

Assignment 2 - Report

Team Members:

Sathyannarayanan R - S20180010154

Darshan G - S20180010046

Jaswanth K - S20180010068

Ejurothu Pavan Sai Santhosh - S20180010053

Part-1 Simulating Chi-Square Distribution

The Chi-Squared Distribution (χ^2 -distribution) is the distribution of the sum of squares of n standard normal random variables, where n is the degrees of freedom of the chi-square random variable ($df = n$).

Mean of the distribution = df

Variance of the distribution = $2 * df$

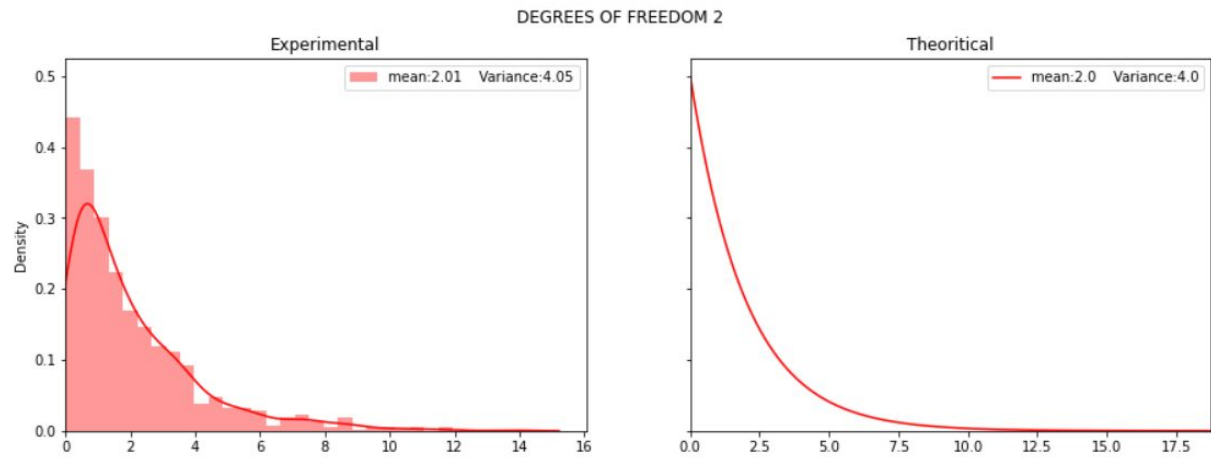
Values of Mean and Variance found by simulation and theoretical:

Degrees of Freedom, (df)	Simulated		Theoretical	
	Mean	Variance	Mean	Variance
1	1.0	1.89	1	2
2	2.01	4.05	2	4
3	2.98	5.98	3	6
4	4.07	8.76	4	8
5	5.06	10.85	5	10
6	6.12	12.69	6	12
7	7.05	14.32	7	14
8	8.13	16.6	8	16
9	9.11	18.04	9	18
10	10.12	20.27	10	20

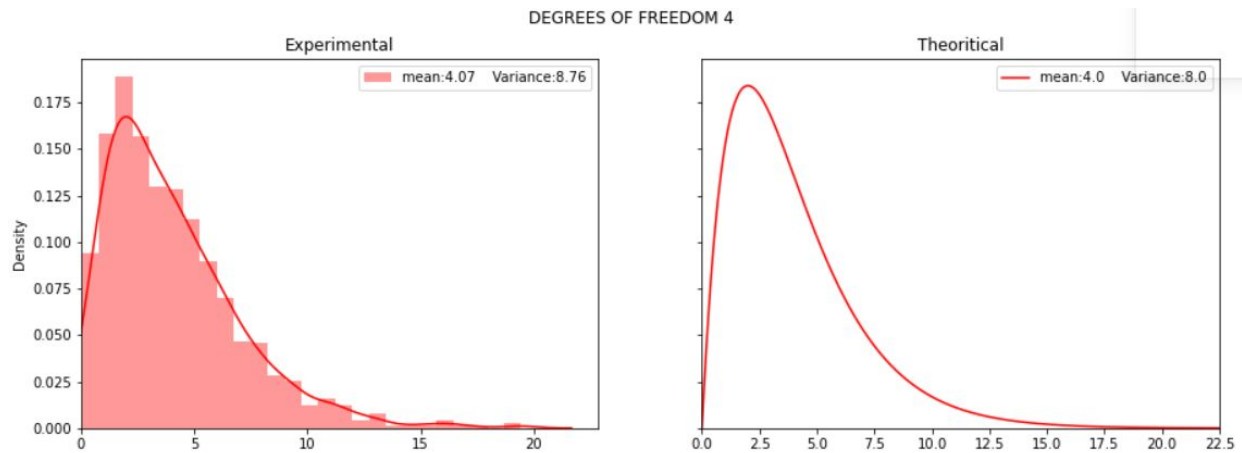
Chi-Square Distribution for the above tabulation:

Random variables were generated for the above parameters and the pdf formula $f(x)$ was applied to each of those points to get the corresponding distribution

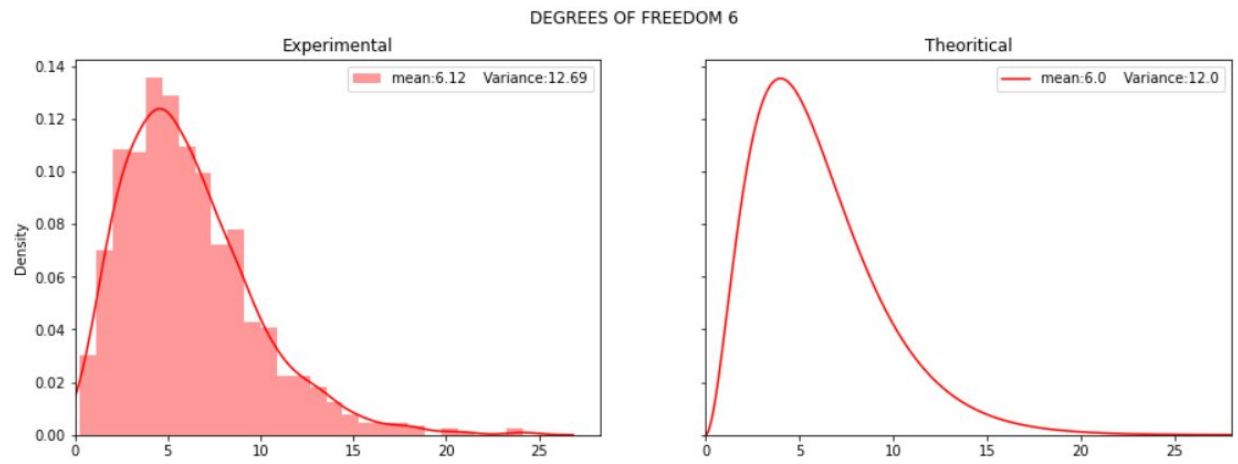
$DF=2$



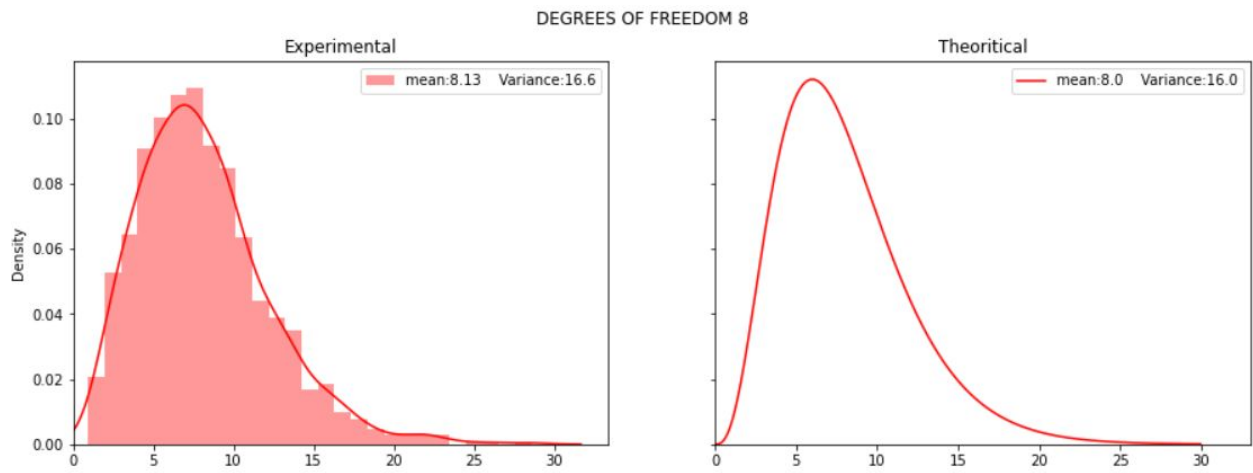
$DF=4$



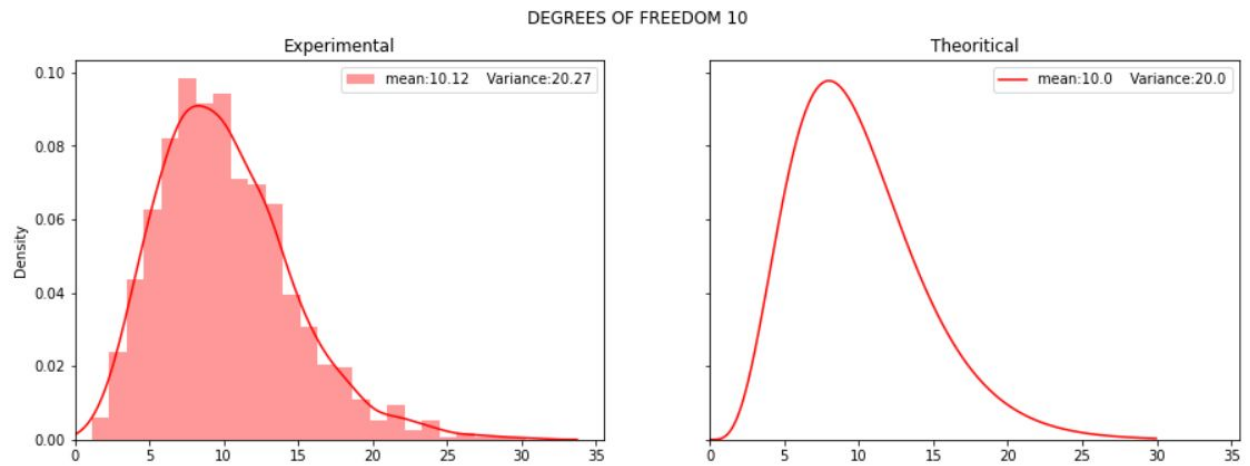
DF=6

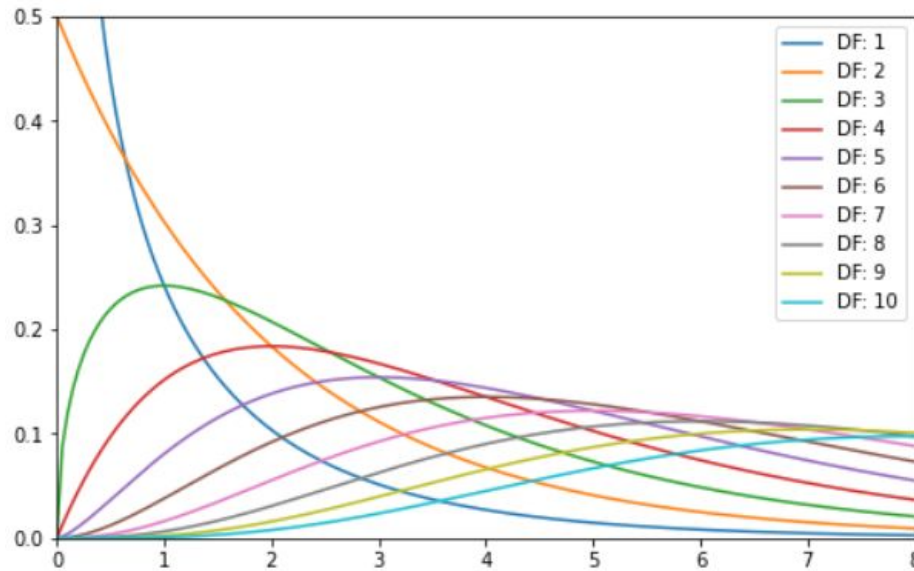


DF=8



DF=10





Conclusion:

The Chi-Square Distribution was successfully simulated by taking random samples and the parameters are found to be approximately equal to the theoretical values.

The mean of a Chi-Square distribution is equal to the DF, the variance is equal to two times DF.

The Chi-Square PDF looks more and more normal with an increase in degrees of freedom.

Part 2 - Simulating t-distributions

A t-distribution can be expressed as
$$\frac{z}{\sqrt{\chi^2/n}}$$

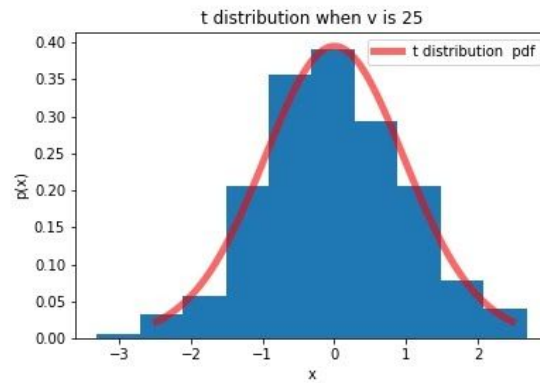
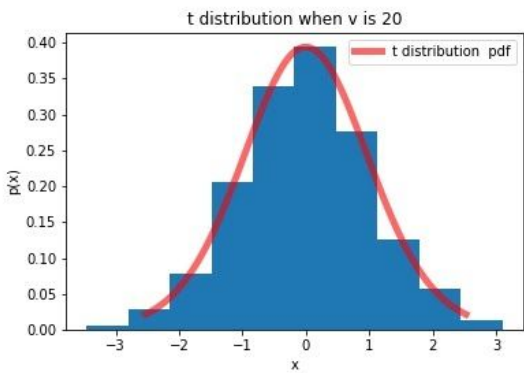
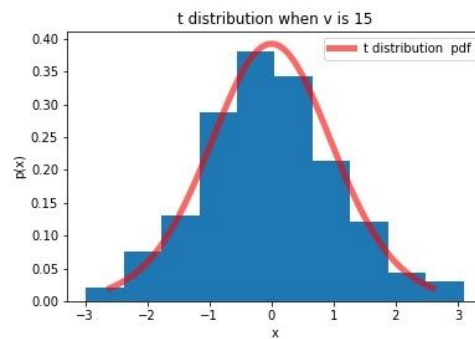
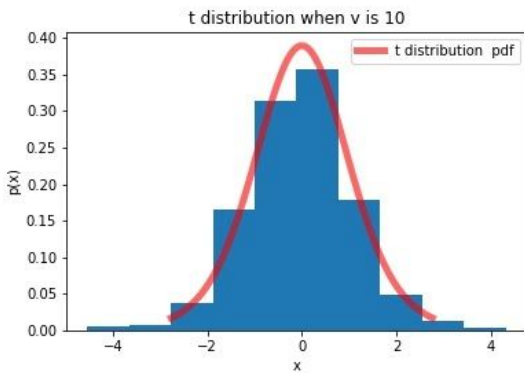
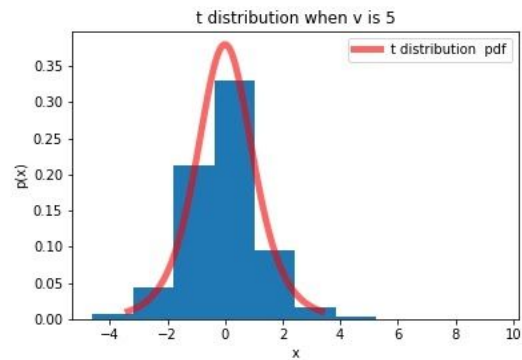
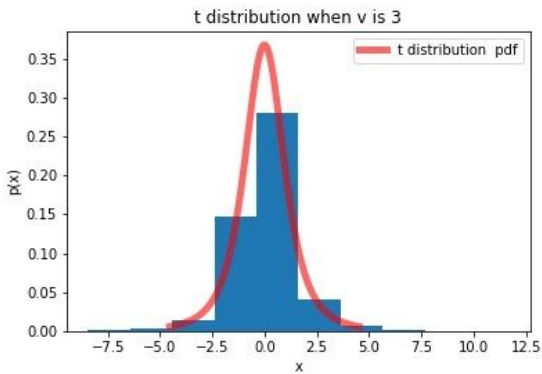
Where Z is a standard normal distribution random variable and χ^2 is a chi-squared distribution with 'n' degrees of freedom.

Mean of a t-distribution is zero and variance is $n/(n-2)$

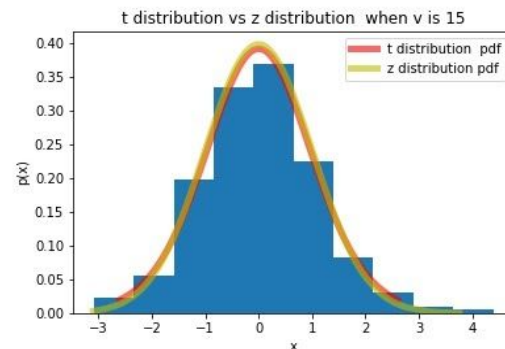
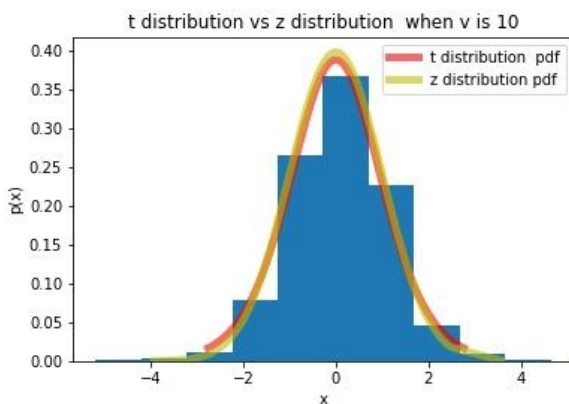
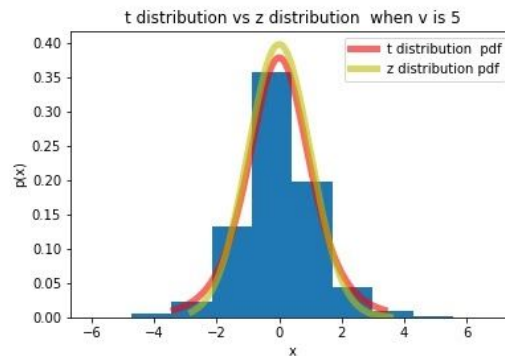
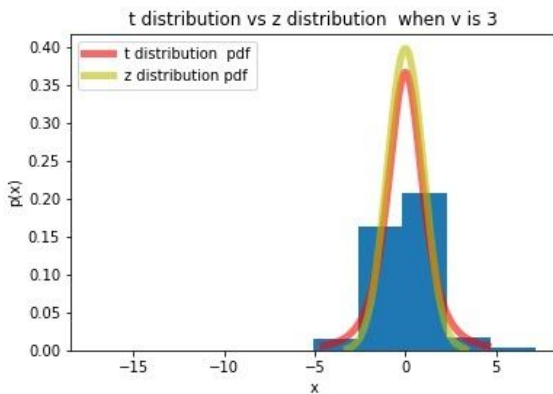
Method used :

Random variables generated for Z-distribution and Chi-squared were used for each degrees of freedom and with the help of the above formula t-distribution was simulated (v - Degrees of freedom)

Plots showing t-distribution with different degrees of freedom



Plots comparing Standard Normal Distribution with t-distribution



Conclusion

In terms of mean both Standard normal distribution and t-distribution are centered around zero. Also, t-distribution has fatter tails than normal distribution. This means that the probability of getting values far away from the mean is larger with a t-distribution than a normal distribution. Based on the observations in the above plots above, as the degrees of freedom increase t-distribution looks similar to the normal distribution

Part-3 Simulating F - Distributions using X^2 :

The F distribution is the ratio of two chi-square distributions with degrees of freedom n_1 and n_2 , respectively, where each chi-square has first been divided by its degrees of freedom.

$$F(n_1, n_2) = (U_1/n_1)/(U_2/n_2)$$

Here, U_1 has X^2 distribution with n_1 degrees of freedom. U_2 has X^2 distribution with n_2 degrees of freedom.

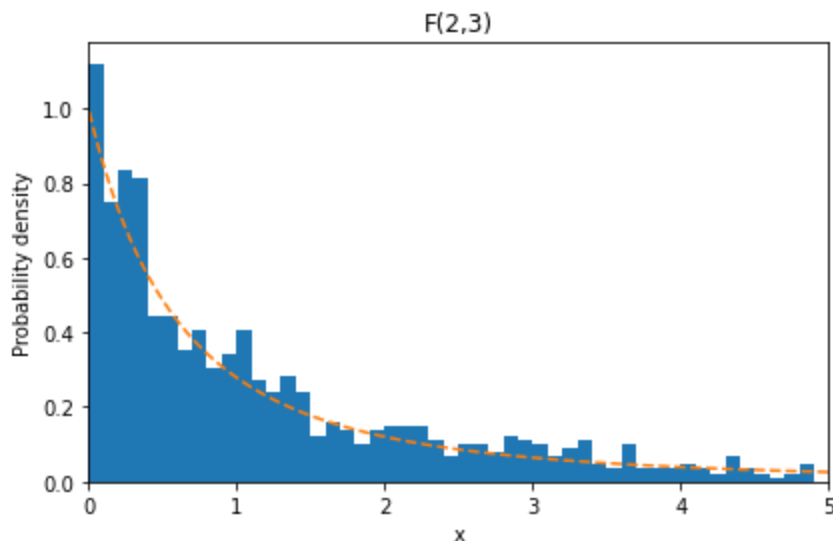
Method:

$F(2,3)$ is simulated using $X^2(2)$ and $X^2(3)$ taken from part-1.

We divide chisq2 and chisq3 by their degrees of freedom and then the ratio is taken to find $F(2,3)$.

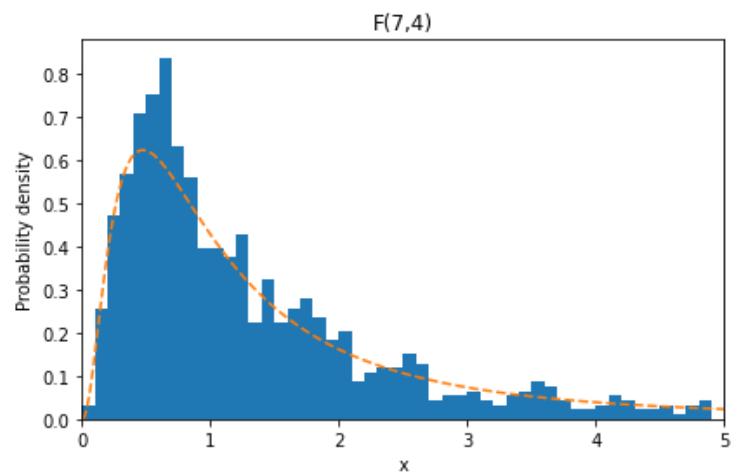
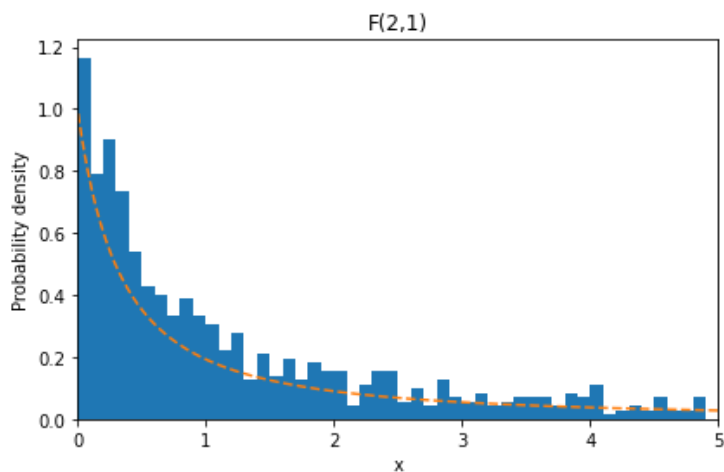
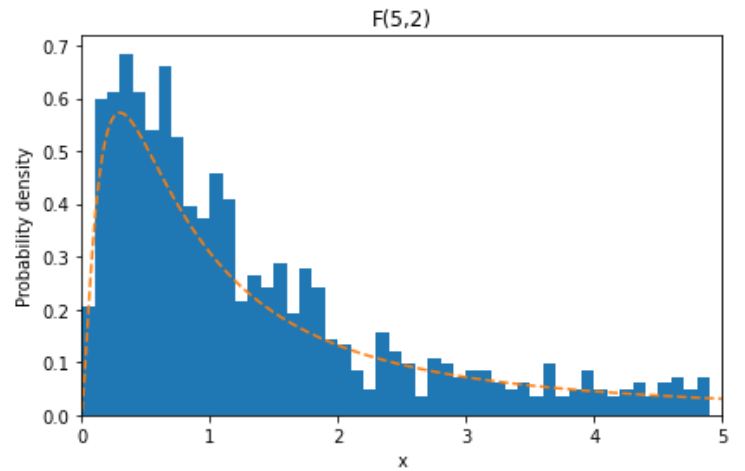
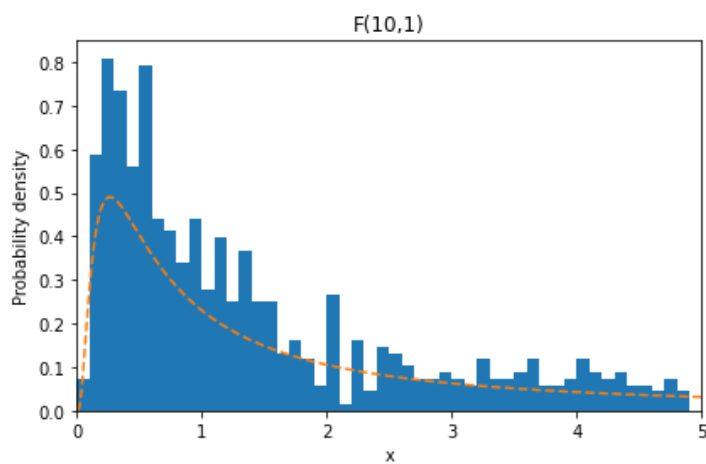
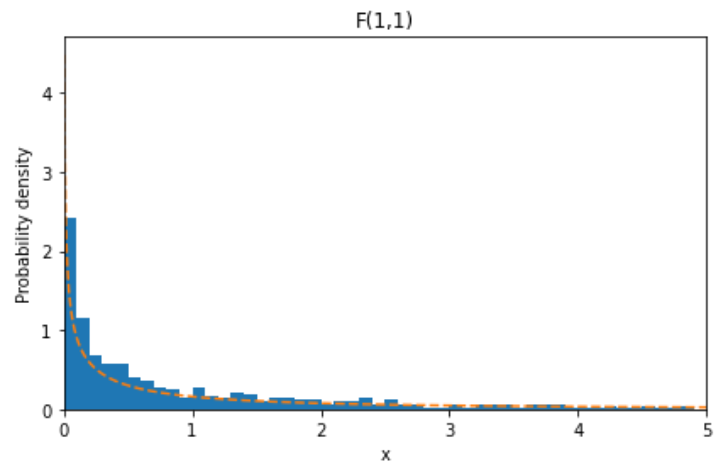
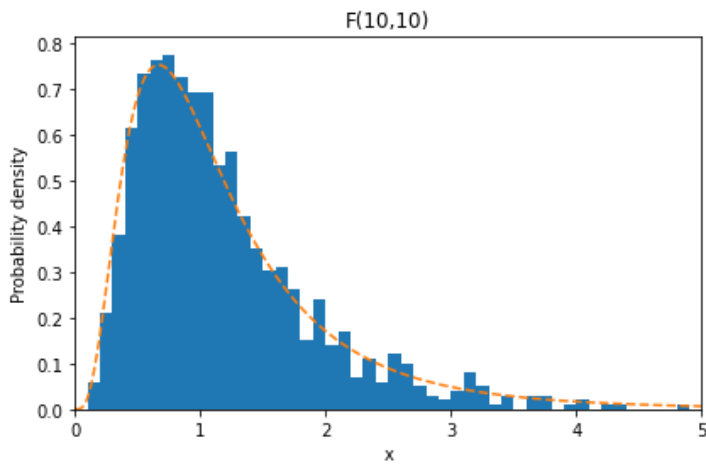
```
dfn = 2
dfd = 3
limit = 5
chn=chisq2/dfn
chd= chisq3/dfd
np.random.shuffle(chd)

x=np.divide(chn,chd)
```



Similarly, the following F- distributions have been simulated and plotted using respective χ^2 - distributions:

*Orange lines(---) denote theoretical pdf



Conclusion:

Since theoretical pdf matches with simulated ones, it can be concluded that F-distribution is successfully simulated.

Part-4 Check Normality:

A multivariate normal distribution was simulated with 3 variables for 100 observations

Mean used to generate random variables = [5,3,7]

Covariance matrix used to generate random variables= [[4,-1,0],[-1,4,2],[0,2,9]]

A) Univariate Normality Check using Q-Q Plots:

Q-Q plots are used to check univariate normality

Method to plot Q-Q plots

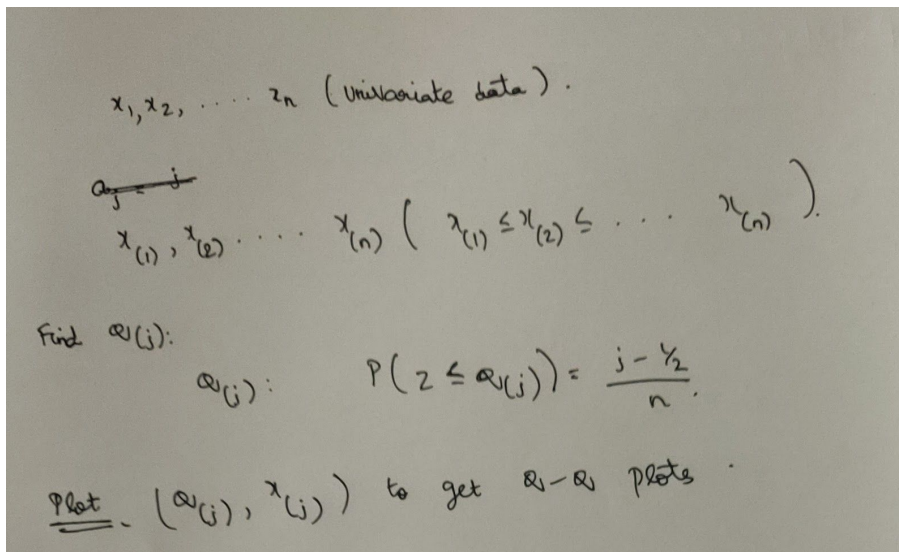
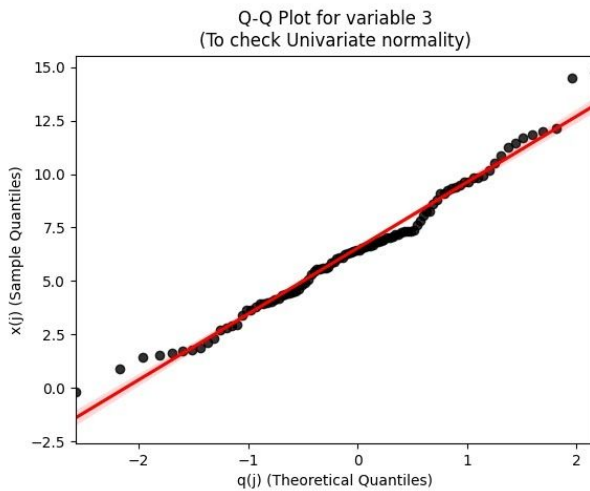
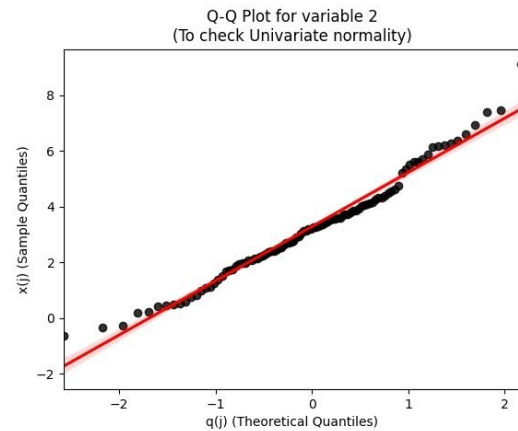
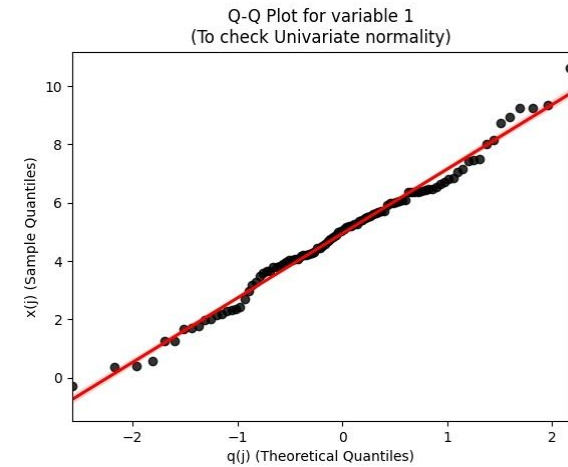


Figure Q-Q plots for each variable:



Observation:

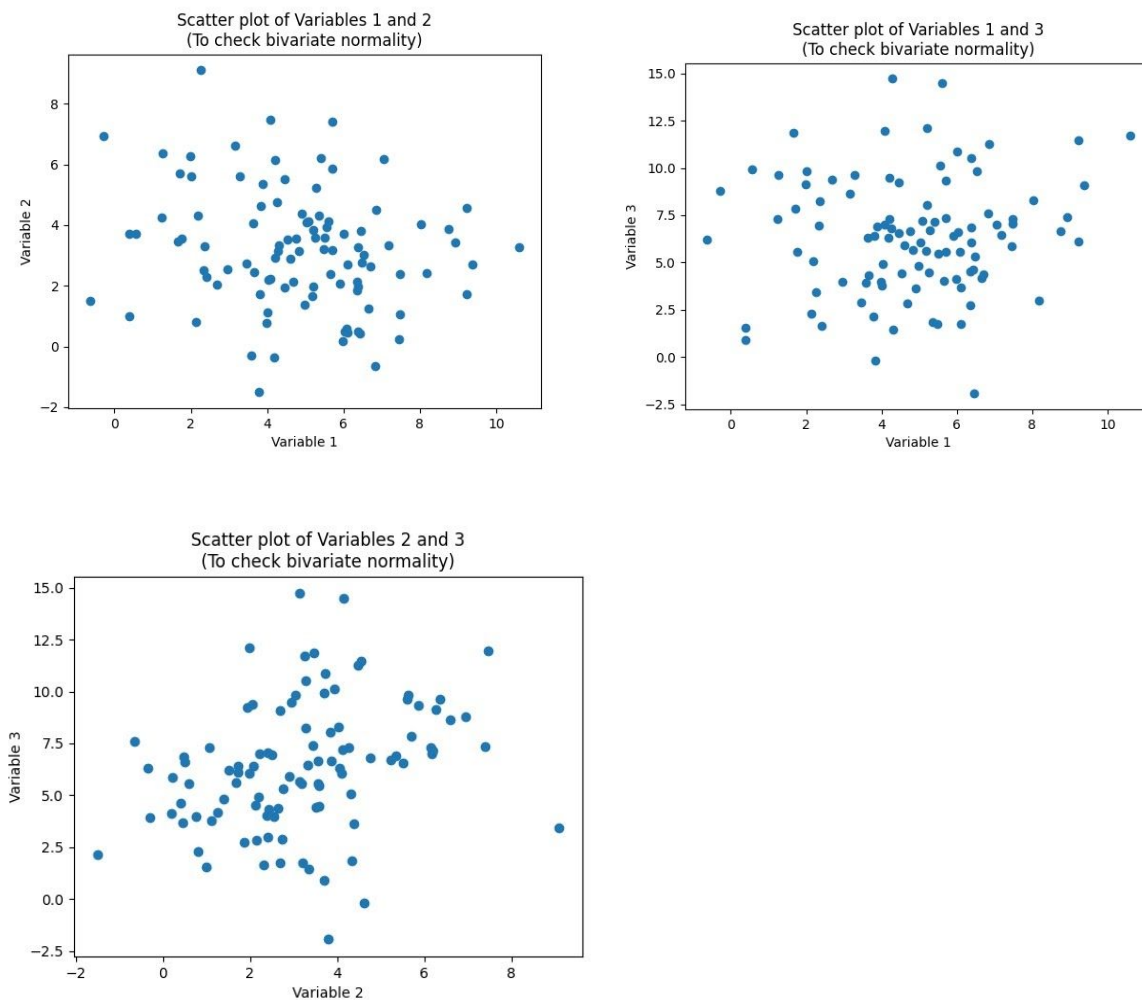
As observed all Q-Q plots have points which can be fitted with a straight line, so this shows that the given univariate data is normal

B)

Bivariate Normality Test:

To do bivariate normality test, two variables are taken at a time their scatter plots are plotted

Figure: Scatter plots of two variables at a time



Observation:

The scatter plots form an elliptical shape which confirms that two variables taken at a time are bivariate normals

Multivariate Normality Test:

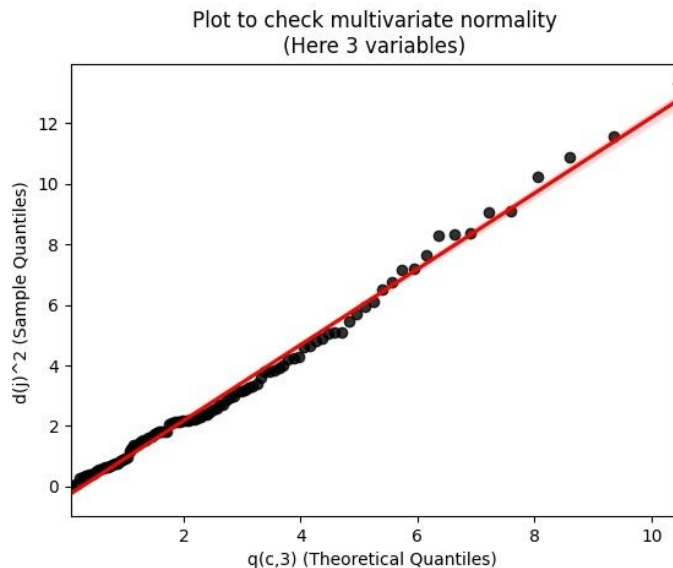
Method used:

$$\begin{aligned} & x_1, x_2, \dots, x_n \text{ (Multivariate data)} \\ & d_j^2 = (x_j - \bar{x})^T S^{-1} (x_j - \bar{x}) \quad [\text{This follows the } \chi^2 \text{ distribution}] \\ & d_1^2, d_2^2, \dots, d_n^2 \\ & \text{Sort it,} \\ & d_{(1)}^2, d_{(2)}^2, \dots, d_{(n)}^2 \quad [d_{(1)}^2 \leq d_{(2)}^2 \leq \dots \leq d_{(n)}^2] \\ & \alpha_{c,p}((j - \frac{1}{2})/n) = \chi_p^2(n - j + \frac{1}{2}/n) \\ & \text{Plot } (\alpha_{c,p}((j - \frac{1}{2})/n), d_{(j)}^2) \end{aligned}$$

Each d_j^2 follows chi-squared distribution and using chi-squared table $q(c,p)$ is found which is plotted along with d_j^2 values

Figure:

Plot to check Multivariate Normality:



Observation:

The scatter plot forms a straight line which shows that the simulated Multivariate data (here Trivariate) is normal