

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import scipy.stats as stats
import numpy as np
import pandas as pd
```

## Problem 1

```
In [4]: def solve_p1(err):
    fig, axs = plt.subplots(1, 2, figsize=(7,3))

    dist = stats.norm(1, err/100)
    flux = dist.rvs(10000)
    flux_fit = np.linspace(0.001, 2, 1000)
    pdf_flux_fit = dist.pdf(flux_fit)

    mag = -2.5 * np.log10(flux)
    mag_fit = -2.5 * np.log10(flux_fit)
    pdf_mag_fit = pdf_flux_fit.copy()
    pdf_mag_fit[1:] /= abs(mag_fit[1:] - mag_fit[:-1])
    pdf_mag_fit /= np.dot(pdf_mag_fit[1:], abs(mag_fit[1:] - mag_fit[:-1]))

    ax1, ax2 = axs

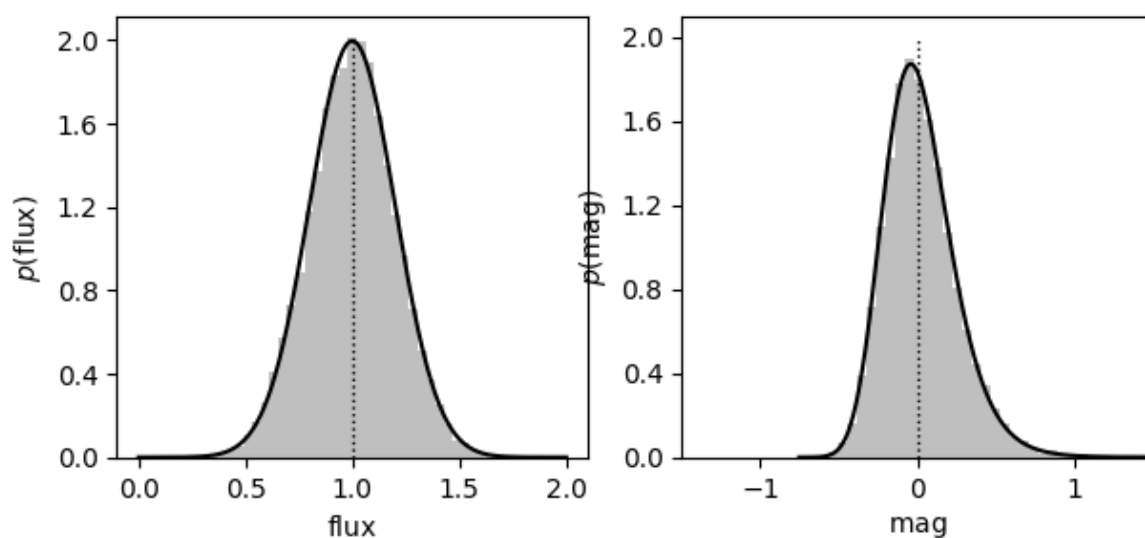
    ax1.hist(flux, bins=np.linspace(0, 2, 50), histtype='stepfilled', fc='
    ax1.plot(flux_fit, pdf_flux_fit, '-k')
    ax1.plot([1, 1], [0, 2], ':k', lw=1)
    ax1.set_xlabel(r'$\rm flux$')
    ax1.set_ylabel(r'$p(\rm flux)$')
    ax1.yaxis.set_major_locator(plt.MultipleLocator(0.4))

    ax2.hist(mag, bins=np.linspace(-1, 2, 50), histtype='stepfilled', fc='
    ax2.plot(mag_fit, pdf_mag_fit, '-k')
    ax2.plot([0, 0], [0, 2], ':k', lw=1)
    ax2.set_xlim(-1.5, 1.5)
    ax2.set_xlabel(r'$\rm mag$')
    ax2.set_ylabel(r'$p(\rm mag)$')
    ax2.yaxis.set_major_locator(plt.MultipleLocator(0.4))

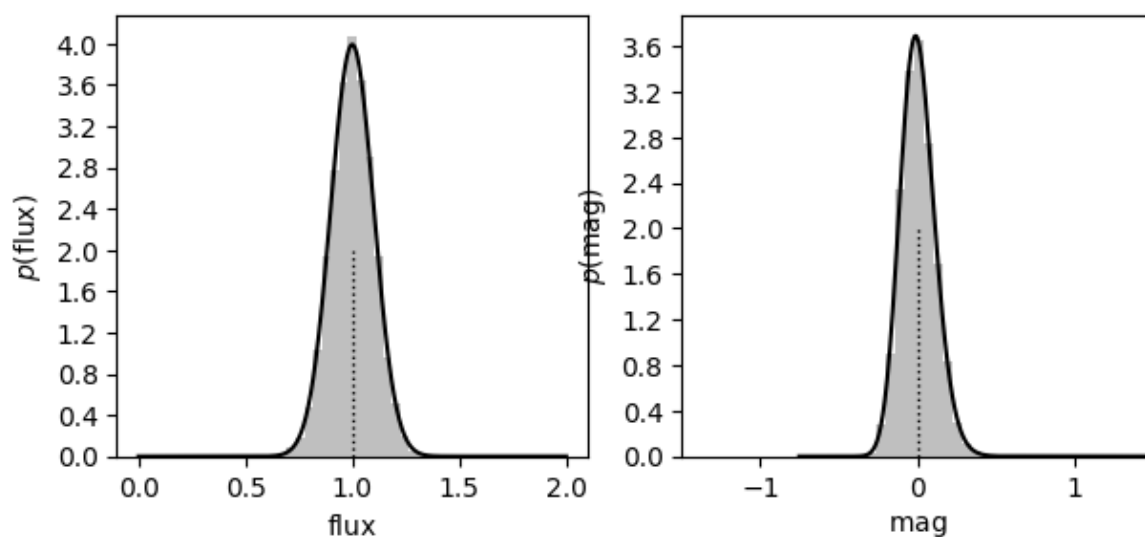
    plt.suptitle('{ }% flux error'.format(err))
```

```
In [5]: solve_p1(20)
solve_p1(10)
solve_p1(5)
```

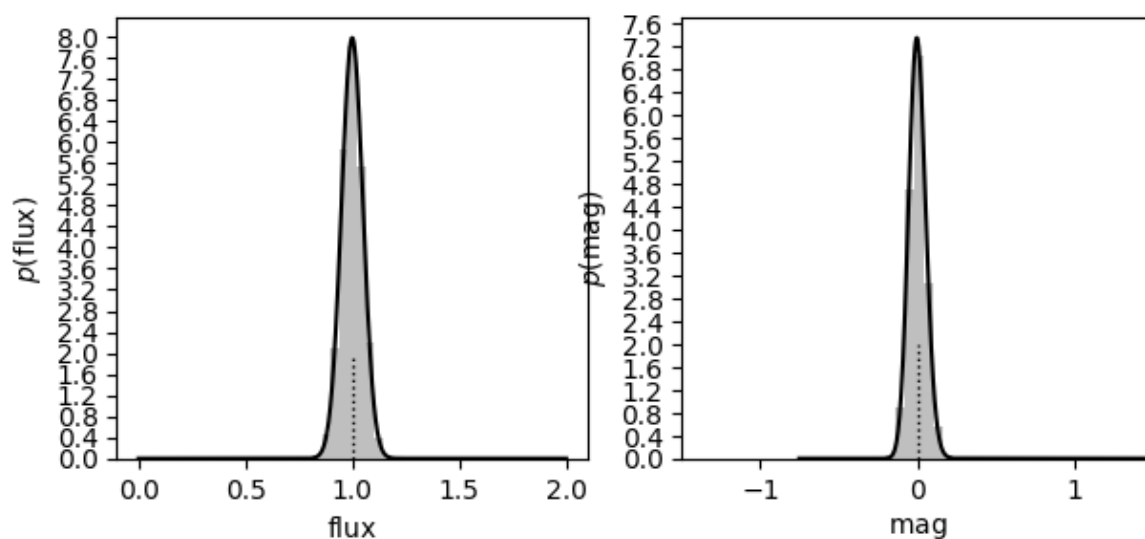
20% flux error



10% flux error



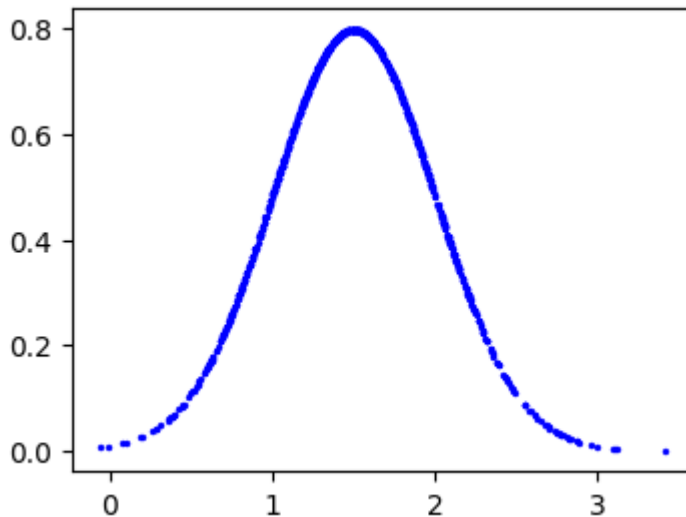
5% flux error



With a decrease in the flux error there is a decrease in the asymmetry of the magnitude graph which stems from the logarithmic dependence of the magnitude on the flux.

## Problem 2

```
In [6]: dist=stats.norm(1.5,0.5)
r=dist.rvs(1000)
p=dist.pdf(r)
plt.figure(figsize=(4,3))
plt.scatter(r,p,color='blue',s=2)
plt.show()
```



```
In [7]: mean,variance,skewness,kurtosis=dist.stats(moments='mvsk')
mad=np.median(np.abs(r-np.median(r)))
sigma_g=0.7413*(np.percentile(r,75)-np.percentile(r,25))

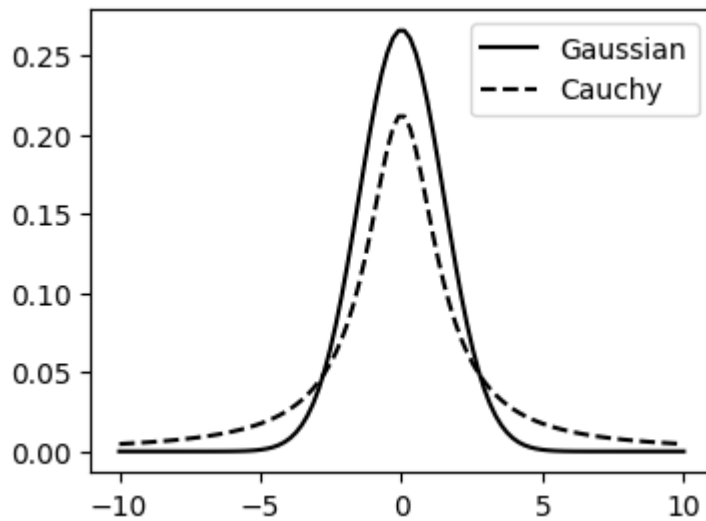
print('mean=',mean)
print('variance=',variance)
print('skewness=',skewness)
print('kurtosis=',kurtosis)
print('Std using MAD=',1.482*mad)
print('Sigma_g=',sigma_g)
```

```
mean= 1.5
variance= 0.25
skewness= 0.0
kurtosis= 0.0
Std using MAD= 0.5124995245167813
Sigma_g= 0.5076696407778214
```

## Problem 3

```
In [8]: x=np.linspace(-10,10,100)
gauss=stats.norm.pdf(x,0,1.5)
cauchy=stats.cauchy.pdf(x,0,1.5)
plt.figure(figsize=(4,3))
plt.plot(x,gauss,color='black',label='Gaussian')
plt.plot(x,cauchy,'--',color='black',label='Cauchy')
plt.legend()
```

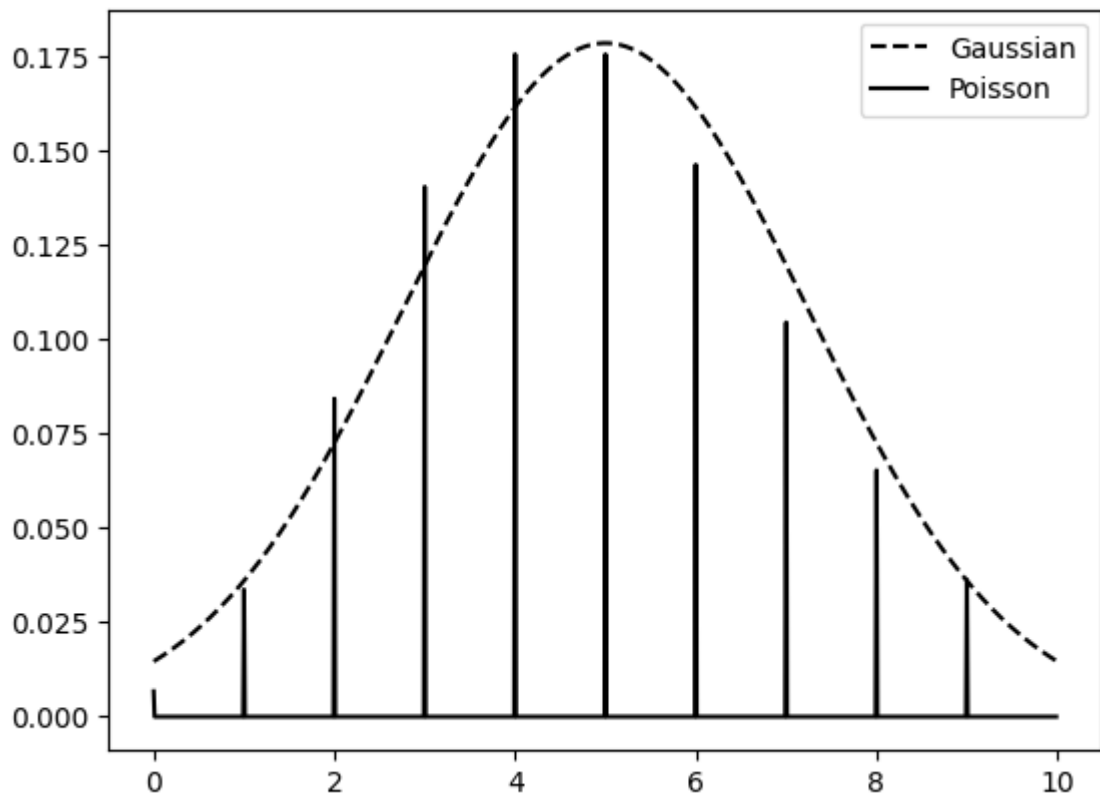
```
Out[8]: <matplotlib.legend.Legend at 0x7fca05110f10>
```



## Problem 4

```
In [9]: x=np.arange(0,10,0.01)
gauss=stats.norm.pdf(x,5,np.sqrt(5))
poisson=stats.poisson.pmf(x,mu=5)
plt.figure()
plt.plot(x,gauss,'--',color='black',label='Gaussian')
plt.plot(x,poisson,color='black',label='Poisson')
plt.legend()
```

Out[9]: <matplotlib.legend.Legend at 0x7fca0517c1f0>



## Problem 5

The expressions for the weighted mean of a quantity and its uncertainty are given by:

$$\bar{x} = \frac{\sum_{i=1}^N x_i / \sigma_i^2}{\sum_{i=1}^N 1 / \sigma_i^2} \quad \sigma_x^2 = \frac{1}{\sum_{i=1}^N 1 / \sigma_i^2}$$

```
In [10]: lt=1e-10*np.array([0.8920,0.881,0.8913,0.9837,0.8958])
delta=1e-10*np.array([0.00044,0.009,0.00032,0.00048,0.00045])
wm=np.sum(lt/delta**2)/np.sum(1/delta**2)
uc=np.sqrt(1/np.sum(1/delta**2))
print('Weighted mean=',wm)
print('Uncertainty=',uc)
```

Weighted mean= 9.089185199574898e-11

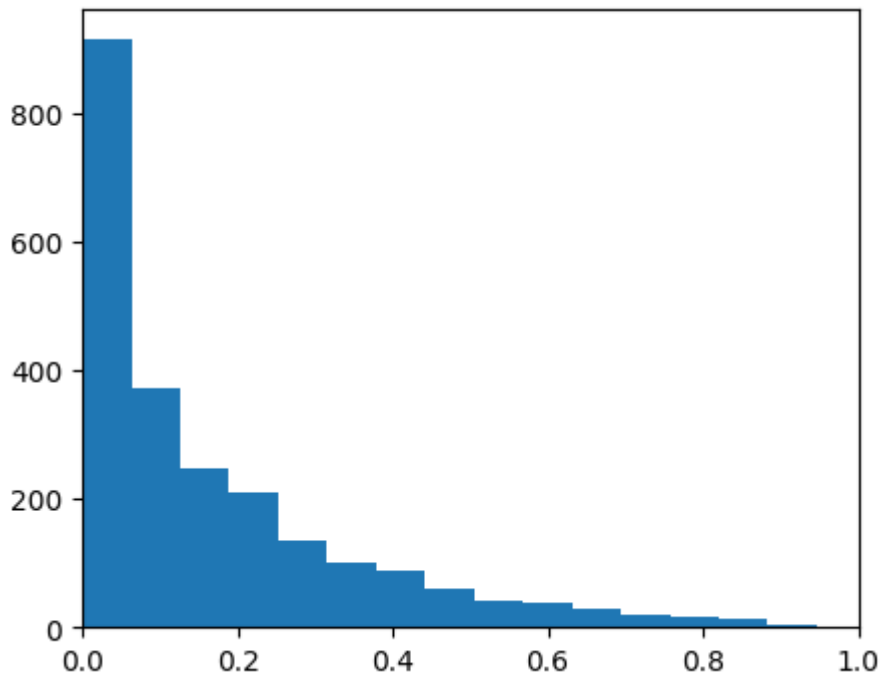
Uncertainty= 2.0318737026848624e-14

Weighted mean lifetime=  $(9.089 \pm 0.002) \times 10^{-11} s$

## Problem 6

```
In [11]: data=pd.read_csv('Downloads/eccen.csv')
e=data['eccentricity'].dropna().to_numpy()
plt.figure(figsize=(5,4))
plt.hist(e,bins=50)
plt.xlim(0,1)
```

Out[11]: (0.0, 1.0)



```
In [12]: e_gauss,_=stats.boxcox(e[e>0])
fig,ax=plt.subplots(figsize=(5,4))
ax.hist(e_gauss, 50)
```

```
Out[12]: (array([ 1.,  1.,  0.,  3.,  2.,  3.,  5.,  7., 10.,  7., 33.,
                21., 46., 67., 39., 88., 75., 112., 98., 98., 106., 88.,
                125., 113., 87., 99., 85., 78., 74., 60., 52., 40., 34.,
                38., 22., 13.,  4.,  0.,  1.,  0.,  0.,  0.,  0.,  0.,
                0.,  0.,  0.,  0.,  0.,  1.]),
          array([-3.75129237, -3.64964098, -3.54798959, -3.44633819, -3.3446868 ,
                -3.24303541, -3.14138401, -3.03973262, -2.93808123, -2.83642984,
                -2.73477844, -2.63312705, -2.53147566, -2.42982427, -2.32817287,
                -2.22652148, -2.12487009, -2.02321869, -1.9215673 , -1.81991591,
                -1.71826452, -1.61661312, -1.51496173, -1.41331034, -1.31165895,
                -1.21000755, -1.10835616, -1.00670477, -0.90505337, -0.80340198,
                -0.70175059, -0.6000992 , -0.4984478 , -0.39679641, -0.29514502,
                -0.19349363, -0.09184223,  0.00980916,  0.11146055,  0.21311195,
                0.31476334,  0.41641473,  0.51806612,  0.61971752,  0.72136891,
                0.8230203 ,  0.92467169,  1.02632309,  1.12797448,  1.22962587,
                1.33127726]),
          <BarContainer object of 50 artists>)
```

