

```
# -*- coding: utf-8 -*-
```

```
"""
```

```
Created on Mon Apr 1 14:24:27 2019
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```
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```

```
"""
```

```
#import libraries:
```

```
import numpy as np
```

```
from numpy import cos,sin
```

```
import matplotlib.pyplot as plt
```

```
from mpl_toolkits.mplot3d import Axes3D
```

```
from scipy import stats
```

```
import time
```

```
from numpy.polynomial.polynomial import polyfit
```

```
import matplotlib.mlab as mlab
```

```
from scipy.stats import norm
```

```
#generates a non-uniform distribution between 0 and pi proportional to sin(theta)
```

```
#inverse-transformation method
```

```
def sin_dis(num):
```

```
    u = np.random.uniform(0,1,num)
```

```
    theta = np.arccos(1-2*u)
```

```
    return theta
```

```
#generates a non-uniform distribution between 0 and pi proportional to sin(theta)
```

```
#reject-accpet method
```

```
def reject_accept(num):
```

```
    ran = []
```

```
    # Counter test to calculate the percentage of points used
```

```
    naccept=0
```

```
    x = np.random.uniform(0,np.pi,num)
```

```
    y = np.random.uniform(0,1,num)
```

```
    criterion = y < np.sin(x)
```

```
    for i in range(num):
```

```
        if criterion[i]:
```

```
            ran.append(x[i])
```

```
            naccept=naccept+1
```

```
    percent_accept = (naccept/num)* 100
```

```
    return ran, percent_accept
```

```
#function that returns the random decay postion and time
```

```
def position_decay(num_events=1000000):
```

```
    tau = 550E-6
```

```
    speed = 2000
```

```
    time = np.random.exponential(scale = tau, size = 1000000)
```

```
position = speed*time
```

```
return time,position
```

```
#function that returns the random decay angle
```

```
def angle_decay(num):
```

```
    theta = sin_dis(num)
```

```
    phi = np.random.uniform(0,2*np.pi,num)
```

```
    theta -= np.pi/2
```

```
    return theta,phi
```

```
#function that returns the hit position on the detector
```

```
def position_on_detector(theta,phi):
```

```
    time,position = position_decay(num_events=1000000)
```

```
    rho = (2 - position)/cos(theta)
```

```
    x = rho*sin(theta)*cos(phi)
```

```
    y = rho*sin(theta)*sin(phi)
```

```
    #Resolution of the detector causes a smear on the gaussian distribution of the hit
```

```
positions
```

```
    res_x = 0.1
```

```
    res_y = 0.3
```

```
    smear_x = x + np.random.normal(0, res_x, 1000000)
```

```
    smear_y = y + np.random.normal(0,res_y, 1000000)
```

```
return x,y,smear_x, smear_y
```

```
#Test of the distribution for a unit sphere of a uniform distribution
```

```
def uniform_dist(num):
```

```
    theta = np.random.uniform(0,np.pi,num) # Opening angle to beam
```

```
    phi = np.random.uniform(0,2*np.pi,num) # Polar angle on screen, about axis of
```

```
beam
```

```
    x1 = sin(theta)*cos(phi)
```

```
    y1= sin(theta)*sin(phi)
```

```
    z1 = cos(theta)
```

```
    return x1,y1,z1
```

```
#Test of the distribution for a unit sphere of a non-uniform distribution
```

```
def non_uniform_dist(num):
```

```
    theta = sin_dis(num) # Opening angle to beam
```

```
    phi = np.random.uniform(0,2*np.pi,num)
```

```
#where rho is 1 in spherical coordinates
```

```
    x2 = sin(theta)*cos(phi)
```

```
    y2 = sin(theta)*sin(phi)
```

```
    z2 = cos(theta)
```

```
return x2,y2,z2
```

#Collider experiment to calculate the cross section of total number of candidate events  
observed is 5 with 95 percent confidence level

```
def confidence_interval(mincross, maxcross):
```

```
    steps = 100
```

```
    repeats = 10000
```

```
    percent = []
```

```
    over5 = []
```

```
    X = []
```

```
    for q in range(steps +1):
```

```
        total_signal = []
```

```
        background = []
```

```
        signal = []
```

```
        for i in range(repeats):
```

```
            bg_noise = np.random.normal(5.7,0.4)#background noise found using a
```

Gaussian distribution

```
            lum_error = np.random.normal(12,0.5)#Integrated luminosity uncertainty found  
using a Gaussian distribution
```

```
cross_sec = mincross + (maxcross-mincross)*q/steps #This allows zooming in  
around in a certain range of cross sections
```

```
bg_prod = np.random.poisson(bg_noise)#Poisson variation in the background  
production
```

```
lum = np.random.poisson(lum_error*cross_sec)#Poisson variation in the  
signal production
```

```
combined = np.random.poisson(lum_error*cross_sec + bg_noise )#combined  
signal production using a poisson distribution
```

```
total_signal.append(combined)
```

```
signal.append(lum)
```

```
background.append(bg_prod)
```

```
x = sum(float(n) > 5 for n in total_signal)
```

```
if 100*x/repeats > 95:
```

```
    X.append(cross_sec)
```

```
over5.append(100*x/repeats)
```

```
percent.append(mincross+(maxcross-mincross)*q/steps)
```

```
return over5, percent, total_signal, background, signal, X
```

```
MyInput = '0'
```

```

while MyInput != 'q':

    MyInput = input('Enter a choice, \n "a)" to investigate the analytical method,\n "b)"
to investigate the reject-accept method and compare the two methods, \n "c)" to
explore the partial experiment, \n "d)" to explore the statistical investigation, \n "q)" to
quit: ')

    print('You entered the choice: ',MyInput)


    if MyInput == 'a':

        print('You have chosen to generate random angles  $0 < \theta < \pi$  in a distribution
proportional to sin using the inverse transformation method.')


        print('Probability density function of the generated deviates compared to the sin(x)
function using the inverse transformation method:')

        num_bins = 100

        theta = sin_dis(100000)

        x = np.linspace(0, np.pi, 100)


        sinx = 1/2 * np.sin(x)

        plt.plot(x, sinx, linewidth=2, color = "red", label="$\\sin(\\theta)$")

        n, bins, patches = plt.hist(theta, num_bins , density = 0.5, facecolor='blue',
alpha=0.7, label="sinusoidal random \n number")

        plt.xlim([0,np.pi])

```

```

plt.legend(loc="upper left", fontsize="x-small", borderpad=1)

plt.xlabel("Random angle,  $\theta$ ")

plt.ylabel("Normalized frequency")

plt.show()

print('Test of the difference between the probability density function of the
generated deviates for each bin and sin(x) function using the inverse transformation
method:')

plt.subplot(2,1,2)

plt.bar(x, (sinx-n), width = 0.01, align='center', alpha=1)

plt.xlabel("Random angle,  $\theta$ ")

plt.ylabel("Difference between \n PDF and sin(x)")

plt.show()


print('The first 4 statistical moments of the sinusoidal distribution between 0 and pi
')

print("Number", " ", "Moment")    #table column headings

for x in range(1,5):

    print(x, ' ', stats.moment(theta, moment = x))

elif (MyInput == 'b'):

    print('Probability density function of the generated deviates compared to the sin(x)
function using the reject and accept method:')

    theta, percent_accept = reject_accept(num=100000)

```



```
print('The percent of points accepted using the reject-accept method:',  
percent_accept, '%')  
  
num_bins = 100  
  
n, bins, patches = plt.hist(theta, num_bins, color="red", density = 1, alpha=0.7,  
label="Sinusoidal random number")
```

```
x = np.linspace(0, np.pi, 100)
```

```
sinx = 1/2* np.sin(x)
```

```
plt.plot(x, sinx, linewidth=2, color="blue", label="$\\sin(\\theta)$")
```

```
plt.xlim([0,np.pi])
```

```
plt.xlabel("Random angle, $\\theta$")
```

```
plt.ylabel("Normalized Frequency")
```

```
plt.legend(loc="upper left", fontsize="x-small", borderpad=1)
```

```
plt.show()
```

```
print('Test of the difference between the probability density function of the  
generated deviates for each bin and sin(x) function:')
```

```
plt.subplot(2,1,2)
```

```
plt.bar(x, (sinx-n), width = 0.01, align='center', alpha=1)
```

```
plt.xlabel("Random angle,  $\theta$ ")
```

```
plt.ylabel("Difference between \n PDF and sin(x)")
```

```
plt.show()
```

```
print('The first 4 statistical moments of the sinusoidal distribution between 0 and pi  
)
```

```
print("Number"," ", "Moment")    #table column headings
```

```
for x in range(1,5):
```

```
    print(x,' ',stats.moment(theta, moment = x))
```

```
N = []
```

```
t_analytic = []
```

```
t_reject = []
```

```
print('Please wait while the speed efficency of the two methods are compared:')
```

```
for num in range(10000,1000000,10000):
```

```
    start1 = time.time()
```

```
    x = sin_dis(num)
```

```
    end1 = time.time()
```

```
    start2 = time.time()
```

```
    x = reject_accept(num)
```

```
end2 = time.time()
```

```
t_analytic.append(end1-start1)
```

```
t_reject.append(end2-start2)
```

```
N.append(num)
```

```
plt.plot(np.unique(N), np.poly1d(np.polyfit(N, t_analytic, 1))(np.unique(N)))
```

```
plt.plot(N,t_analytic, color = "b",label = "Analytic")
```

```
plt.plot(np.unique(N), np.poly1d(np.polyfit(N, t_reject, 1))(np.unique(N)))
```

```
plt.plot(N,t_reject, color = "r",label = "Reject")
```

```
plt.legend(title = "Method",fontsize = 15)
```

```
plt.xlabel('Number of angles generated',fontsize = 15)
```

```
plt.ylabel('Time (s)',fontsize = 15)
```

```
plt.show()
```

```
elif (MyInput == 'c'):
```

```
time,position = position_decay(num_events=1000000)
```

```
print('Exponential distribution of the random decay times and the random decay  
positions ')
```

```
plt.subplot(2,1,1)
```

```
plt.hist(time, 100 ,color="pink")
```

```
#plt.xlim([0,0.0025])
```

```
plt.xlabel("Time after injection,  $s^{-1}$ ")
```

```
plt.ylabel("Relative Frequency")
```

```
plt.title("Random Decay Times")
```

```
plt.subplot(2,1,2)
```

```
plt.hist(position, 100, color="purple")
```

```
plt.xlabel("Position from beam injection,  $m^{-1}$ ")
```

```
#plt.xlim([0,6])
```

```
plt.ylabel("Relative Frequency")
```

```
plt.title("Random Decay Position")
```

```
plt.tight_layout()
```

```
plt.show()
```

```
theta,phi = angle_decay(num = 1000000)
```

```
print('Unit sphere of randomly distribution angles with a uniform distribution and  
non-uniform distribution using the inverse transformation method ')
```

```
num = 10000
```

```
x1,y1,z1 = uniform_dist(num)
```

```
x2,y2,z2 = non_uniform_dist(num)
```

```
fig = plt.figure(figsize=plt.figaspect(0.5))  
ax1 = fig.add_subplot(1,2,1, projection='3d')  
ax2 = fig.add_subplot(1,2,2, projection='3d')  
  
ax1.scatter(x1,y1,z1, s = 0.1)  
ax1.set_xlabel("x",fontsize = 15)  
ax1.set_ylabel("y",fontsize = 15)  
ax1.set_zlabel("z",fontsize = 15)  
ax1.set_title("Uniform distribution",fontsize = 15)
```

```
ax2.scatter(x2,y2,z2, s = 0.1)  
ax2.set_title("Non-uniform distribution",fontsize = 15)  
ax2.set_xlabel("x",fontsize = 15)  
ax2.set_ylabel("y",fontsize = 15)  
ax2.set_zlabel("z",fontsize = 15)  
plt.show()
```

```
print('Histogram plot of decay angles phi and theta in a random distribution ')
```

```
plt.hist2d(phi, theta, bins=100, cmap=plt.cm.BuPu)  
plt.xlabel("Angle  $\theta$ ")
```

```
plt.ylabel("Angle  $\varphi$ ")
```

```
cbar = plt.colorbar()
```

```
cbar.solids.set_edgecolor("face")
```

```
plt.draw()
```

```
plt.show()
```

```
x,y,smear_x,smear_y = position_on_detector(theta,phi)
```

```
# Range limits of the detector
```

```
xmin = -1
```

```
xmax = 1
```

```
ymin = - 1
```

```
ymax = 1
```

```
detector_range = [[xmin,xmax],[ymin,ymax]]
```

```
print('2D Histogram plot of the detector hits without the smearing due to  
resolution')
```

```
plt.hist2d(x, y,bins=40, range=detector_range, cmap=plt.cm.BuPu)
```

```
plt.xlabel("$X$ position on detector, m")
```

```
plt.ylabel("$Y$ position on detector, m")
```

```
cbar = plt.colorbar()
```

```
cbar.solids.set_edgecolor("face")
```

```
plt.draw()
```

```
plt.show()
```

```
print('2D Histogram plot of the detector hits with smearing due to resolution of x=  
0.1m and y = 0.3m ')
```

```
plt.hist2d(smear_x, smear_y, bins=40, range=detector_range, cmap=plt.cm.BuPu)
```

```
plt.xlabel("$X$ position on detector, m")
```

```
plt.ylabel("$Y$ position on detector, m")
```

```
cbar = plt.colorbar()
```

```
cbar.solids.set_edgecolor("face")
```

```
plt.draw()
```

```
plt.show()
```

```
meanx = np.mean(smear_x)
```

```
variancex = np.var(smear_x)
```

```
sigmax = np.sqrt(variancex)
```

```
print('1D Histogram plot of the detector hits in the x with smearing due to  
resolution of  $x = 0.1\text{m}$ ')
```

```
plt.hist(smear_x, bins=np.arange(-1, 1, 0.01), density = 1)
```

```
plt.xlabel("$X$ position on detector, m")
```

```
plt.ylabel("Frequency of hit on detector")
```

```
plt.show()
```

```
print('The mean value is', meanx, 'the variance is', variancex, 'and the standard  
deviation is ', sigmax)
```

```
print('1D Histogram plot of the detector hits in the y with smearing due to  
resolution of  $y = 0.3\text{m}$ ')
```

```
plt.hist(smear_y, bins=np.arange(-1, 1, 0.01), density = 1)
```

```
plt.xlabel("$Y$ position on detector, m")
```

```
plt.ylabel("Frequency of hit on detector")
```

```
plt.show()
```

```
meany = np.mean(smear_y)
```

```
variancey = np.var(smear_y)
```

```
sigmay = np.sqrt(variancey)
```



```
print('The mean value is',mean,'the variance is',variance, 'and the standard  
deviation is ',sigmay)
```

```
elif (MyInput == 'd'):
```

```
over5, percent, total_signal, background, signal , X =  
confidence_interval(mincross = 0.01 ,maxcross = 1)
```

```
print('The first cross section that produces a 95% confidence level that the event  
is over 5 is', min(X), 'nb')
```

```
print('Confidence level over 5 events for a range of cross sections, the red dashed  
line marks a 95% confidence interval with a range of 0.01 to 1:')
```

```
plt.hlines(y = 95, xmin = 0.01, xmax= 1, color='red', linestyle='dashed', label='95  
confidence')
```

```
plt.xlabel("Diameter of cross section")
```

```
plt.ylabel("Percent of events that occur over 5")
```

```
plt.plot(percent,over5)
```

```
plt.show()
```

```

over5, percent, total_signal, background, signal , X =
confidence_interval(mincross = 0.35 ,maxcross = 0.45)

print('Confidence level over 5 events for a range of cross sections, the red dashed
line marks a 95% confidence interval with a range of 0.35 to 0.45:')

plt.hlines(y = 95, xmin = 0.35, xmax= 0.45, color='red', linestyle='dashed',
label='95 confidence')

plt.xlabel("Diameter of cross section")
plt.ylabel("Perecent of events that occur over 5")
plt.plot(percent,over5)
plt.show()

```

```

over5, percent, total_signal, background, signal , X =
confidence_interval(mincross = 0.01 ,maxcross = 1)

num_bins = 100

print('Background and signal production modelled using Poisson distributions')

n, bins, patches = plt.hist(background, num_bins, density = 1, color="red",
alpha=0.5, label = 'Background')

n, bins, patches = plt.hist(signal,num_bins, density = 1, color="blue",
alpha=0.5,label = 'Signal')

plt.xlabel('Number of candiate events')

```

```
plt.ylabel('Production Frequency')
```

```
plt.legend(loc="upper right", fontsize="medium", borderpad=1)
```

```
plt.show()
```

```
meanb = np.mean(background)
```

```
varianceb = np.var(background)
```

```
sigmab = np.sqrt(varianceb)
```

```
means = np.mean(signal)
```

```
variances = np.var(signal)
```

```
sigmas = np.sqrt(variances)
```

```
print('The mean value of the background is',meanb,'the variance is',varianceb,  
'and the standard deviation is ',sigmab)
```

```
print('The mean value of the signal is',means,'the variance is',variances, 'and the  
standard deviation is ',sigmas)
```

```
print('The combined background and signal production modelled using Poisson  
distributions')
```

```
num_bins = 100
```

```
n, bins, patches = plt.hist(total_signal, num_bins, density =1, color="red",  
alpha=0.5)
```

```
plt.xlabel('Number of candiate events')
```

```
plt.ylabel('Production Frequency')
```

```
plt.show()
```

```
meant = np.mean(total_signal)
```

```
variancet = np.var(total_signal)
```

```
sigmat = np.sqrt(variancet)
```

```
print('The mean value of the total signal is',meant,'the variance is',variancet, 'and  
the standard deviation is ',sigmat)
```

```
elif MyInput != 'q':
```

```
    print('This is not a valid choice')
```

```
print('You have chosen to finish - goodbye.')
```