

BM20BTECH11001-Lab15(Resubmission)

November 30, 2021

1 Question-a

- We assume $x(t)$ to be a 3×1 matrix with each row representing a coordinate in 3D space.
- If $x(t)$ is a gaussian random process, then all the rows of $x(t)$ are gaussian.
- The filters $y(t)$ and $z(t)$ are just a combination of $x(t)$

```
[1]: from mpl_toolkits.mplot3d import Axes3D
import matplotlib.pyplot as plt
import numpy as np
from pylab import *
import matplotlib.colors as colors
```

1.1 $x(t)$ filter

1.1.1 Cross-covariance matrix of $x(t)$ filter

```
[2]: A = [[0.5,0,0,0,0], [0.5, 0.5, 0, 0, 0], [0, 0.5, 0.5, 0, 0], [0, 0, 0.5, 0.5, 0],
→0], [0, 0, 0, 0.5, 0.5]]

matrix = np.dot(A, np.transpose(A))

print(matrix)
```

```
[[0.25 0.25 0.    0.    0.   ]
 [0.25 0.5   0.25 0.    0.   ]
 [0.    0.25 0.5   0.25 0.   ]
 [0.    0.    0.25 0.5   0.25]
 [0.    0.    0.    0.25 0.5  ]]
```

1.1.2 Plotting $x(t)$

```
[3]: A = [[0.5, 0, 0], [0.5, 0.5, 0], [0, 0.5, 0.5]]
matrix = np.dot(A, np.transpose(A))
print(matrix)
mean = 0
Y = lambda x,y,z : np.transpose([[x-mean, y-mean, z-mean]])

#Inverse of covariance matrix
```

```

Inv = np.linalg.inv(matrix)
f_x = []

x = np.linspace(-0.4, 0.4, 20)
y = np.linspace(-0.4, 0.4, 20)
z = np.linspace(-0.4, 0.4, 20)
x_1 = []
y_1 = []
z_1 = []

for d in x:
    for e in y:
        for f in z:
            x_1.append(d)
            y_1.append(e)
            z_1.append(f)
            t = np.dot(np.transpose(Y(d,e,f)), Inv)
            a = np.dot(t, Y(d,e,f))
            gauss = ((np.e)**(-a[0][0]/2))/(((2*np.pi)**(3/2))*(np.linalg.
→det(np.array(matrix))))
            f_x.append(gauss)

my_cmap = plt.get_cmap('seismic')

fig = plt.figure(figsize=(15, 15))
ax = fig.add_subplot(111, projection='3d')

# creating the heatmap
img = ax.scatter3D(x_1, y_1, z_1, c=f_x, cmap=my_cmap, marker='^')
fig.colorbar(img, ax = ax, shrink = 0.5, aspect = 5, label="PDF values")
ax.set_xlabel('X-axis')
ax.set_ylabel('Y-axis')
ax.set_zlabel('Z-axis')
ax.set_title('x(t) filter application');

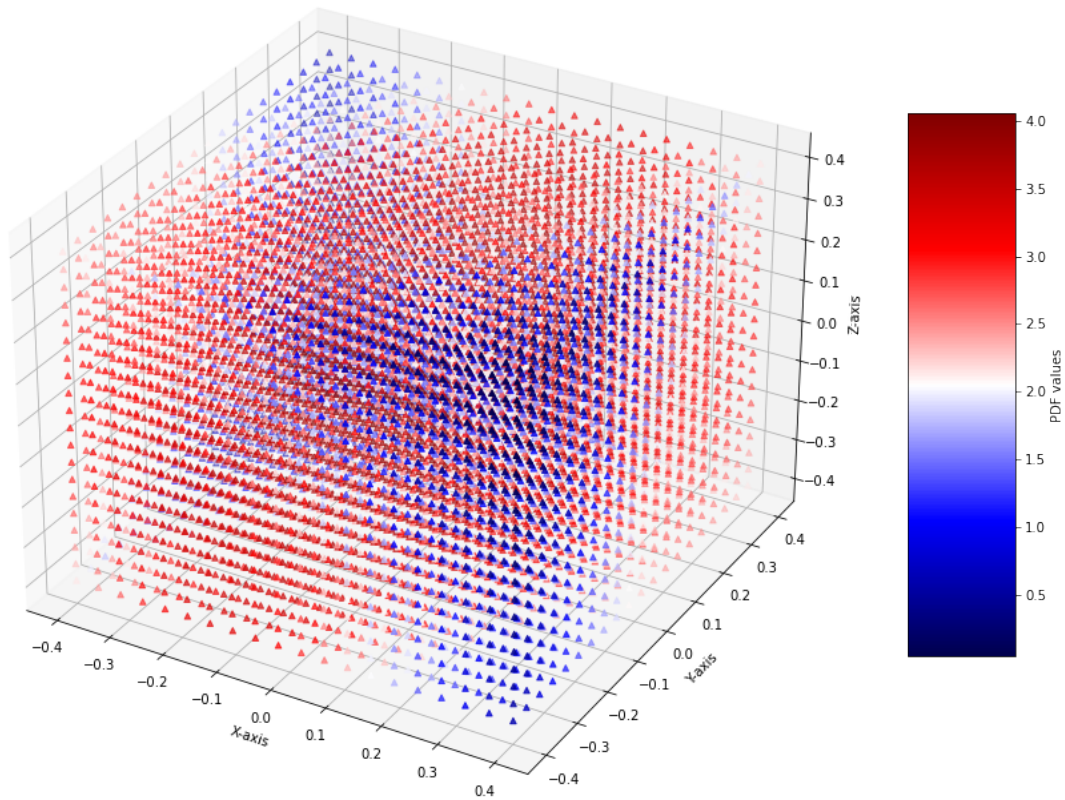
```

```

[[0.25 0.25 0. ]
 [0.25 0.5  0.25]
 [0.   0.25 0.5 ]]

```

x(t) filter application



1.1.3 Analysis of the graph

- The graph is oriented across $y=x$ direction and is elongated along z

1.2 y(t) filter

1.2.1 Cross-covariance matrix for y(t)

```
[4]: A = [[1,0,0,0,0], [-1, 1, 0, 0, 0], [0, -1, 1, 0, 0], [0, 0, -1, 1, 0], [0, 0, 0, -1, 1]]
      ↪0, -1, 1]]

matrix = np.dot(A, np.transpose(A))

print(matrix)
```

```
[[ 1 -1  0  0  0]
 [-1  2 -1  0  0]
 [ 0 -1  2 -1  0]
```

```
[ 0  0 -1  2 -1]
[ 0  0  0 -1  2]]
```

1.2.2 Plotting $y(t)$

```
[5]: A = [[1, 0, 0], [-1, 1, 0], [0, -1, 1]]
matrix = np.dot(A, np.transpose(A))
print(matrix)
Y = lambda x,y,z : np.transpose([[x-mean, y-mean, z-mean]])

#Inverse of covariance matrix
Inv = np.linalg.inv(matrix)
f_x = []

x = np.linspace(-0.5, 0.5, 20)
y = np.linspace(-0.5, 0.5, 20)
z = np.linspace(-0.5, 0.5, 20)
x_1 = []
y_1 = []
z_1 = []

for d in x:
    for e in y:
        for f in z:
            x_1.append(d)
            y_1.append(e)
            z_1.append(f)
            t = np.dot(np.transpose(Y(d,e,f)), Inv)
            a = np.dot(t, Y(d,e,f))
            gauss = ((np.e)**(-a[0][0]/2))/(((2*np.pi)**(3/2))*(np.linalg.
↪det(np.array(matrix))))
            f_x.append(gauss)

my_cmap = plt.get_cmap('seismic')

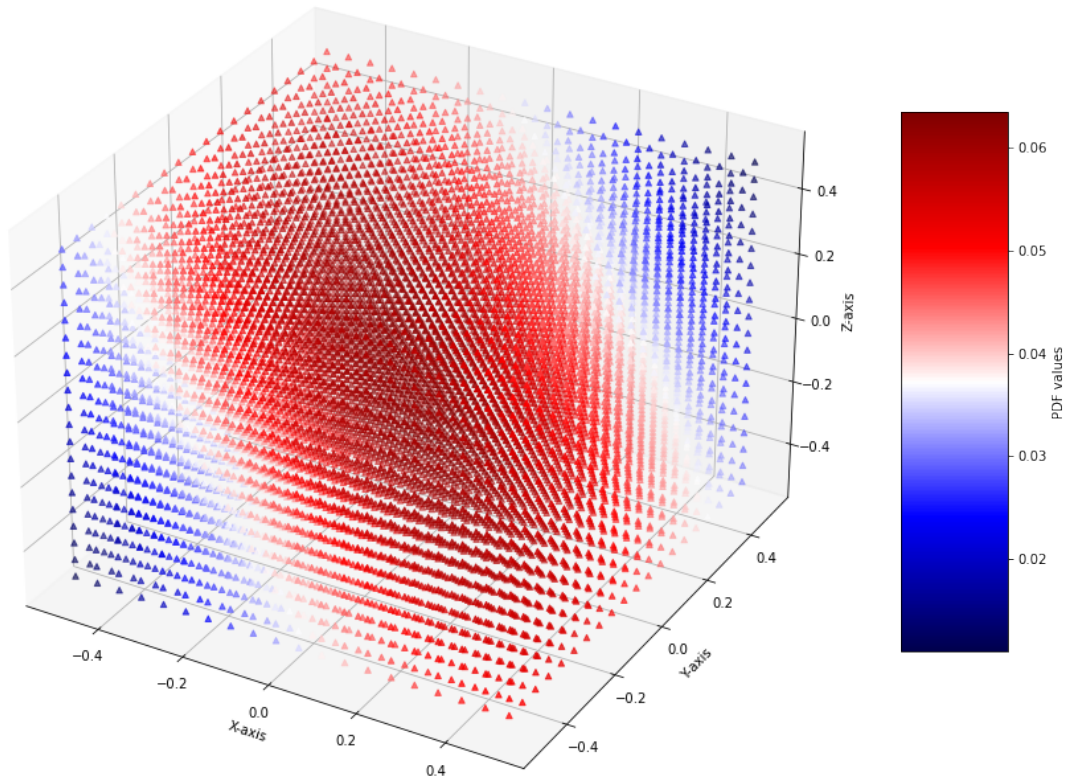
fig = plt.figure(figsize=(15, 15))
ax = fig.add_subplot(111, projection='3d')

# creating the heatmap
img = ax.scatter3D(x_1, y_1, z_1, c=f_x, cmap=my_cmap, marker='^')
fig.colorbar(img, ax = ax, shrink = 0.5, aspect = 5, label="PDF values")
ax.set_xlabel('X-axis')
ax.set_ylabel('Y-axis')
ax.set_zlabel('Z-axis')
ax.set_title('y(t) filter application');
```

```
[[ 1 -1  0]
```

```
[-1  2 -1]
[ 0 -1  2]]
```

y(t) filter application



1.2.3 Analysis of the graph

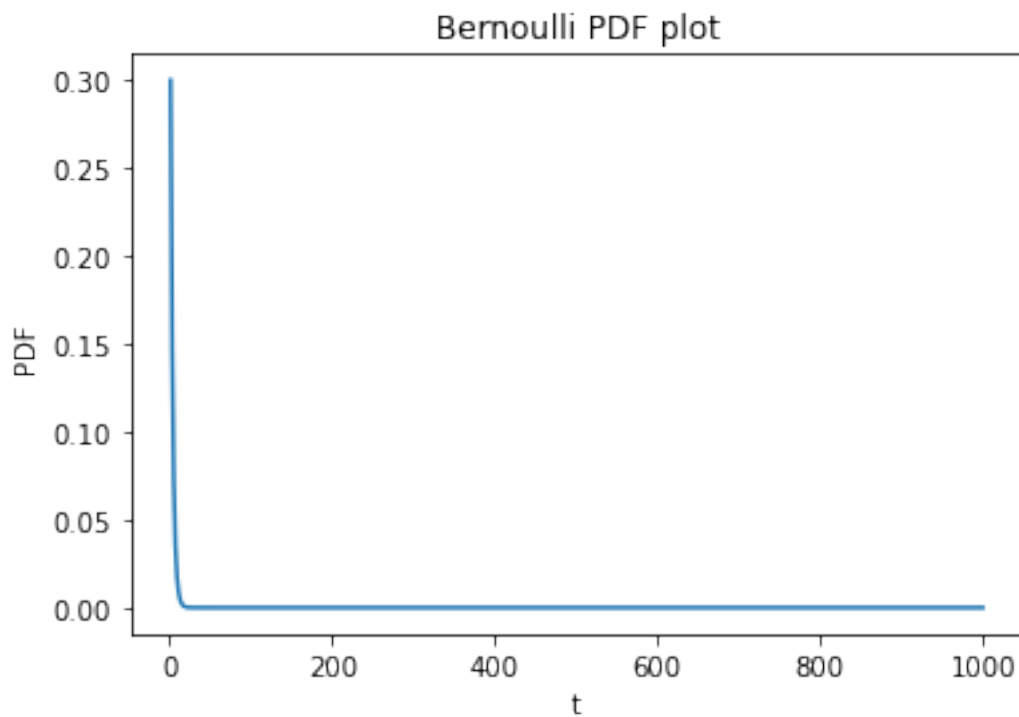
- The graph is oriented along $x+z = c$ and is elongated along y-axis.

1.3 Question-b

```
[6]: x = np.linspace(1, 1000, 1000)
y = []
p = 0.3
bernoulli = lambda m: ((1-p)**(m-1))*p
for a in x:
    y.append(bernoulli(a))
```

```
[7]: fig, ax = plt.subplots()
ax.plot(x, y);
ax.set_title("Bernoulli PDF plot")
```

```
ax.set_xlabel('t')
ax.set_ylabel('PDF');
```

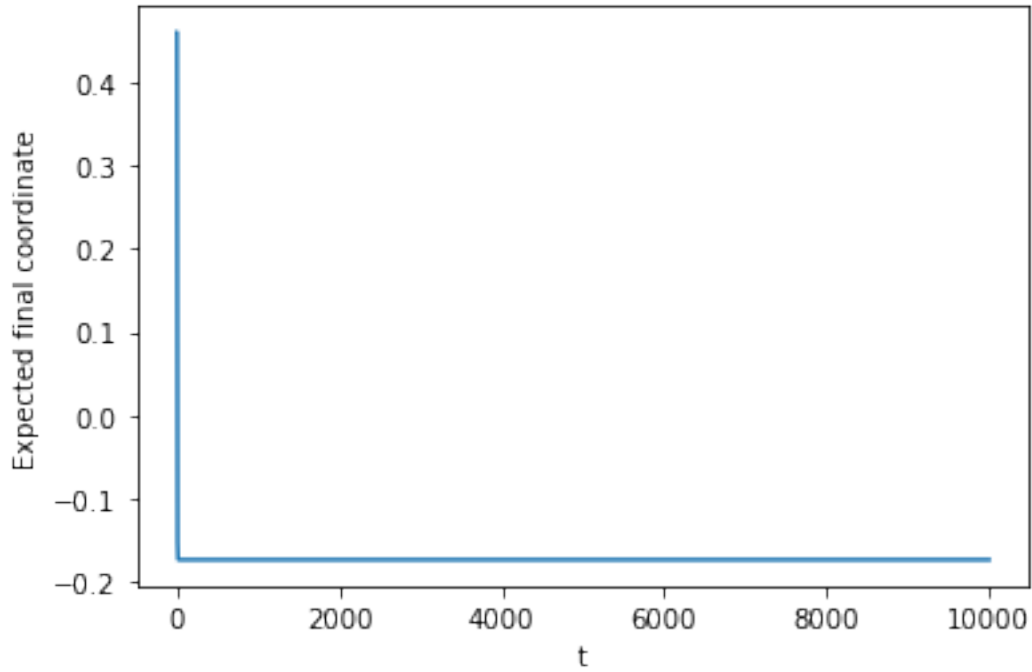


1.4 Question-c

- Let's assume that the particle can either move 1 step forward or 1 step backward(+1 for forward step and -1 for backward step)
- If the probability of getting +1 coordinate is p , the probability of getting +1 coordinate after m trials is $((1-p)^{(m-1)})p$ and the expected coordinate will be $((-1)(m-1)) + 1 = 2-m$

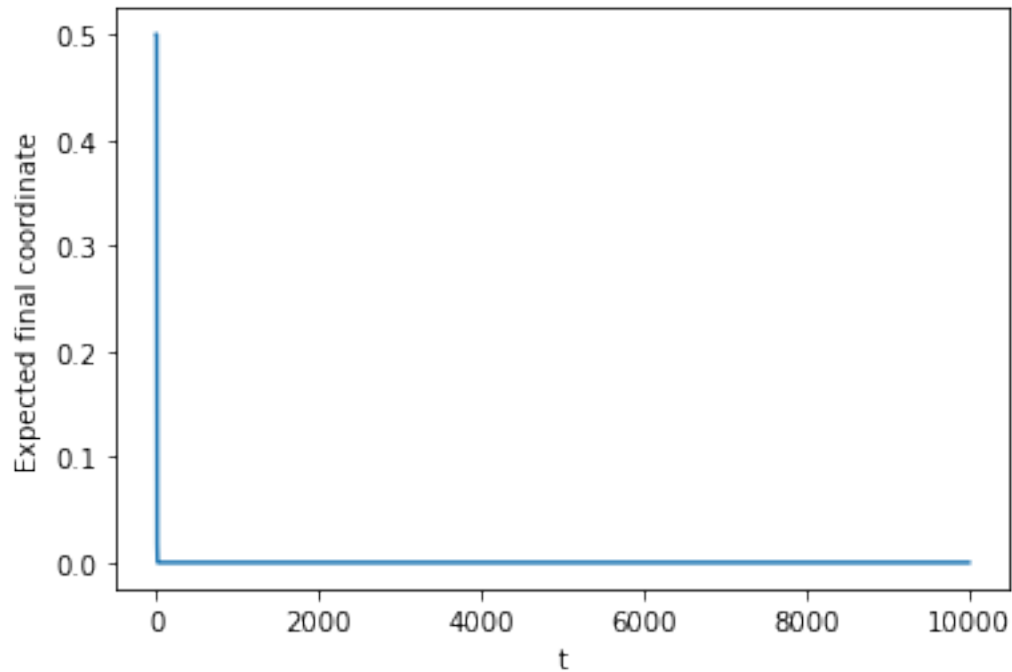
```
[8]: x = np.linspace(1, 10000, 10000)
x_t = []
p = 0.46
bernoulli = lambda m: ((1-p)**(m-1))*p
for a in x:
    x_t.append((2-a)*bernoulli(a))
y = []
sum_x = 0
for i in range(0, 10000):
    sum_x += x_t[i]
    y.append(sum_x)
```

```
[9]: fig, ax = plt.subplots()
      ax.plot(x, y);
      ax.set_xlabel('t')
      ax.set_ylabel('Expected final coordinate');
```



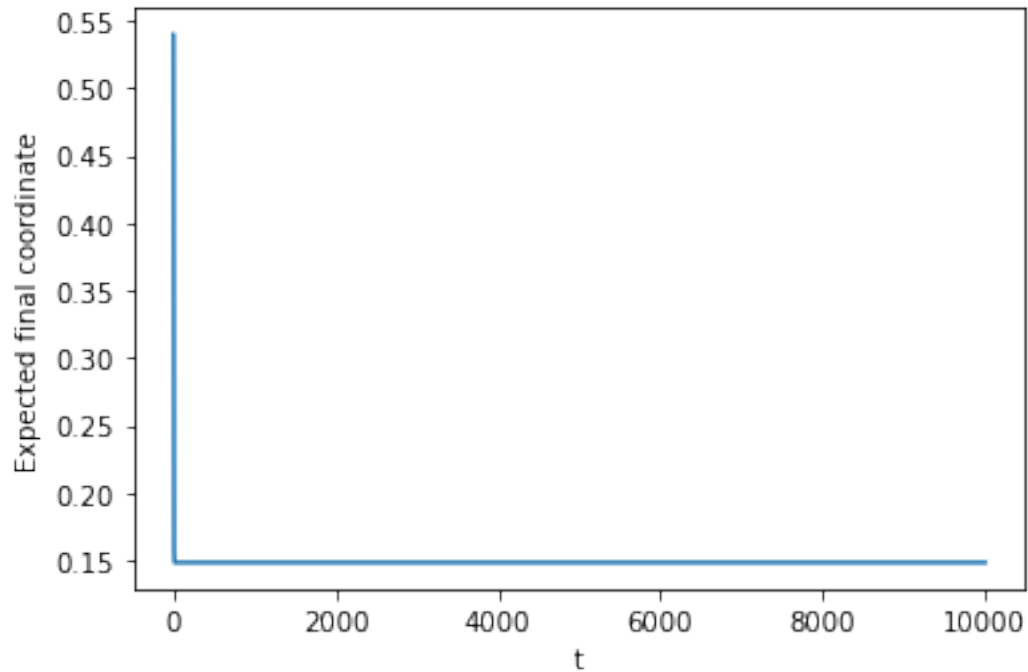
```
[10]: x = np.linspace(1, 10000, 10000)
      x_t = []
      p = 0.5
      bernoulli = lambda m: ((1-p)**(m-1))*p
      for a in x:
          x_t.append((2-a)*bernoulli(a))
      y = []
      sum_x = 0
      for i in range(0, 10000):
          sum_x += x_t[i]
          y.append(sum_x)
```

```
[11]: fig, ax = plt.subplots()
      ax.plot(x, y);
      ax.set_xlabel('t')
      ax.set_ylabel('Expected final coordinate');
```



```
[12]: x = np.linspace(1, 10000, 10000)
x_t = []
p = 0.54
bernoulli = lambda m: ((1-p)**(m-1))*p
for a in x:
    x_t.append((2-a)*bernoulli(a))
y = []
sum_x = 0
for i in range(0, 10000):
    sum_x += x_t[i]
    y.append(sum_x)
```

```
[13]: fig, ax = plt.subplots()
ax.plot(x, y);
ax.set_xlabel('t')
ax.set_ylabel('Expected final coordinate');
```

1.5 Question-d

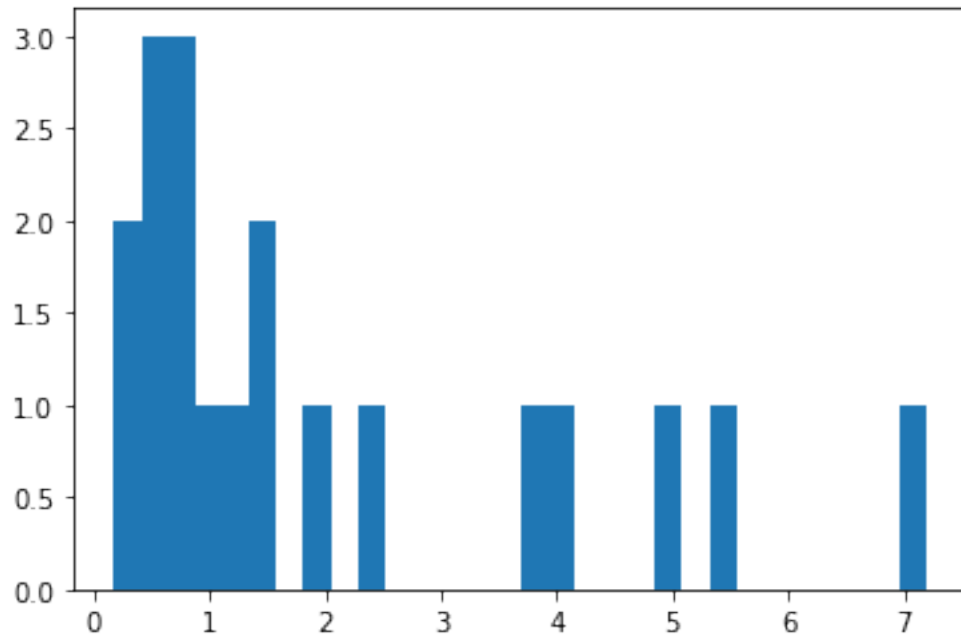
```
[14]: #Scale = (1/lambda)
arr = np.zeros(20)
for i in range(1, 20):
    arr[i] = arr[i-1] + (np.random.exponential(scale = 2.3))
```

```
[15]: dist = []
for i in range(1, 20):
    dist.append(arr[i]-arr[i-1])
```

1.5.1 Histogram of difference in exponentials

```
[16]: counts, bins = np.histogram(dist, bins=30)
plt.hist(dist, bins = 30);
counts
```

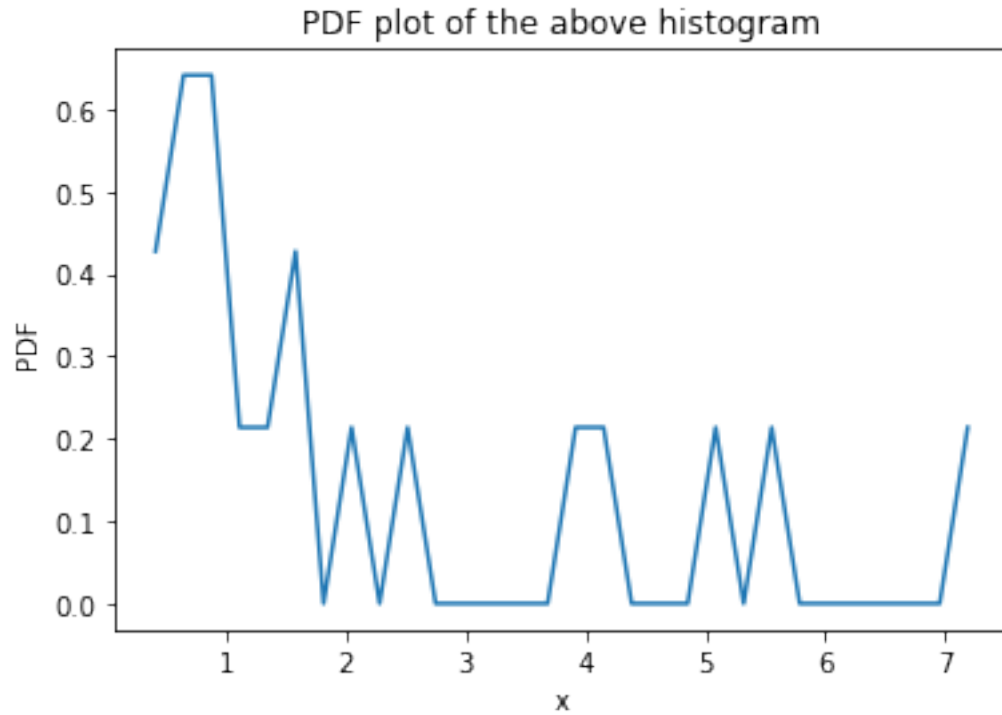
```
[16]: array([2, 3, 3, 1, 1, 2, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0,
          1, 0, 0, 0, 0, 0, 0, 1])
```



```
[17]: counts = list(counts)
      bins = list(bins)
      for i in range(0, len(counts)):
          counts[i] = counts[i]/20
```

```
[18]: pdf = counts/(bins[1]-bins[0])
```

```
[19]: fig, ax = plt.subplots()
      ax.set_xlabel("x")
      ax.set_ylabel("PDF")
      ax.set_title("PDF plot of the above histogram")
      ax.plot(bins[1:len(bins)], pdf);
```



1.5.2 Analysis of the graph

- Here, the mean(highest number of events) is at $x = (1/2.3) = 0.43$ where lambda of exponential was $1/2.3$.
- Also, $\lambda = 0.4310 \sim 3$ the highest frequency(or mean).
- Therefore, the above plotted histogram and PDF graph correspond to a poisson distribution.