4, size=N-1)]  
        walks[k, 1:] = steps.cumsum(axis=0)  
  
    distances = np.linalg.norm(walks[:, -1, :], axis=1)  
    return walks, distances
```

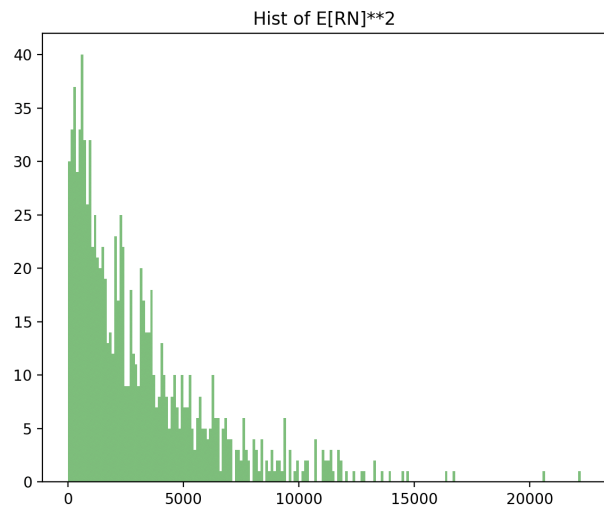
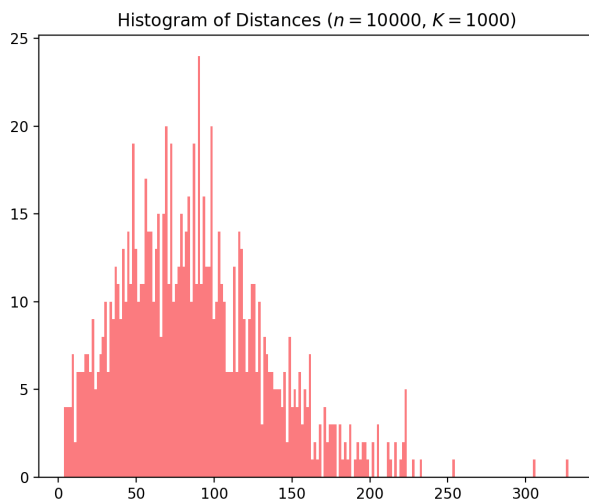
``(Python 3.11.7)

This are the result for one drunkard:



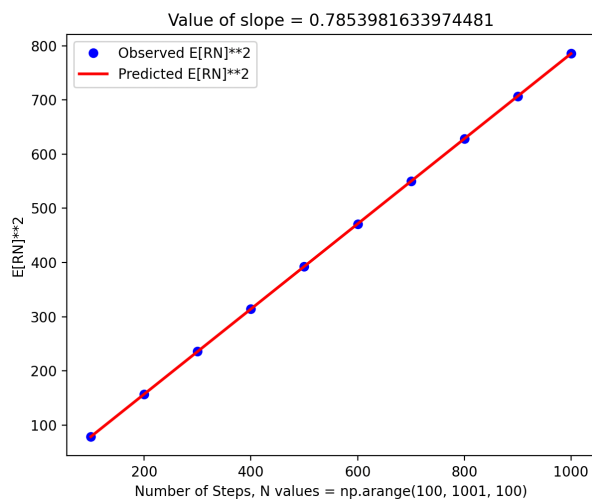
1) 2D random $N = 100$, $N = 1000$, $N = 10\,000$

We can simulate this for K drunkards and plot $\mathbb{E}[R_N]$ where $\mathbb{E}[R_N] = \sqrt{x_N^2 + y_N^2}$, and then we perform $(R_N)^2$ and plotting the histogram with following results:



3) Histogram of $\mathbb{E}[R_N]$ and $\mathbb{E}[R_N]^2$

The linear regression for $\mathbb{E}[R_N]^2$ to N steps:



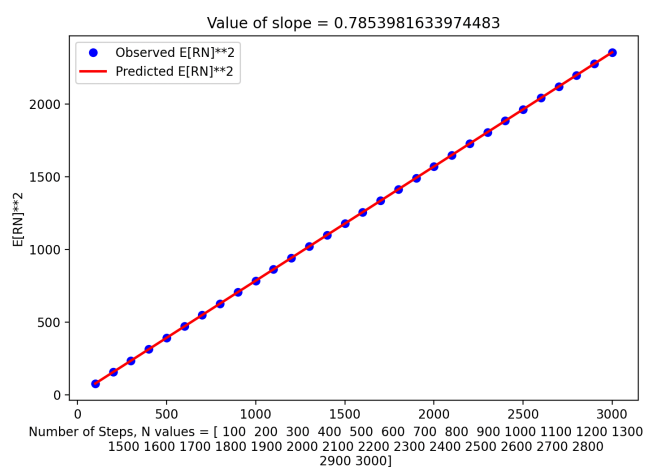
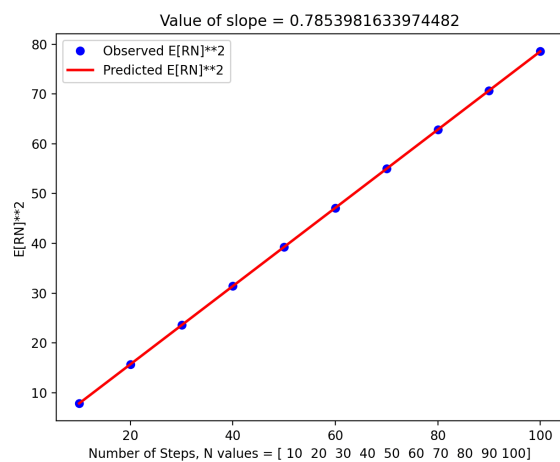
Linear regression for:

$$\mathbb{E}[R_N]^2 = \frac{\pi}{4} \times N$$

$$m \approx 0.78 \quad (\text{Value of slope})$$

4) Linear regression for $N = \{100, 200, \dots, 1000\}$

And some more examples for bigger N_{step} values:



4) Plot for $N_{\text{values}} = \{10, 20, \dots, 100\}$ and $N_{\text{values}} = \{100, 200, \dots, 3000\}$

```(WHOLE CODE)

```
import numpy as np
import matplotlib.pyplot as plt
import random
import math
from sklearn.linear_model import LinearRegression
from scipy.stats import linregress

def random_walk(N, K):
 """
 Simulates a random walk in 2D for N steps and K walkers.
 Returns the final positions and the Euclidean distances from the origin.
 """
 directions = np.array([[1, 0], [-1, 0], [0, 1], [0, -1]])
 walks = np.zeros((K, N, 2))
 for k in range(K):
 steps = directions[np.random.randint(0, 4, size=N-1)]
 walks[k, 1:] = steps.cumsum(axis=0)

 distances = np.linalg.norm(walks[:, -1, :], axis=1)
 return walks, distances

def calculate_avg_squared_dist(N_values):
 """
 Calculates the average squared distance for a set of steps.
 Theoretical calculation using $(\pi/4) * N$.
 """
 return [(math.pi / 4) * N for N in N_values]

def main():
 # 36 GB RAM
 # N = 30000 # Number of steps
 # K = 100000 # Number of Drunkards

 N = 3000
 K = 1000

 fig, ax = plt.subplots(2, 2, figsize=(12, 10))

 # Perform random walk and plot the first one for visualization
 walks, distances = random_walk(N, K)
 ax[0, 0].plot(walks[0, :, 0], walks[0, :, 1])
 ax[0, 0].set_title("Random Walk ($n = {}$ steps)".format(N))

 # Histogram of distances
 ax[0, 1].hist(distances, bins=200, color='red', alpha=0.5)
 ax[0, 1].set_title(
 "Histogram of Distances ($n = {}$, $K = {}$)".format(N, K))

 # Prepare data for linear regression
 N_values = np.arange(100, 3001, 100)
 avg_squared_dists = calculate_avg_squared_dist(N_values)

 ax[1, 0].set_title("Hist of $E[RN]**2$ ")
 ax[1, 0].hist(np.power(distances, 2), bins=200,
 color='g', alpha=0.5, label='X')

 # Fit linear regression
 model = LinearRegression()
 model.fit(N_values.reshape(-1, 1), avg_squared_dists)
 predictions = model.predict(N_values.reshape(-1, 1))

 slope = linregress(N_values, avg_squared_dists).slope

 # Plotting the linear regression results
 slope, _, _, _ = linregress(N_values, avg_squared_dists)
 ax[1, 1].plot(N_values, avg_squared_dists, 'bo', label="Observed $E[RN]**2$ ")
 ax[1, 1].plot(N_values, predictions, 'r-',
 linewidth=2, label="Predicted $E[RN]**2$ ")
 ax[1, 1].set_title(
 f"Value of slope = {slope}")
 ax[1, 1].set_xlabel(
 f"Number of Steps, N values = {N_values}")
 ax[1, 1].set_ylabel('E[RN]**2')
 ax[1, 1].legend()

 plt.tight_layout()
 plt.show()

if __name__ == "__main__":
 main()
```

```Python 3.11.7)