

# Assignment 1

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## Imports

```
In [ ]: import numpy as np
import matplotlib.pyplot as plt
import scipy.stats as st
import astropy as ap
import astroML.stats as aml
import pandas as pd
```

## Question 1

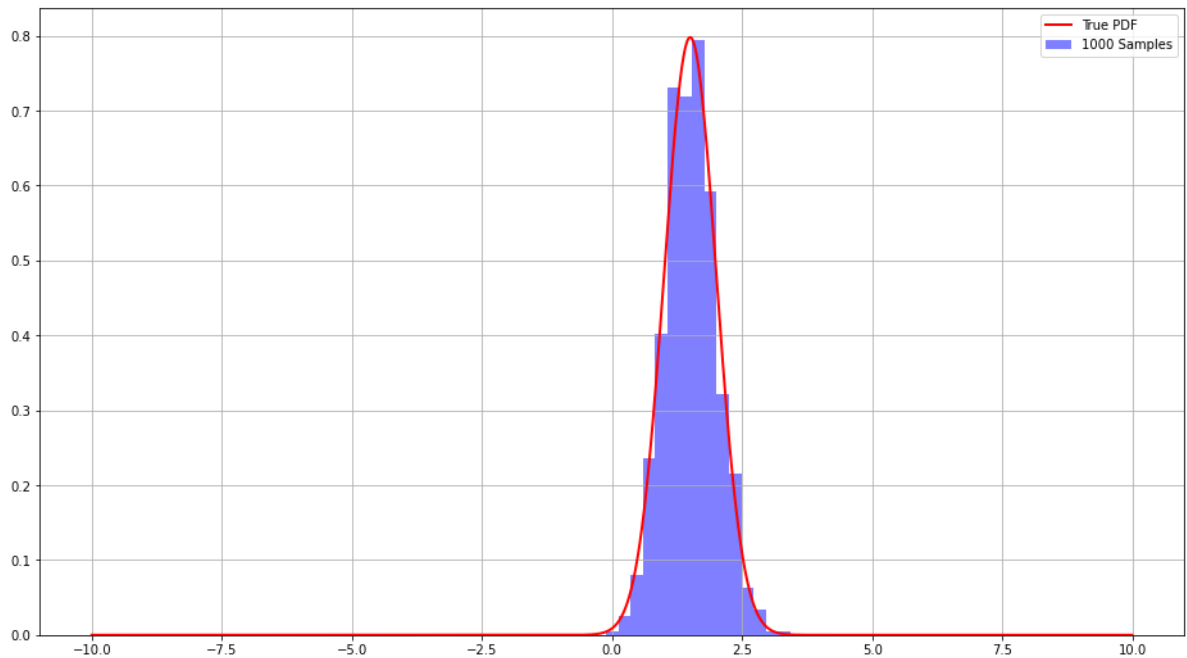
Create 1000 draws from a normal distribution of mean of 1.5 and standard deviation of 0.5.\nPlot the pdf. Calculate the sample mean, variance, skewness, kurtosis as well as standard deviation using MAD and  $\sigma_G$  of these samples.

```
In [ ]: np.random.seed(42)
obj = st.norm(1.5, 0.5)
rv = obj.rvs(size=1000)

print(f"Sample Mean: {rv.mean():.3f}")
print(f"Sample Variance: {float(rv.var()*1000/999.0):.3f}")
print(f"Skewness: {st.skew(rv):.3f}")
print(f"Kurtosis: {st.kurtosis(rv):.3f}")
print(f"MAD: {st.median_abs_deviation(rv):.3f}")
print(f"Standard Deviation: {(1.482*st.median_abs_deviation(rv)):.3f}")
print(f"sigma_G: {aml.sigmaG(rv):.3f}")

plt.figure(figsize=(16, 9))
plt.plot(np.arange(-10, 10, 0.01), obj.pdf(np.arange(-10, 10, 0.01)), c='r', lw=2,
plt.grid()
plt.hist(rv, bins=15, density=True, alpha=0.5, color='b', label='1000 Samples')
plt.legend()
plt.show()
```

```
Sample Mean: 1.510
Sample Variance: 0.240
Skewness: 0.117
Kurtosis: 0.066
MAD: 0.323
Standard Deviation: 0.479
sigma_G: 0.480
```



## Question 2

Plot a Cauchy distribution with  $\mu=0$  and  $\gamma=1.5$  superposed on the top of a Gaussian distribution with  $\mu=0$  and  $\sigma=1.5$ . Use two different line styles to distinguish between the Gaussian and Cauchy distribution on the plot and also indicate these in the legends.

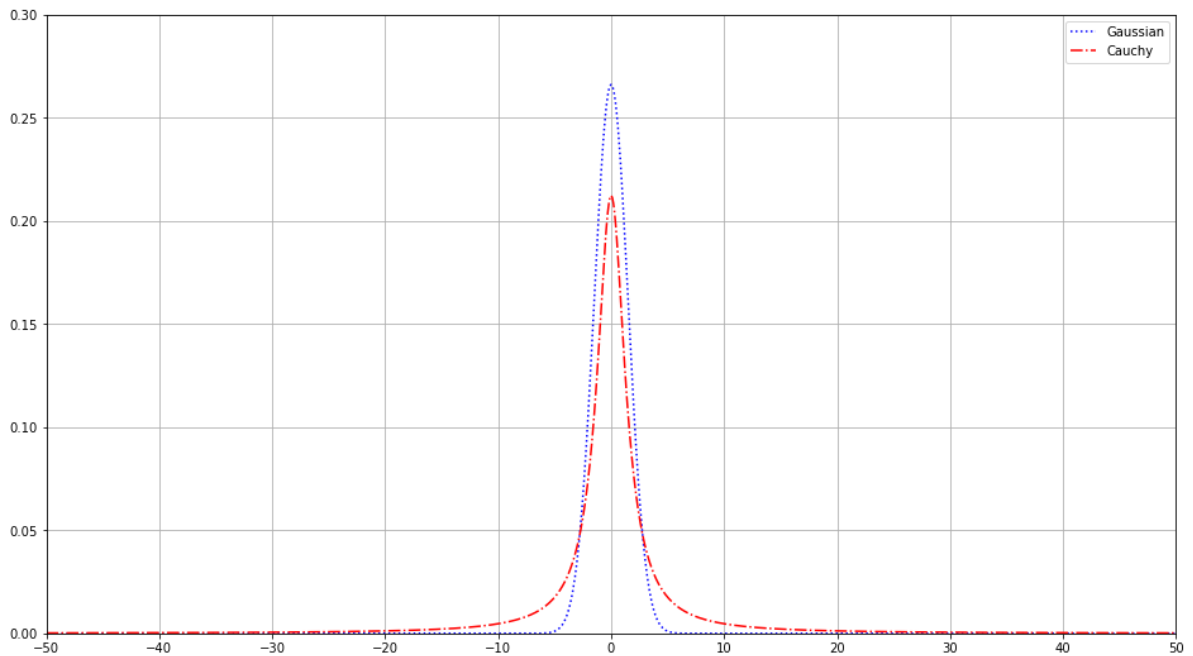
```
In [ ]: pdf_cauchy = st.cauchy(0, 1.5).pdf(np.arange(-50, 50, 0.1)) # Cauchy PDF
pdf_gaussian = st.norm(0, 1.5).pdf(np.arange(-50, 50, 0.1)) # Gaussian PDF

plt.figure(figsize=(16, 9))

# Gaussian Plot
plt.plot(np.arange(-50, 50, 0.1), pdf_gaussian, label="Gaussian", color='b', ls =

# Cauchy Plot
plt.plot(np.arange(-50, 50, 0.1), pdf_cauchy, label="Cauchy", color='r', ls = '-.')

plt.grid(which='both')
plt.ylim(0, 0.3)
plt.xlim(-50, 50)
plt.xticks(np.arange(-50, 51, 10))
plt.legend()
plt.show()
```



## Question 3

Plot Poisson distribution with mean of 5, superposed on top of a Gaussian distribution with mean of 5 and standard deviation of square root of 5. Use two different line styles for the two distributions and make sure the plot contains legends for both of them.

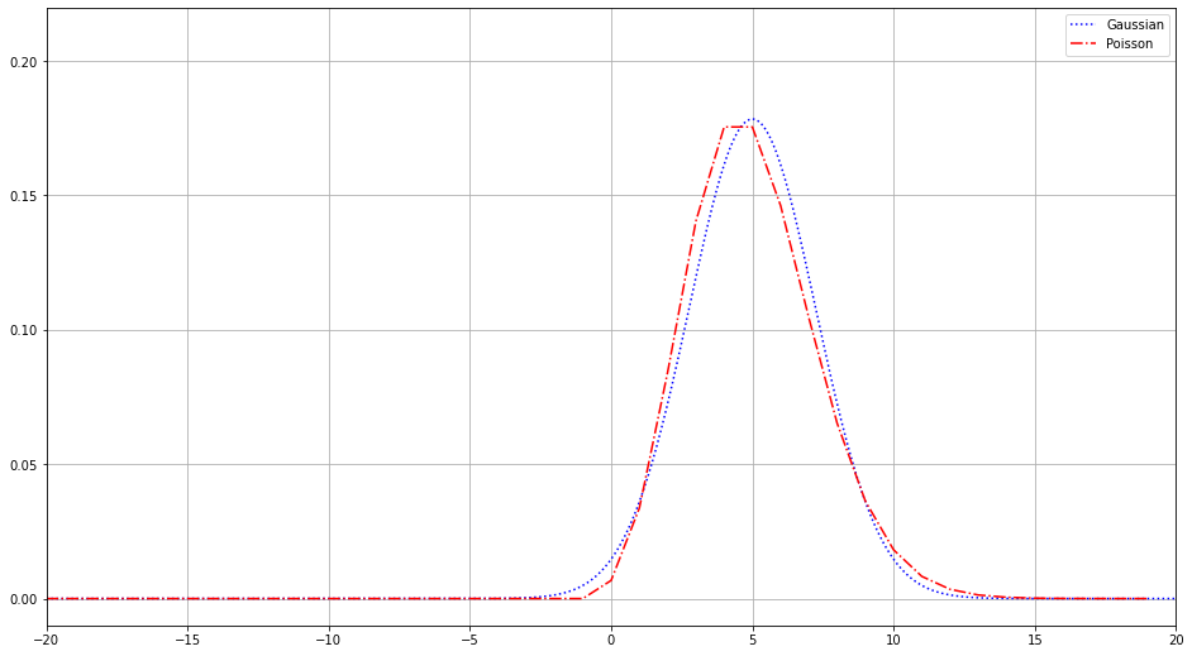
```
In [ ]: pdf_poisson = st.poisson(5).pmf(np.arange(-20, 20, 1)) # Poisson PDF
pdf_gaussian = st.norm(5, np.sqrt(5)).pdf(np.arange(-20, 20, 0.01)) # Gaussian PDF

plt.figure(figsize=(16, 9))

# Gaussian Plot
plt.plot(np.arange(-20, 20, 0.01), pdf_gaussian, label="Gaussian", color='b', ls = 'dotted')

# Poisson Plot
plt.plot(np.arange(-20, 20, 1), pdf_poisson, label="Poisson", color='r', ls = '-.')

plt.grid(which='both')
plt.ylim(-0.01, 0.22)
plt.xlim(-20, 20)
plt.legend()
plt.show()
```



## Question 4

The following were the measurements of mean lifetime of K meson (as of 1990) (in units of  $10^{-10}$  s) :  $0.8920 \pm 0.00044$ ;  $0.881 \pm 0.009$ ;  $0.8913 \pm 0.00032$ ;  $0.9837 \pm 0.00048$ ;  $0.8958 \pm 0.00045$ . Calculate the weighted mean lifetime and uncertainty of the mean.

```
In [ ]: lifetime = np.asarray([0.892, 0.881, 0.8913, 0.9837, 0.8958])
error = np.asarray([0.00044, 0.009, 0.00032, 0.00048, 0.00045])

weighted_mean = np.sum(np.divide(lifetime, np.square(error)))/np.sum(np.divide(1, np.square(error)))

uncertainty = 1/np.sqrt(np.sum(np.divide(1, np.square(error))))

print(f"Weighted Mean: {weighted_mean:.3f}")
print(f"Uncertainty of Weighted Mean: {uncertainty:.3f}")
```

Weighted Mean: 0.909

Uncertainty of Weighted Mean: 0.000

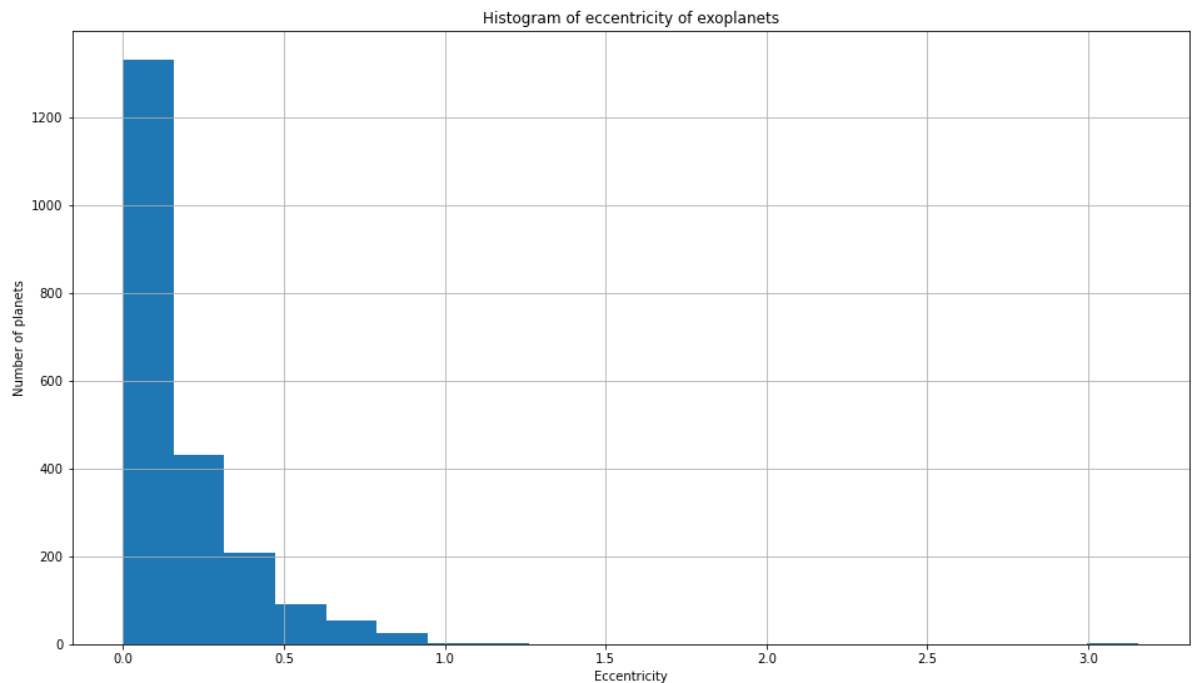
## Question 5

Download the eccentricity distribution of exoplanets from the exoplanet catalog <http://exoplanet.eu/catalog/>. Look for the column titled e, which denotes the eccentricity. Draw the histogram of this distribution. Then redraw the same histogram after Gaussianizing the distribution using Box-transformation either using `scipy.stats.boxcox` or from first principles using the equations shown in class or in arXiv:1508.00931. Note that exoplanets without eccentricity data or with  $e=0$ , can be ignored.

```
In [ ]: # Reading data
df = pd.read_csv("exoplanet.eu_catalog.csv")
print(df['eccentricity'].count())
```

2145

```
In [ ]: # Plotting original data
plt.figure(figsize=(16, 9))
plt.hist(df['eccentricity'], bins=20)
plt.grid(which='both')
plt.xlabel("Eccentricity")
plt.ylabel("Number of planets")
plt.title("Histogram of eccentricity of exoplanets")
plt.show()
```



```
In [ ]: # Removing data with e=0
for x in df.index:
    if not(df.loc[x, "eccentricity"] > 0):
        df.drop(x, inplace = True)
print(df['eccentricity'].count())
```

1704

```
In [ ]: # Plotting Gaussianized data(Using Box-Cox transformation)
plt.figure(figsize=(16, 9))
plt.hist(st.boxcox(df['eccentricity'])[0], bins=20)
plt.xlabel("Eccentricity")
plt.ylabel("Number of planets")
plt.title("Box-Cox transformation of Histogram eccentricity of exoplanets")
plt.show()
```

