

```
In [ ]: import numpy as np
import matplotlib.pyplot as plt

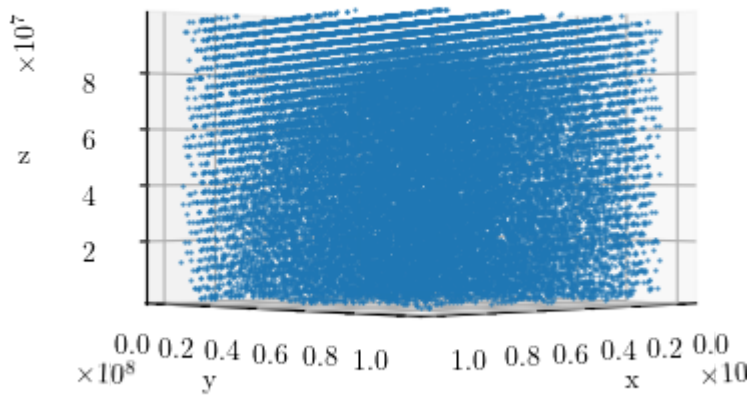
plt.rc('text', usetex=True)
plt.rc('font', family='serif')
plt.rcParams['figure.dpi'] = 100
```

## Question 1

I could not figure out how to use the C pseudo-random number generator given by Jon. So I'll just use the txt file provided. Also, I'll check if the Python ( `numpy.random.randint` ) PRNG is flawed or not.

```
In [ ]: # flawed C-PRNG
# Loading the txt file provided
points=np.loadtxt('rand_points.txt')
x=points[:,0]
y=points[:,1]
z=points[:,2]

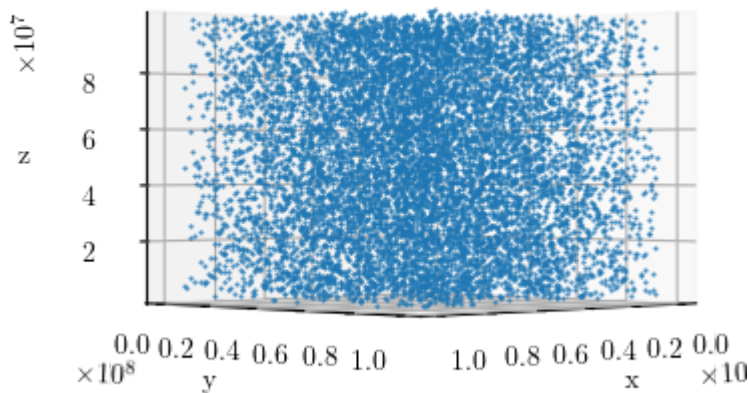
# plotting stuff
fig=plt.figure()
ax=fig.add_subplot(projection='3d')
ax.plot(x,y,z,'.',markersize='1')
ax.view_init(0, 45)
ax.set_xlabel('x')
ax.set_ylabel('y')
ax.set_zlabel('z')
plt.show()
```



We can see lines; it really does look like the 'randomly' generated points lie on a set of planes!  
Not so random.

```
In [ ]: # Numpy PRNG
# getting (x,y,z) coordinates between 0 and 1e8, using numpy random number
# generator
x=[];y=[];z=[]
for i in range(10000):
    x.append(np.random.randint(0,1e8))
    y.append(np.random.randint(0,1e8))
    z.append(np.random.randint(0,1e8))

# plotting stuff
fig=plt.figure()
ax=fig.add_subplot(projection='3d')
ax.plot(x,y,z,'.',markersize=1)
ax.view_init(0, 45)
ax.set_xlabel('x')
ax.set_ylabel('y')
ax.set_zlabel('z')
plt.show()
```



It does not look like the points lie on a set of planes; completely randomly generated points!

## Question 2

Theoretically, I could use any of the Lorentzians, Gaussians or power laws to generate the exponential deviates, so long as the bounding distribution is greater than the exponential for all values of  $x$ . The Gaussian function will be a little harder to use, because the CDF does not really have simple form (it is the ERF, error function), but I don't doubt that it is possible to work with it in Python. However, in this question I will just use the Lorentzian because why not.

A Lorentzian has the form:  $\text{PDF}(x) = \frac{1}{1+x^2}$ . So the CDF is:

$$\text{CDF}(x) = \int_0^x \frac{1}{1+x^2} dx = \arctan(x)$$

Set the CDF equal to some random number between 0 and 1:

$$\text{CDF}(x) = \arctan(x) = r.$$

Solve for  $x$ :

$$x = \tan(r).$$

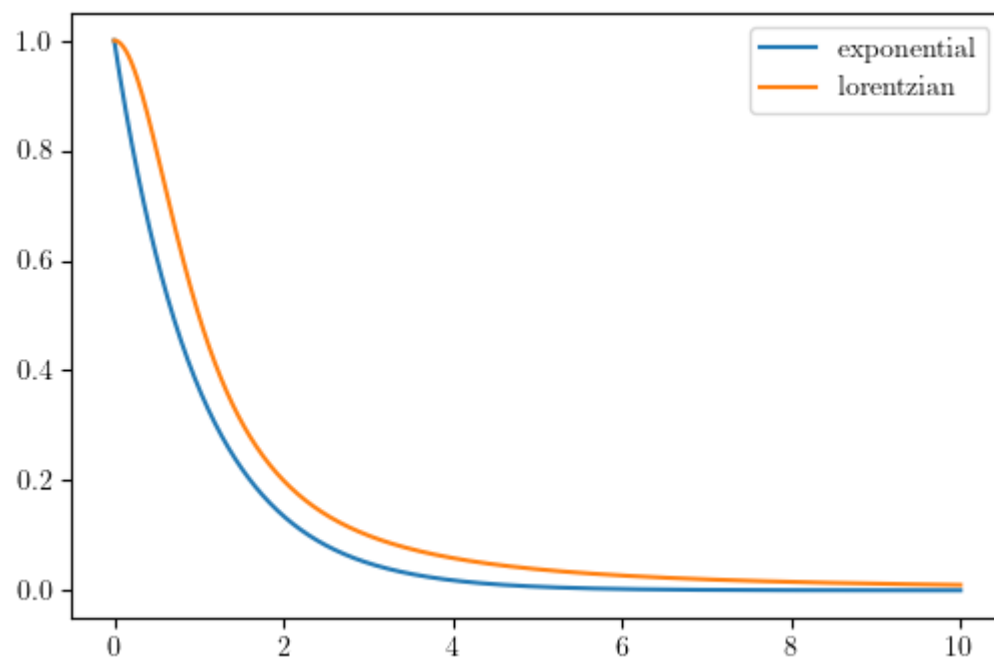
```
In [ ]: # relation we derived above: x=tan(r)
def lorentzian(n):
    r=np.pi*(np.random.rand(n)-0.5) # tan only takes numbers between -pi
    and pi
    return np.tan(r)
```

```
# first let's check if the Lorentzian is always above the exponential we
want to generate
x=np.linspace(0,10,1001)

def model(x):
    return np.exp(-x)

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

plt.plot(x,model(x),label='exponential')
plt.plot(x,lorentz(x),label='lorentzian')
plt.legend()
plt.show()
print('all good')
```



all good

```
In [ ]: n=10000000 # number of random numbers

# generating the non-uniform random numbers
l=lorentzian(n)
u=lorentz(l)*np.random.rand(n)

# selecting only those that lie *below* the exponential
accept=u<model(l)
```

```

use=l[accept]

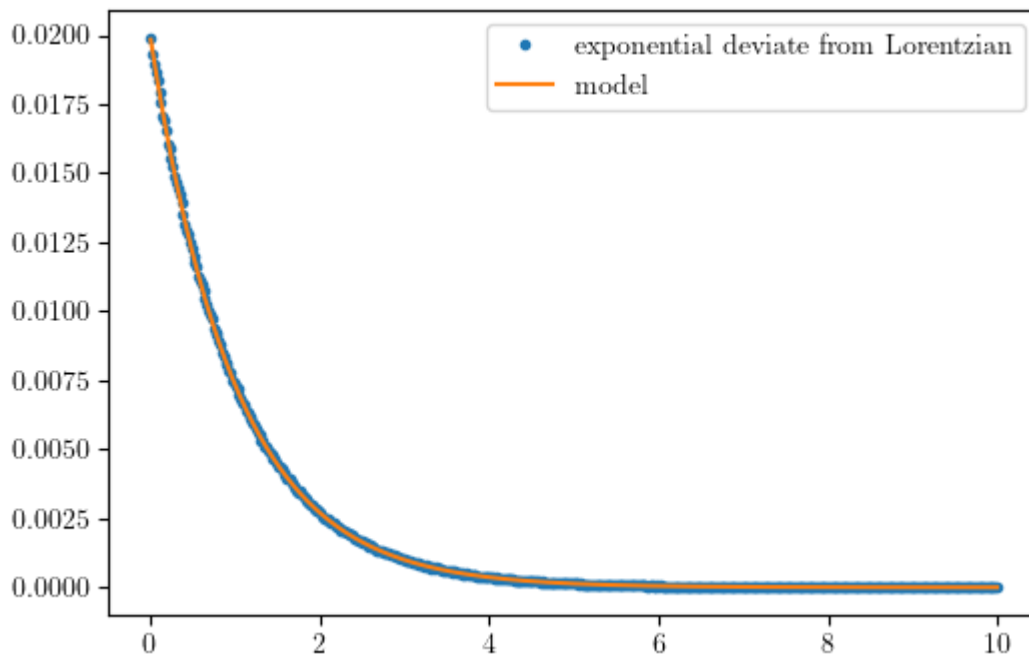
# effeciency (how many samples we actually use)
eff=np.size(use)/np.size(accept)*100

# plotting stuff
a,b=np.histogram(use,np.linspace(0,10,500))
a=a/a.sum()
bb=0.5*(b[1:]+b[:-1]) # get center of bins
pred=model(bb)
pred=pred/pred.sum()

plt.plot(bb,a,'.', label='exponential deviate from Lorentzian')
plt.plot(bb,pred,label='model')
plt.legend()
plt.show()
print('Effeciency of generator is: ',eff)

```

C:\Users\Greg\AppData\Local\Temp\ipykernel\_3820\1798579393.py:10: RuntimeWarning: overflow encountered in exp  
 return np.exp(-x)



Effeciency of generator is: 81.80398

Looks good!

## Question 3

Same as problem 2, but now I use a ratio-of-uniforms generator.

The math is that we take a  $(u, v)$  plane, where

$$0 < u < \sqrt{\text{PDF}(v/u)}.$$

In our case,  $\text{PDF}(x) = e^{-x}$  and  $u$  runs from 0 to 1

So we have

$$0 < u < \sqrt{e^{-v/u}},$$

$$0 < u^2 < e^{-v/u},$$

$$\ln(0) < \ln(u^2) < -v/u,$$

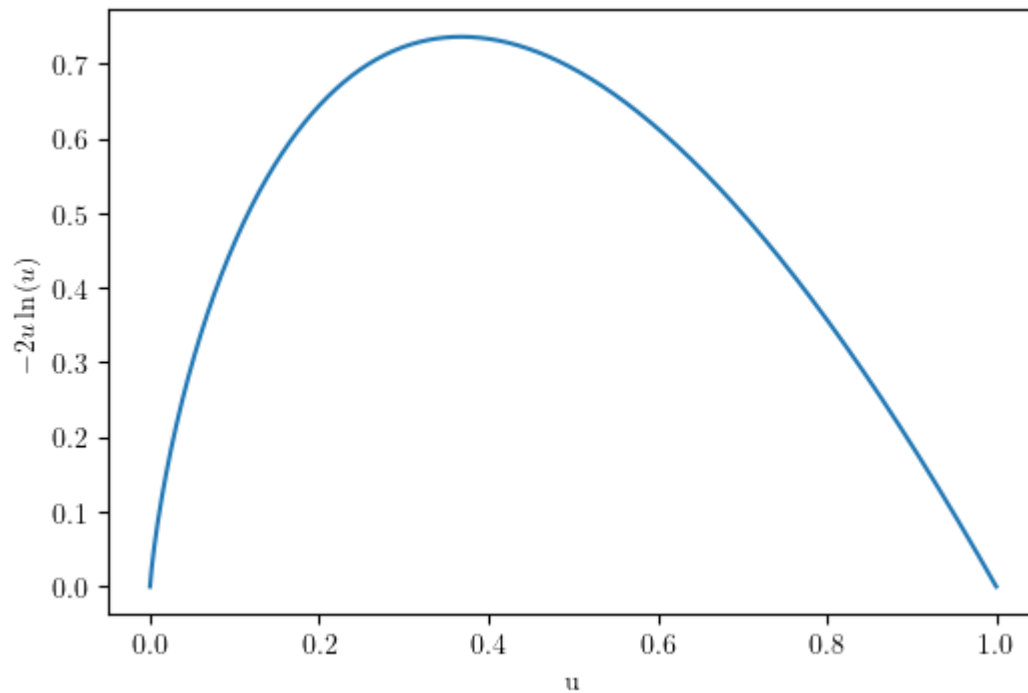
$$\ln(0) > -2u \ln(u) > v.$$

$\ln 0 = \infty$  so we can discard that bound (of course  $v$  will always be smaller than  $\infty$ ). So we are left with

$$-2u \ln(u) > v.$$

To find the upper bound on  $v$  (the lower bound is 0) let's pick the  $u$ , between 0 and 1, that will maximize  $-2u \ln(u)$

```
In [ ]: x=np.linspace(0.0000001,1,1001)
plt.plot(x,-2*x*np.log(x))
plt.xlabel('u')
plt.ylabel('$-2u\ln\{u\}$')
plt.show()
```



There is one extremum (it is a maximum) between 0 and 1 so let's take the derivative and set it equal to zero:

$$\frac{d}{du}(-2u \ln(u)) = -2 - 2 \ln(u) = 0,$$

$$\ln u = -1,$$

$$u = e^{-1}.$$

So the upper bound on  $v$  is  $-2(e^{-1}) \ln(e^{-1}) = 2e^{-1}$ .

Since the lower bound is 0:

$$0 < v < 2e^{-1}.$$

```
In [ ]: n=10000000 # number of random numbers

u=np.random.rand(n) # random numbers from 0 to 1
v=(np.random.rand(n))*2*np.exp(-1) # random numbers from 0 to e^-1

# ratio of uniforms
a=u<np.sqrt(np.exp(-v/u))
use=v[a]/u[a]

# effeciency (how many ratios we actually use)
eff=np.size(use)/np.size(a)*100
```

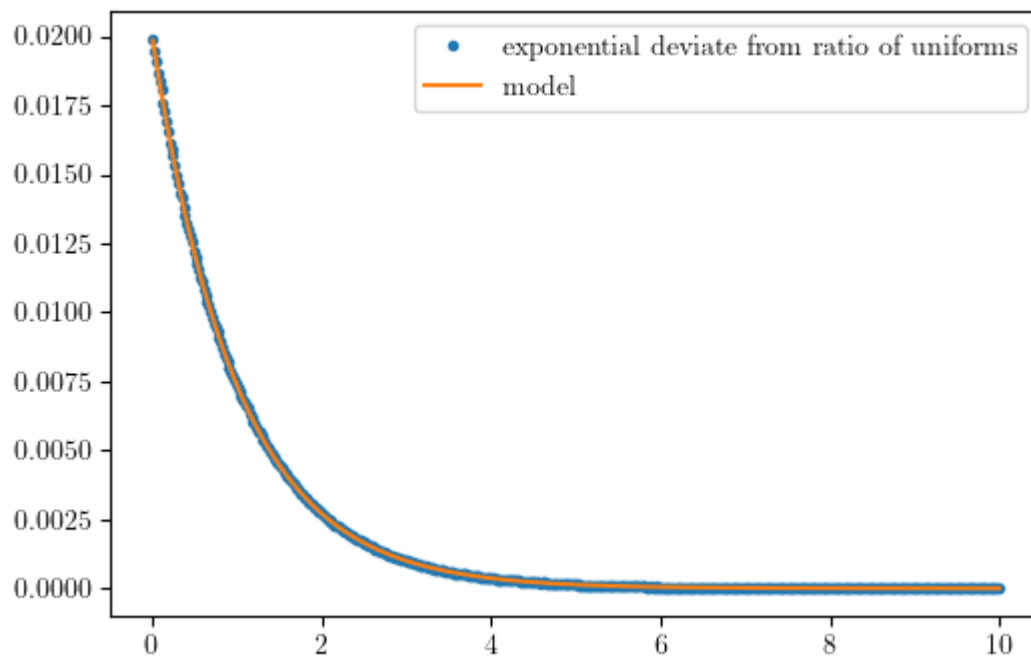
```

# plotting stuff
a,b=np.histogram(use,np.linspace(0,10,500))
a=a/a.sum()
bb=0.5*(b[1:]+b[:-1]) # get center of bins
pred=model(bb)
pred=pred/pred.sum()

plt.plot(bb,a,'.', label='exponential deviate from ratio of uniforms')
plt.plot(bb,pred,label='model')
plt.legend()
plt.show()

print('Effeciency of generator is: ',eff)

```



Effeciency of generator is: 67.96546

Not bad!!