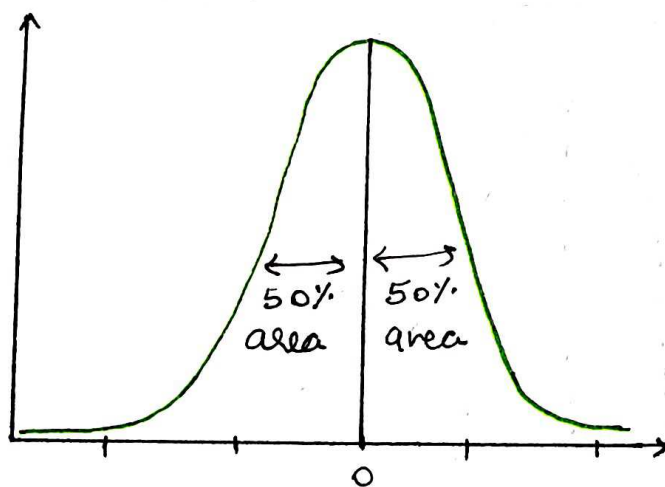


Properties

1. Symmetry

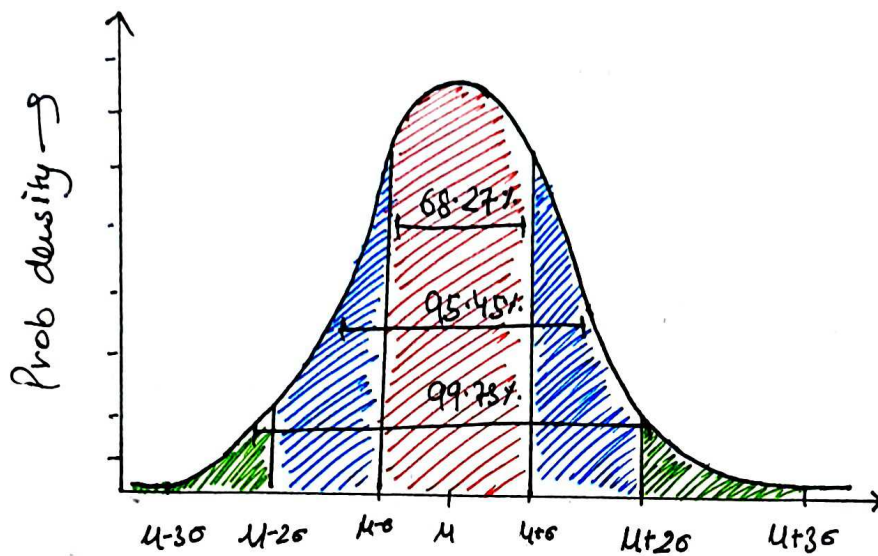
The normal distⁿ is symmetric about its mean which means that the probability of observing a value above the mean is the same as the probability of observing a value below the mean. The bell-shaped curve of the normal distribution reflects this symmetry.



2. Measure of Central Tendencies are equal
Mean, Median, Mode

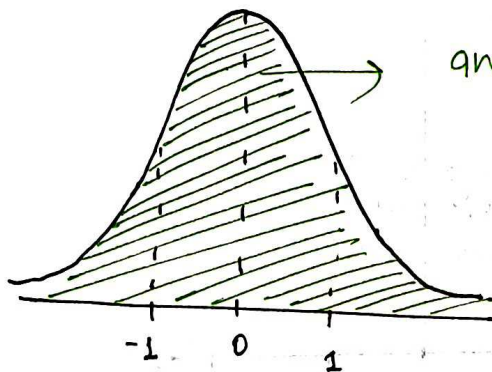
3. Empirical Rule

The normal distⁿ has well-known empirical rule also called 68-95-99.7 rule, which states that approximately 68% of the falls within one standard deviation of the mean, about 95% of the data falls within two standard deviation of the mean, and about 99.7% of the data falls within three std of the mean.



68-95-99.7 Rule

4. The Area under the curve

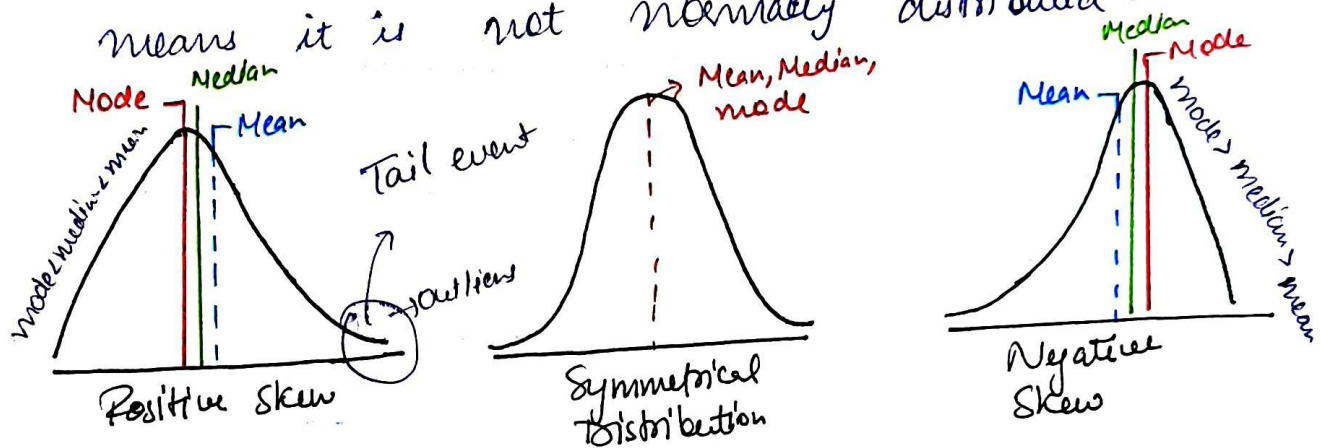


area under curve is 1 for every PDF

Skewness

* What is skewness?

A normal distribution is a bell-shaped, symmetrical distribution with a specific mathematical formula that describes how the data is spread out. Skewness indicates that the data is not symmetrical, which means it is not normally distributed.



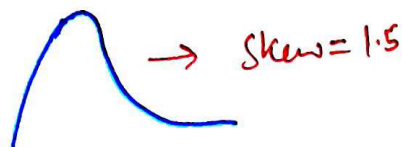
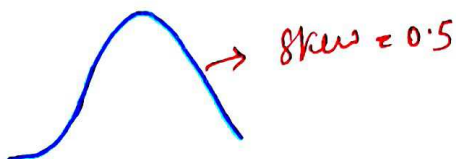
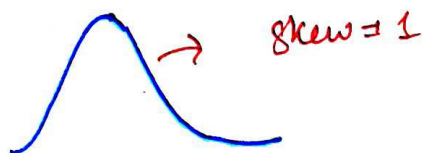
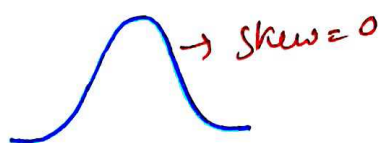
The greater the skew the greater distⁿ betⁿ mode, median, mean.

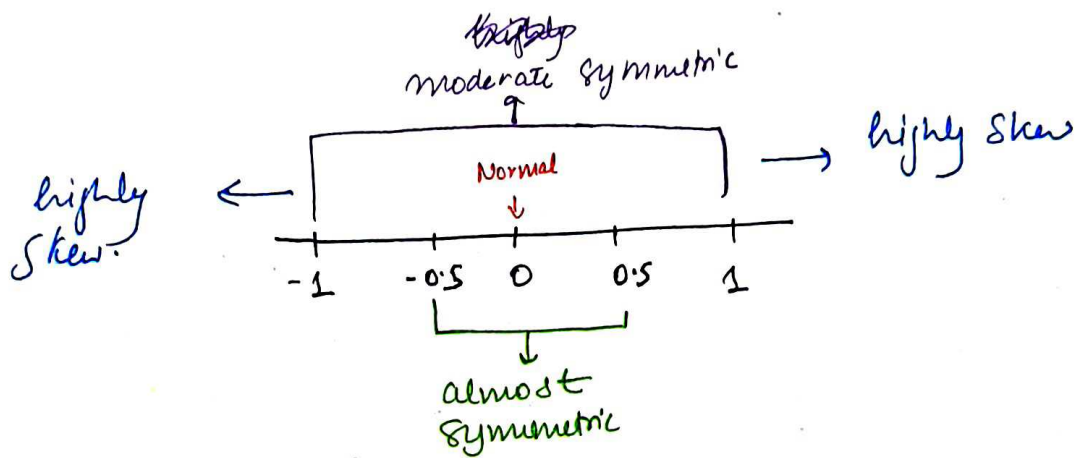
* How skewness is calculated?

$$\frac{n}{(n-1)(n-2)} \sum \left(\frac{(x-\bar{x})^3}{s} \right) \rightarrow \text{3rd Moment}$$

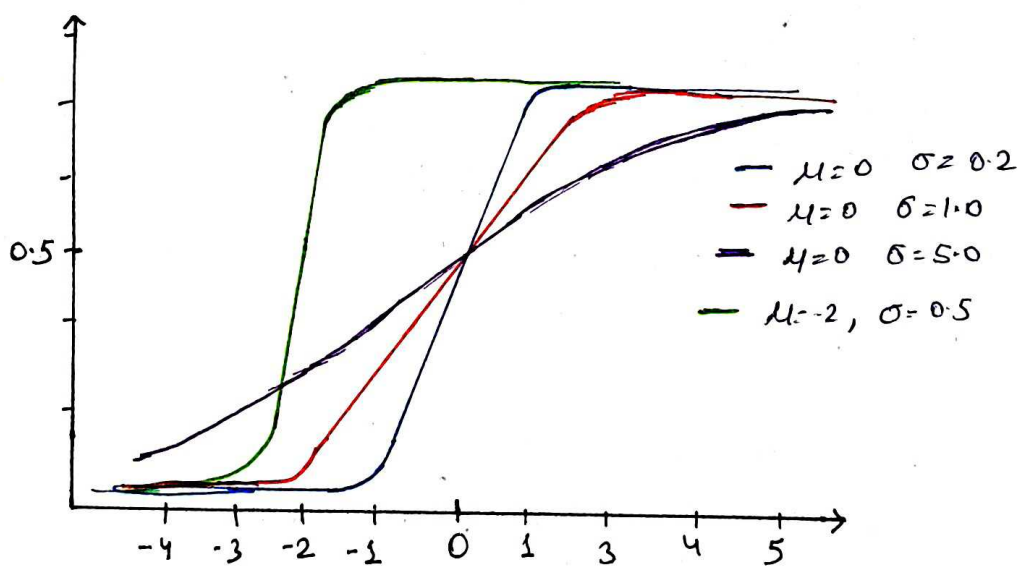
1st moment \rightarrow mean

2nd moment \rightarrow Variance



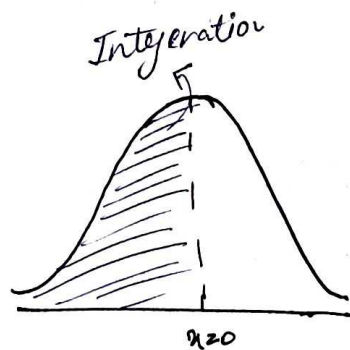


CDF of Normal Distribution



$$f(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$$

$$= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{(t-\mu)^2}{2\sigma^2}} dt$$



Use in data science

→ Outlier detection

→ Assumption on data for ML algo \Rightarrow LR and GMM

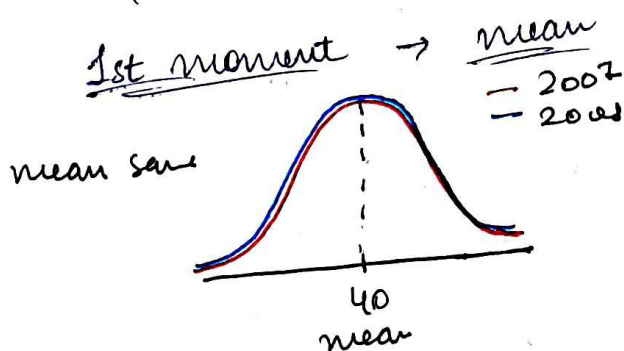
→ Hypothesis testing

→ Central Limit Theorem

Kurtosis

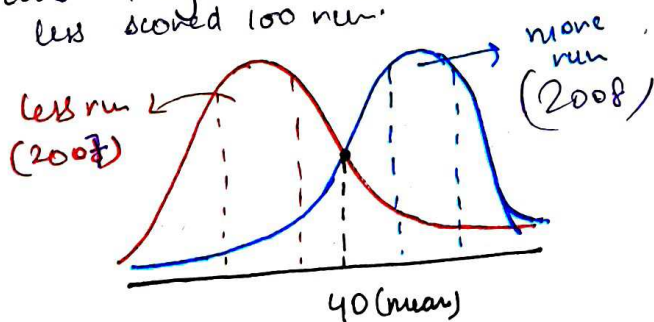
* What is kurtosis?

Kurtosis is the 4th statistical moment. In probability theory and statistics, kurtosis (meaning "curved, arching") is a measure of the "tailedness" of the probability distribution of a real-valued random variable. Like skewness, kurtosis describes a particular aspect of a probability distribution.



2nd moment \rightarrow Std

Sachin more consistent in 2008
bcz most of the near the mean.
and they are less out in 0 run.
less scored 100 run.



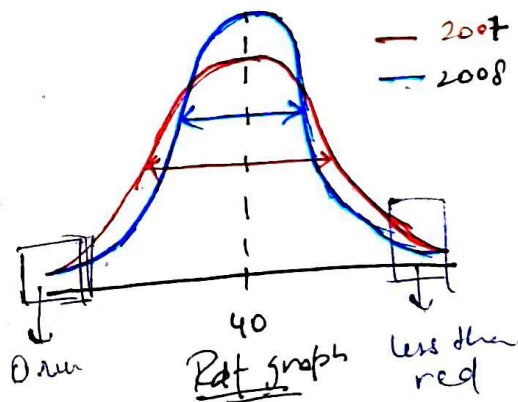
4th moment \rightarrow Kurtosis

In 2008, Sachin out maximum time at 0 and also scored maximum time. As compare to 2007.

Batsman \rightarrow Sachin

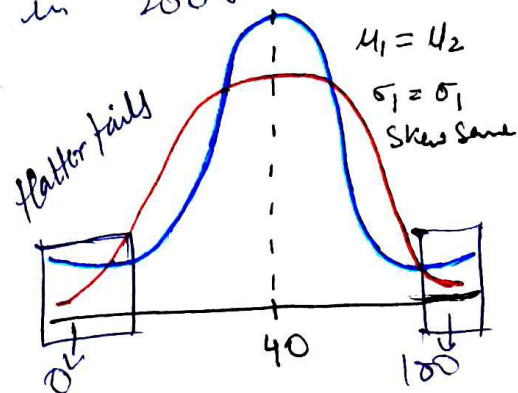
2007 \rightarrow 100 matches \rightarrow 40 avg

2008 \rightarrow 100 matches \rightarrow 40 avg



3rd moment \rightarrow Skew

Sachin scored maximum run in 2008 and most of the less run in 2007.



Formula

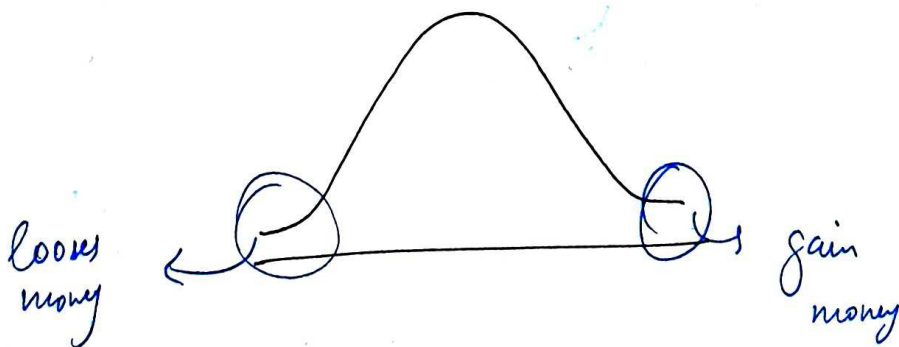
$$\left\{ \frac{n * (n+1)}{(n-1) * (n-2) * (n-3)} * \sum_j^n \left(\frac{x_i - \bar{x}}{s} \right)^2 \right\} - \frac{3 * (n-1)^2}{(n-2) * (n-3)}$$

Practical use-case

In finance, kurtosis risk refers to the risk associated with the possibilities of extreme outcomes or "fat tails" in the distribution of returns of a particular asset or portfolio.

If a distribution has higher kurtosis, it means that there is a higher likelihood of extreme events occurring, either positive or negative, compared to a normal distribution.

In finance, kurtosis risk is important to consider because it indicates that there is a greater probability of larger losses or gains occurring, which can have significant implications for investors. As a result, investors may want to adjust their investment strategies to account for kurtosis risk.



Excess Kurtosis & Types

Excess kurtosis measure of how much tailed, flatter or thinner tail of our graph or distribution as compared to given normal distribution.

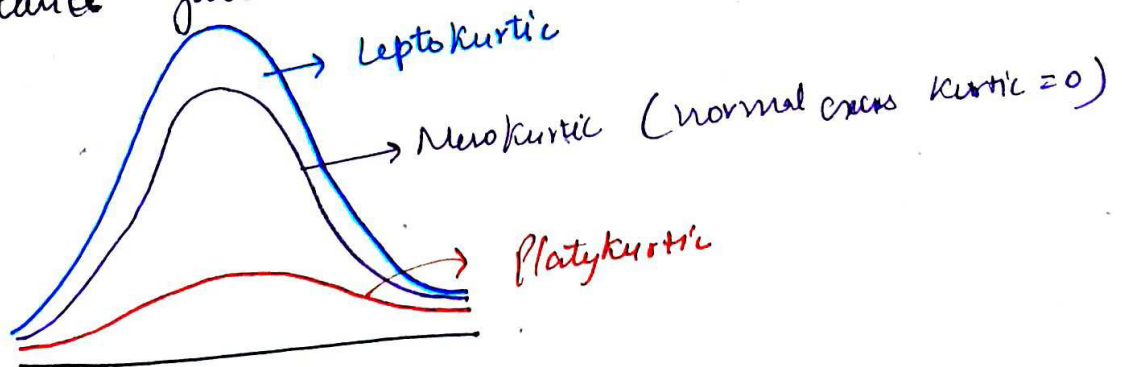
Normal distribution's kurtosis is 0. Excess kurtosis calculated by subtracting 3 from the sample kurtosis coefficient.

Leptokurtic

A distribution with positive excess kurtosis is called leptokurtic. "Lepto" mean "slender". In term of shape, a leptokurtic distribution has fatter tails. This indicates that there are more extreme values or outlier in the distribution.

$$\boxed{\text{Kurtosis} - 3 > 0}$$

Example - Assets with positive excess kurtosis are riskier and more volatile than those with a normal distribution, and they may experience sudden price movements that can result in significant gains or losses.



Platykurtic

A distribution with negative excess kurtosis is called platykurtic. "Platy" means "~~flat~~ bread". In terms of shape, a platykurtic distribution has thinner tails. This indicates that there are fewer - extreme values or outliers in the distribution.

Assets with negative excess kurtosis are less risky and less volatile than those with a normal distribution, and they may experience more gradual price movements that are less likely to result in large gains or losses.

$$\boxed{\text{kurtosis} - 3 < 0}$$

Mesokurtic

Distribution with zero excess kurtosis are called mesokurtic. The most prominent example of a mesokurtic distribution is the normal distribution family, regardless of the value of its parameters.

Mesokurtic is a term used to describe a distribution with a excess kurtosis of 0, indicating that it has the same degree of "peakedness" or "flatness" as a normal distribution.

Example: In finance, a mesokurtic distribution is considered to be the ~~total~~ ideal distribution for assets or portfolio, as it represents a balance between risk and return.

Q-Q Plot

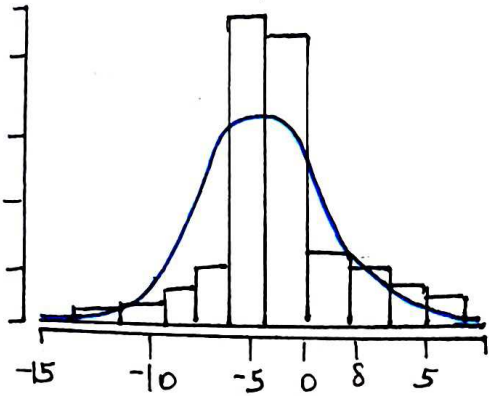
* How to find if a given distribution is normal or not?

Visual inspection: One of the easiest ways to check for normality is to visually inspect a histogram or a density plot of the data. A normal distribution has a bell shaped curve, which means that the majority of the data falls in the middle, and the tails taper off symmetrically. If the distribution looks approximately bell shaped, it is likely to be normal.

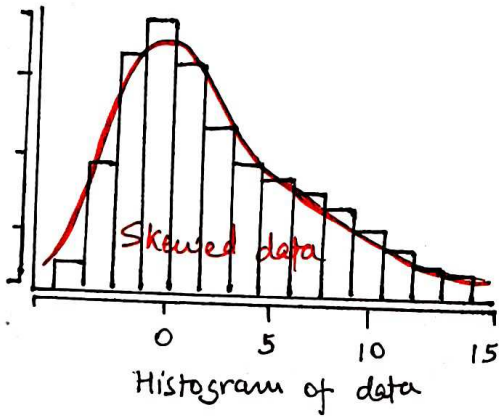
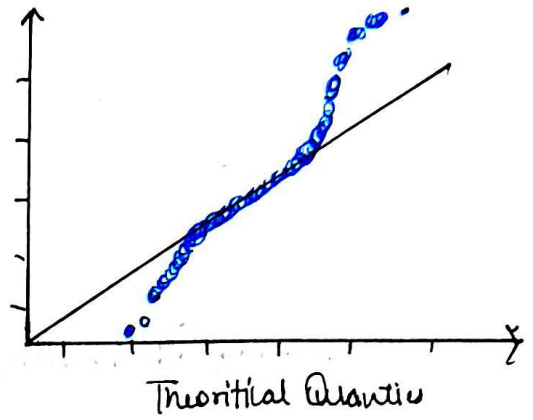
QQ Plot: Another way to check for normality is to create a normality test. A normal probability plot plots the observed prob of the data against the expected values of a normal distn. If the data points fall along a straight line, the distribution is likely to be normal.

Statistical tests: There are several statistical tests that can be used to test for normality, such as the Shapiro-Wilk test, the Anderson-Darling test and the Kolmogorov-Smirnov test. These tests compare the observed data to the expected value of a normal distn and provide a p-value that indicates whether the data is likely to be normal or not. A p-value less than the significance level (usually 0.05) suggest that the data is not normal.

Data too peaked in middle

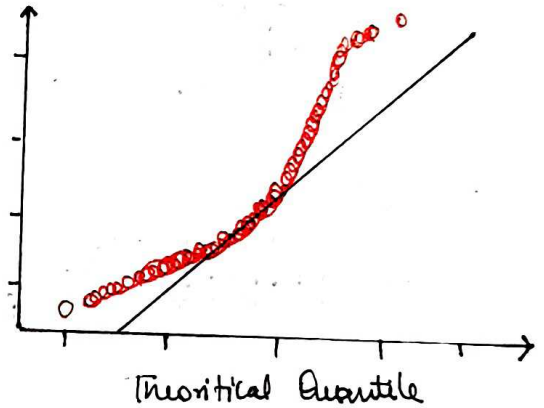


Histogram of data

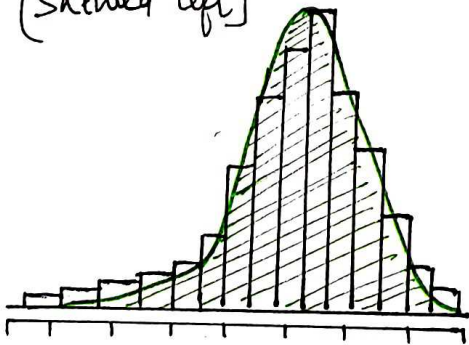


Skewed data

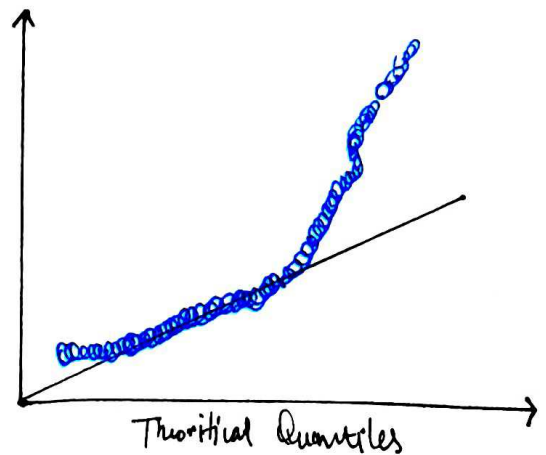
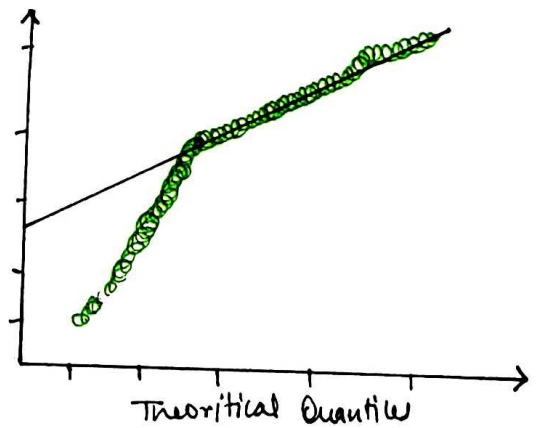
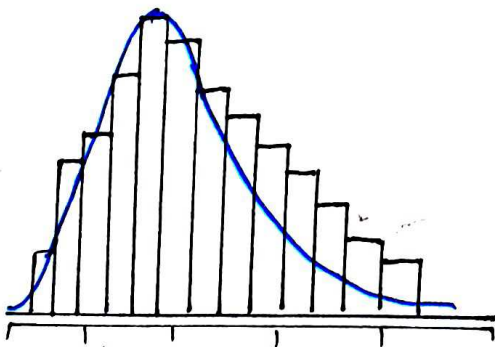
Histogram of data



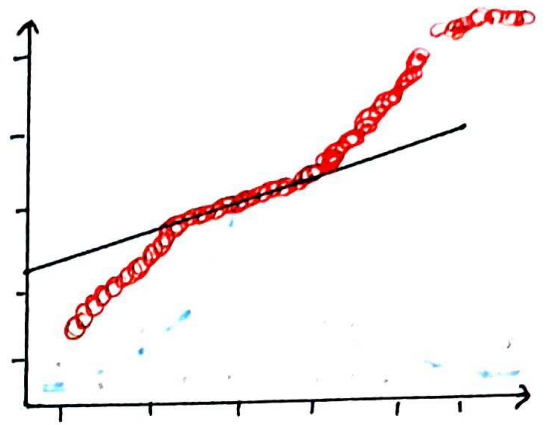
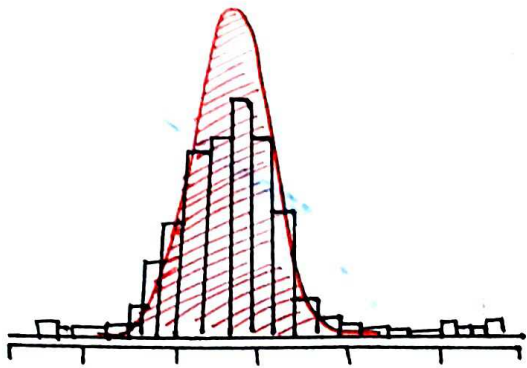
[Skewed left]



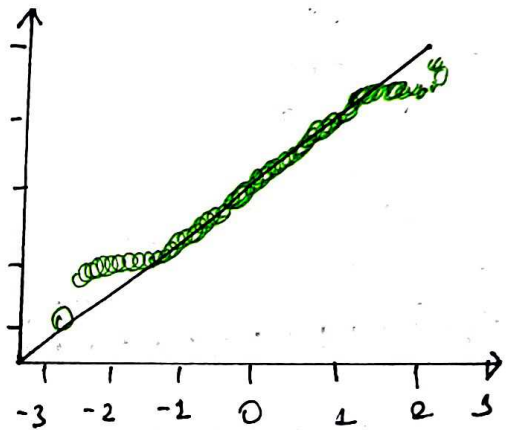
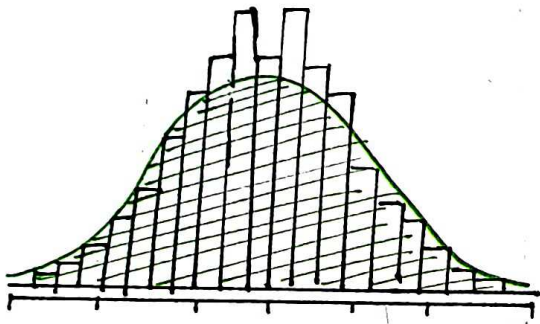
[Skewed Right]



{Fat Tails}



{Thin Tails}



* Does QQ plot only detect normal distribution?