

PHYS512_HW7

November 18, 2022

```
[1]: import numpy as np
import math
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
import matplotlib as mpl
mpl.style.use('seaborn')
import random
```

1 Problem 1

```
[2]: data = np.loadtxt('/Users/junalex Sugiyama/Desktop/PHYS512/repo/phys512-2022/
    ↪problem_sets/rand_points.txt')

x = data[:, 0]
y = data[:, 1]
z = data[:, 2]
```

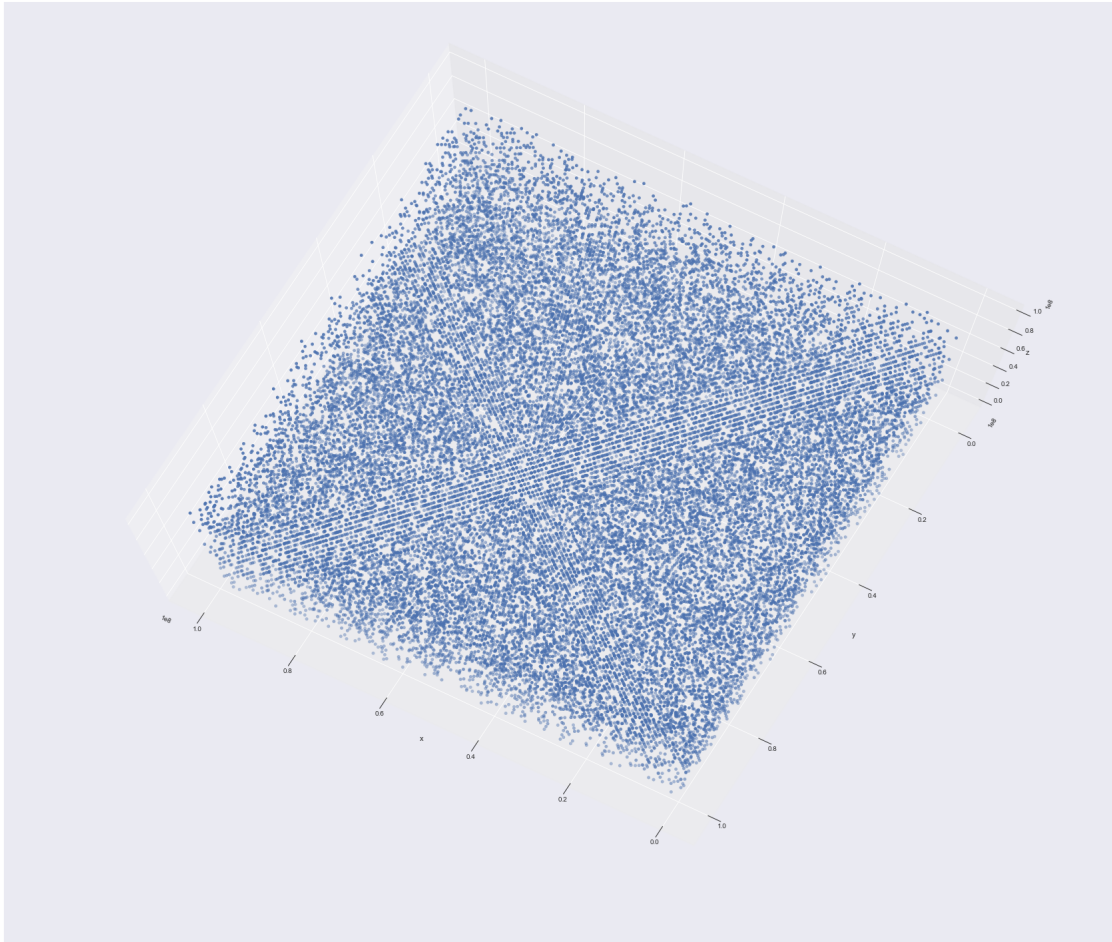
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```
[60]: fig = plt.figure(figsize = (30, 26))
ax = fig.add_subplot(111, projection='3d')
ax.scatter(x, y, z)
ax.set_xlabel('x')
ax.set_ylabel('y')
ax.set_zlabel('z')

# # All of those angles show some sort of structure existing in the "random" ↪
↪coordinates
elev_l = [-42.0, -42.0, -40, 60.0, 90 ]
azim_l = [105.5, 106.5, 100.5, 100.5, 100]
elev = 80
azim = 119
ax.view_init(elev, azim)

plt.show()
```

```
print('ax.azim {}'.format(ax.azim))
print('ax.elev {}'.format(ax.elev))
```



```
ax.azim 119
ax.elev 80
```

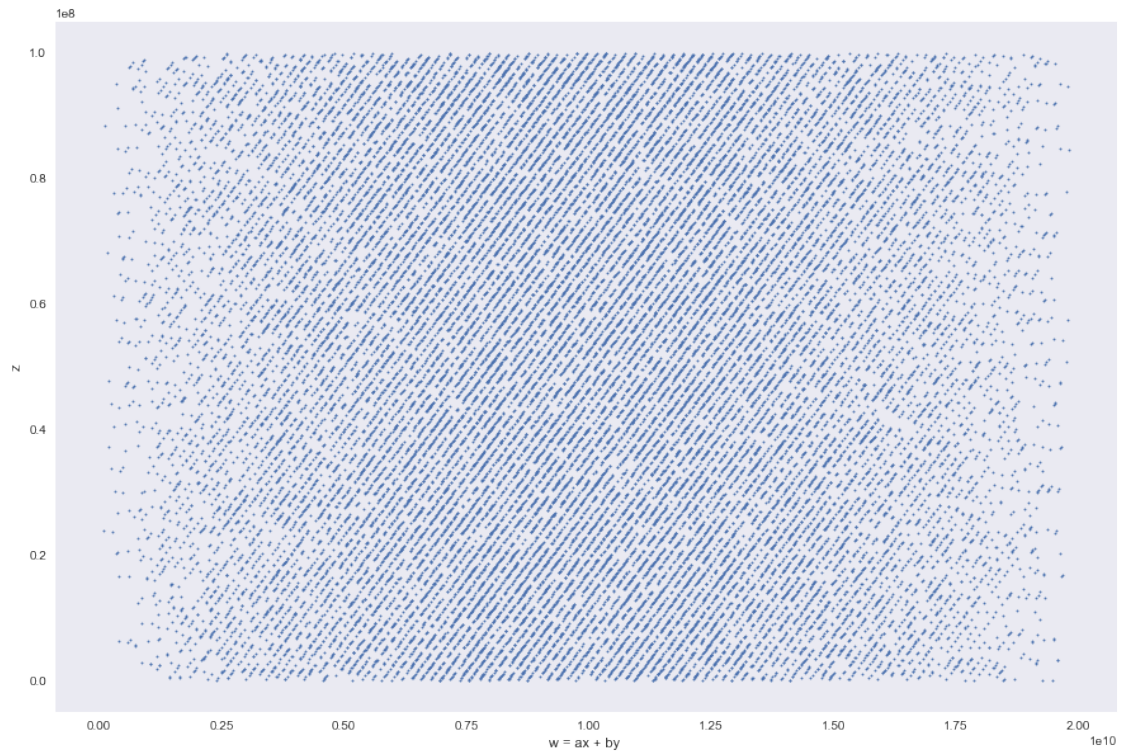
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[77]: # I just kept plugging in random numbers until i saw the pattern appear.
      #However, i took and inspiration from elevation and azimuth values in 3D plot
      a = 80
      b = 119

      w = a * x + b * y

      plt.figure(figsize = (15, 10))
      plt.scatter(w, z, s=2)
      plt.xlabel('w = ax + by ')
```

```
plt.ylabel('z')
plt.grid(b=None)
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2 Problem 2

```
[23]: def gauss(x):
        return 1/np.sqrt(2 * np.pi) * np.exp(-0.5 * x**2/2)

def lorentz(x):
    return 1/(np.pi * (1 + x**2))

def power(x):
    return x**(-1)

n=10000
yy=np.pi*(np.random.rand(n)-0.5)

M = 0.1
```

```

u1=np.pi * (np.random.rand(n)*3 - 0.5)
u2=np.pi * (np.random.rand(n) - 0.5)

idx = np.zeros([3, len(u1)])
funcs = ['gauss', 'lorentz', 'power']

idx_g, = np.where(u2<=gauss(u1)/max(gauss(u1)))
idx_l, = np.where(u2<=lorentz(u1)/max(lorentz(u1)))
idx_p, = np.where(u2<=power(u1)/max(power(u1)))

f_g = u1[idx_g]
f_l = u1[idx_l]
f_p = u1[idx_p]

ef_g = f_g.size/u1.size
ef_l = f_l.size/u1.size
ef_p = f_p.size/u1.size

print('Efficiency of Gaussian as bounding distribution: ', ef_g)
print('Efficiency of Lorentzian as bounding distribution: ', ef_l)
print('Efficiency of Power Law as bounding distribution: ', ef_p)

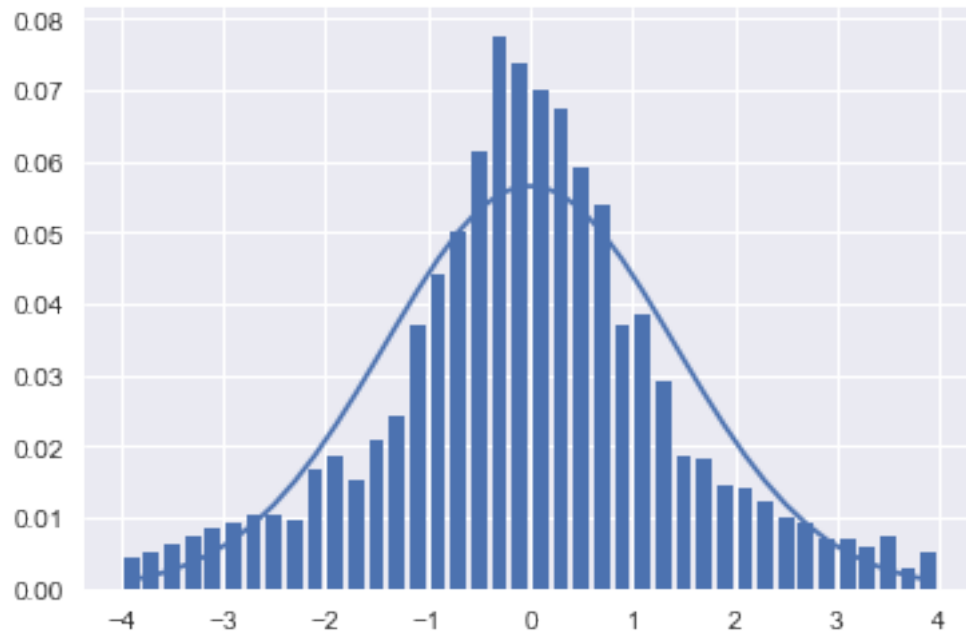
```

Efficiency of Gaussian as bounding distribution: 0.6054
 Efficiency of Lorentzian as bounding distribution: 0.581
 Efficiency of Power Law as bounding distribution: 0.4963

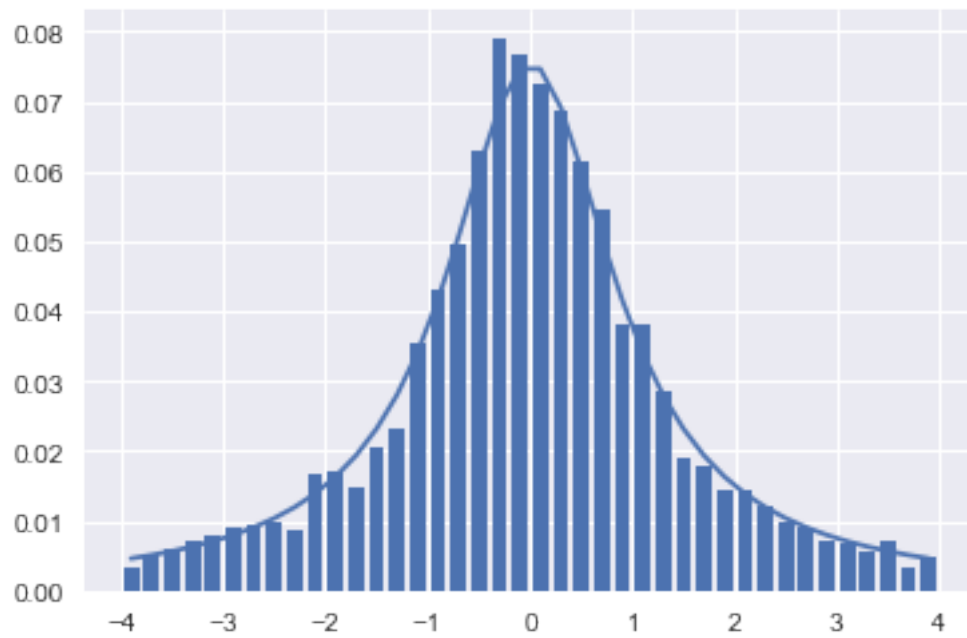
```

[24]: #Gaussian
y_use=f_g
x_use=np.tan(y_use)
aa,bb=np.histogram(x_use,np.linspace(-4,4,41))
b_cent=0.5*(bb[1:]+bb[:-1])
pred=gauss(b_cent)
pred=pred/pred.sum()
aa=aa/aa.sum()
plt.figure(2)
plt.clf()
plt.plot(b_cent,pred)
plt.bar(b_cent,aa,0.15)
plt.show()

```



```
[25]: #Lorentzian
y_use=f_l
x_use=np.tan(y_use)
aa,bb=np.histogram(x_use,np.linspace(-4,4,41))
b_cent=0.5*(bb[1:]+bb[:-1])
pred=lorentz(b_cent)
pred=pred/pred.sum()
aa=aa/aa.sum()
plt.figure(2)
plt.clf()
plt.plot(b_cent,pred)
plt.bar(b_cent,aa,0.15)
plt.show()
```



[]: