

Elastic Net [L_1 & L_2 Norm]

$$\text{Cost func} = \frac{1}{m} \sum_{i=1}^m (h_0(x^{(i)}) - y^{(i)})^2 + \lambda_1 \sum_{i=1}^m (\text{slope})^2 + \lambda_2 \sum_{i=1}^m |\text{slope}|$$

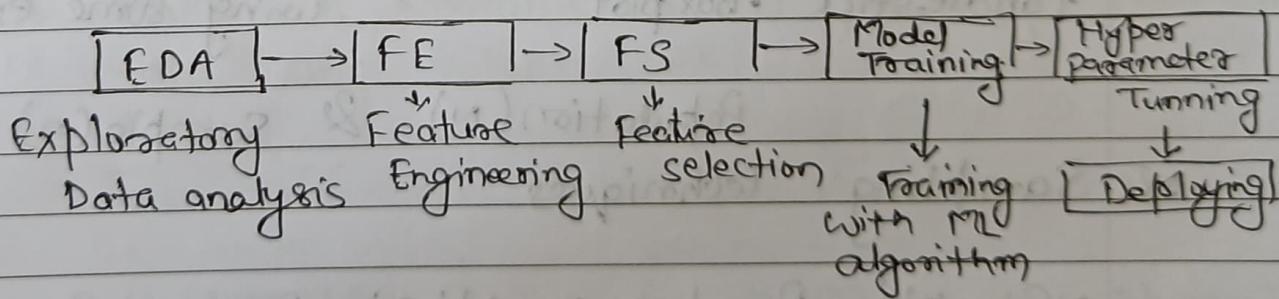
Ridge Lasso

Statistics

- Find the average size of shark throughout the world
- Amazon Big Billion Day sale → which month should you choose.

Life cycle of Data science project

Requirement
Gathering



Statistics will be used almost every stage in the life cycle of a Data science project

Average \Rightarrow Measure of central Tendency.

↓
Descriptive stats

Statistics: Statistics is the science of collecting, organising and analysing the data.

Data: "facts or pieces of information"

statistics

↓
Descriptive stats [EDA + FE] Inferential stats

It consists of organising and summarizing the data. It consists of collecting sample data and

- Histogram
- Bar chart
- pie
- Distribution
- Candlestick
- Box plot

making conclusion about population data using some experiments.

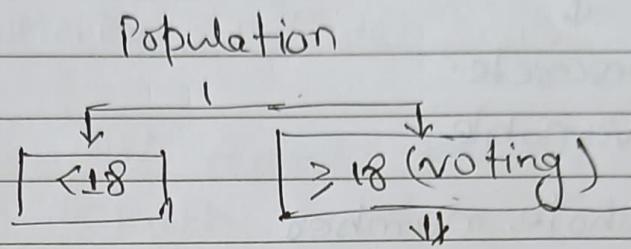
Population (N) & Sample (n)

Sampling Techniques

• Simple Random sampling: Every member of the population (N) has an equal chance of being selected for your sample (n).

• Stratified Sampling: strata \rightarrow layers \rightarrow characters \rightarrow Group.

Stratified Sampling:



Random Sampling

Systematic Sampling:

Select every n^{th} individual out of population (N)

Convenience Sampling:

Only those who are interested in the survey will only participate.

Survey Regarding new technology → Convenience Sampling

Variable: A variable is a property that can take any values.

Two different types of Variable

• Quantitative variable → Measured Numerically
& Mathematical operation &.

ex:- Age, weight, height, rainfall, temp, distance.

• Qualitative variables → Categorical Variables
Eg:- Gender, Types of movie, ...

Quantitative Variable

↓
Discrete
Variable

Eg:- whole numbers

No. of Bank account

No. of children

↓
Continuous
Variable.

Eg:- Height, weight,
ages, Rainfall.

→ Ass	Gender	Categorical
→	Marital status	Categorical
→	Ganga River length	Continuous
→	Movie duration	Continuous
→	Picode	discrete.
→	IQ	continuous discrete

Histogram:

Ages := {10, 12, 14, 18, 24, 26, 30, 35, 36, 37, 40, 41, 42, 43, 50, 51, 65, 68, 78, 90, 95, 100}

① Sort the numbers

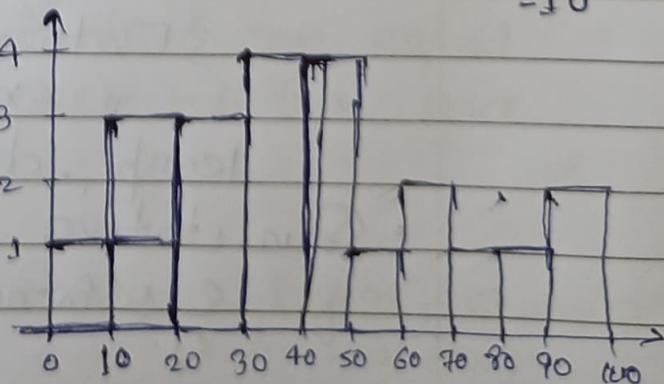
$$\min = 10 \quad \text{Bin size} = \frac{100}{10}$$

② Bin → No. of groups

$$\max = 100$$

③ Bin size : Size of Bins

$$= 10$$



Measure of Central Tendency:

- Mean
 - Median
 - Mode

A measure of central tendency is a single value that attempts to describe a set of data identifying the central position.

$$\text{population mean } (\mu) = \frac{1}{N} \sum_{i=1}^N x_i$$

$N \gg n$

Practical Application (Feature Engineering)

Age	Salary	Family size
-	NAN	-
-	-	NAN NAN - Not a number
NAN	-	-
-	-	NAN
-	NAN	-

we have some data that are missing, if we remove whole row data if any one data is missing in row, we will lose data or information that's why measure of central tendency is important, we can replace missing data with one of the measure of central tendency.

W Median

$$\{1, 2, 3, 4, 5\}$$

$$\bar{x} = 3$$

$$\{1, 2, 3, 4, 5, 100\}$$

$$\bar{x} = 19.16$$

↳ due to outliers

Steps to find out median:-

- sort the numbers

- find the central number

- if the no. of elements are even we find the average of central elements

- if the no. of elements are odd we find the central element

$$\{0, 1, 2, 3, 4, 5, 6, 7, 8, 100, 120\}$$

Median = 5

Mean = 25.6

If outlier is present,
use Median.

W Mode: most frequent occurring element

$$\{1, 2, 2, 3, 3, 3, 4, 5\}$$

↳ Mode is generally used for categorical data sets.

\times Measure of Dispersion:-

- Variance (σ^2)
- Standard deviation (σ)

\times Variance:

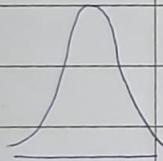
Population Variance (σ^2)	Sample Variance (s^2)
------------------------------------	---------------------------

$$\sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$$

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$\{1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}$

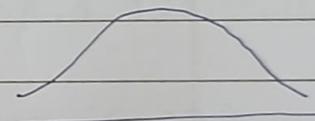
(Variance)_A



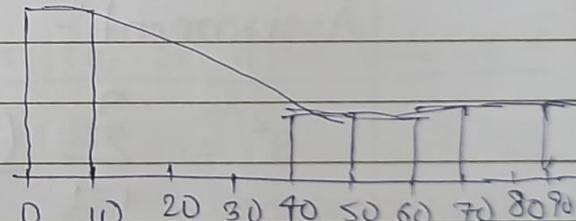
(Variance)_B > (Variance)_A

$\{1, 2, 3, 4, 50, 60, 70, 100\}$

(Variance)_B



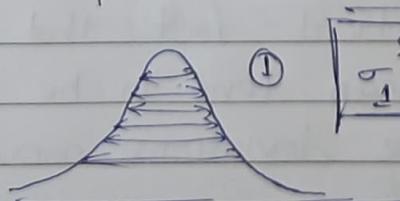
spread is less



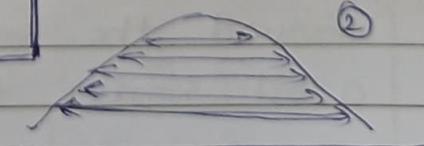
spread is more

$$\sigma^2 = []$$

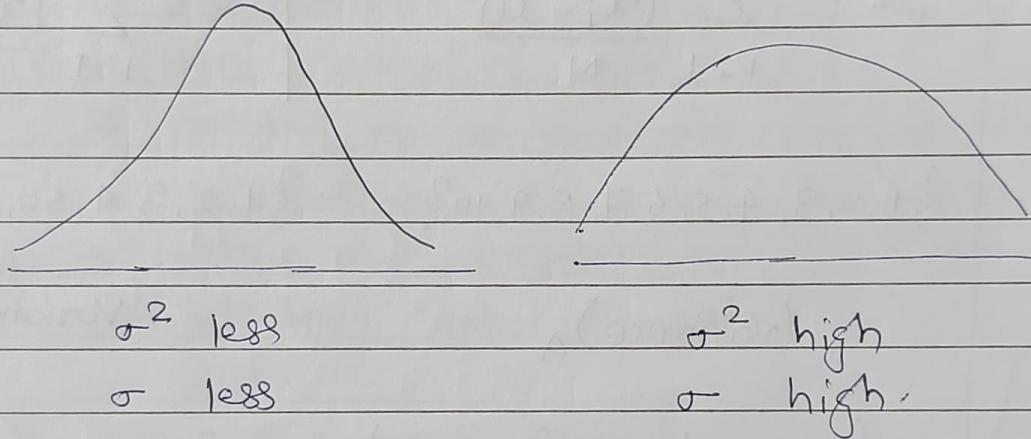
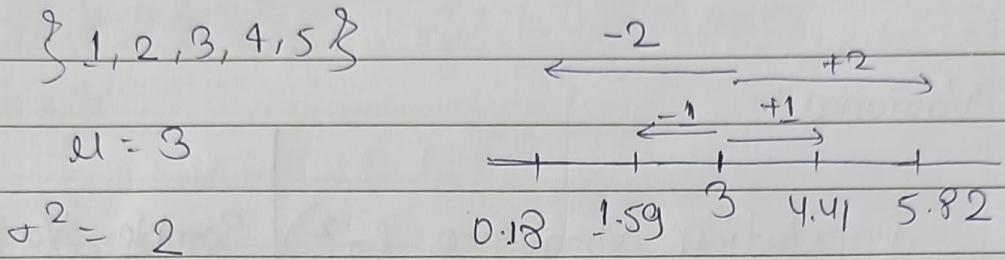
- As spread increases, Variance increases.
- As spread is less, Variance is less.



$$\sigma_1^2 < \sigma_2^2$$



Standard Deviation :-



Assignment:-

$$s^2 = \frac{1}{n-1} \sum_{i=1}^{n-1} (x_i - \bar{x})^2$$

Why here $n-1$?

Since, we compute deviations from the mean of all the items in the population, rather than the deviation from which the sample was drawn.

However population mean is generally unknown, so the sample mean is used in its place.

It is a mathematical fact that the deviations around the sample mean tend to be ~~smaller~~ than a bit smaller than the deviations around the

population mean and that dividing by $n-1$ rather than n provides exactly the right correction. It is also known as Bessel's correction.

v) Percentiles and Quartiles: A percentile is a value below which a certain percentage of observations lie.

99 percentile = It means the person has got better marks than 99% of the entire students.

Dataset : 2, 2, 3, 4, 5, 5, 5, 6, 7, 8, 8, 8, 8, 8, 9, 9, 10, 11, 11, 12
↓ ↓ ↓ ↓ ↓ ↓
0 1 2 3 4 5

What is the percentile rank of 10 percentile, rank of $x = \frac{\# \text{ No. of values below } x}{n}$

Percentile rank of 10 = $\frac{16}{20} = 80 \text{ percentile.}$

Percentile rank of 8 = $\frac{9}{20} = 45\%$

* What is the value that exists at 25 percentile

$$\text{Value} = \frac{\text{Percentile} \times n}{100}$$

$$\frac{25}{100} \times 20 = 5^{\text{th}} \text{ Index}$$

O/P = 5.

Index starts with '0'

IV

5 number summary :-

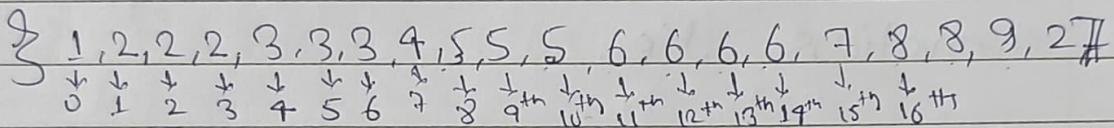
- Minimum
- First Quartile
- Median
- Third Quartile
- Maximum

}

Using Box plot

↑

Remove the outliers



[Lower fence \leftrightarrow Higher fence]

$$\text{Lower fence} = Q_1 - 1.5(\text{IQR})$$

$$\text{IQR} = Q_3 - Q_1$$

↓

Inter Quartile Range.

$$Q_1 = \frac{25}{100} \times (n+1) = \frac{25}{100} \times 21 = 5.25 \Rightarrow \text{Index}$$

$$Q_1 = \frac{25}{100} \times 21 = 5.25 \text{ index} = 3 \quad [\text{Average of } 5^{\text{th}} \text{ & } 6^{\text{th}} \text{ index value}]$$

$$Q_3 = \frac{75}{100} \times 21 = 15.75 \text{ index} = \frac{8+7}{2} = 7.5$$

$$\text{Lower fence} = 3 - (1.5)(4.5) = -3.65$$

$$\text{Higher fence} = 7.5 + (1.5)(4.5) = 14.25$$

[-3.65 \leftrightarrow 14.25]

so 27 is outlier in the above data set.

five number summary:-

minimum = 1

Q₁ = 3

median = 5

Q₃ = 7.5

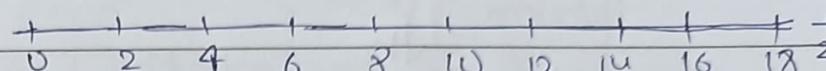
maximum = 9

Box plot

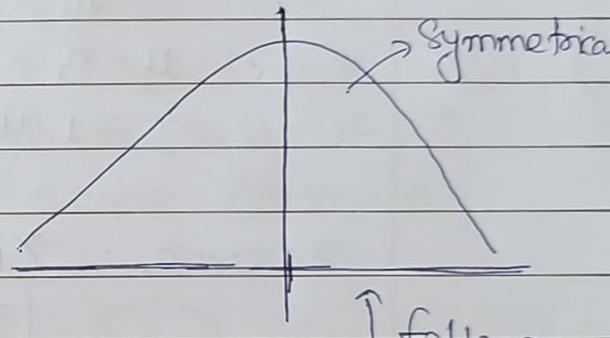
↑



outlier



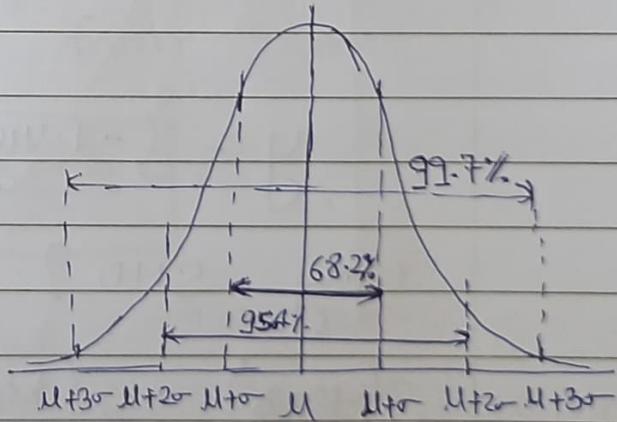
* Gaussian / Normal Distribution



Age, weight, height

Says (Domain expert)

* Empirical Rule of Normal Distribution



*

Q-Q plot for Distribution is Gaussian or Not?

W Standard Normal Distribution:

$X \approx$ Gaussian Distribution (μ, σ)

\Downarrow

$Y \approx$ Standard Normal Distribution ($\mu=0, \sigma=1$)

$$X = \{1, 2, 3, 4, 5\}$$

\Downarrow

$$\mu = 3$$

$$\sigma = 1.41$$

$$Z\text{-score} = \frac{X_i - \mu}{(\sigma / \sqrt{n})}$$

$$\text{Standard Error} = \frac{\sigma}{\sqrt{n}}$$

since we are applying Z-score formula
for all the values so $n = 1$

$$Z\text{-score} = \frac{X_i - \mu}{\sigma}$$

$$Y = \{-1.414, -0.707, 0, 0.707, 1.414\}$$

why SND? \rightarrow Standardization

(year)	(kg)	(cm)
Age	Weight	Height
24	72	150
26	78	160
32	84	165
33	92	170
34	87	150
28	83	180

In Machine Learning \rightarrow Math Required

↓

Algorithm Mathematical model

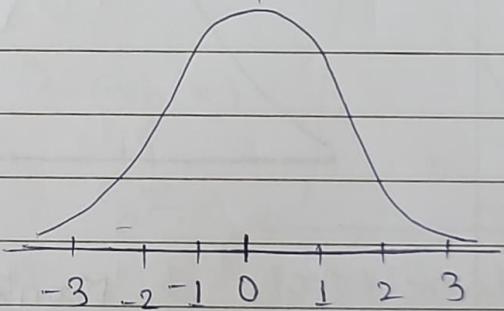
Time $\uparrow\uparrow \leftarrow$ Mathematical calculation \leftarrow

With the help of standard scaling, we try to bring data into same scale because of which calculation become easy.

~~Not standardization~~ with the help of Z-score formula

$$\mu = 0, \sigma = 1$$

almost all the data will range between $[-3, 3]$



because almost 99.7% of data is between 3 standard deviation from mean.

$$\boxed{Z\text{-score} = \frac{x_i - \mu}{\sigma}}$$

~~Normalization~~: you decide the Range in all the data will lie.

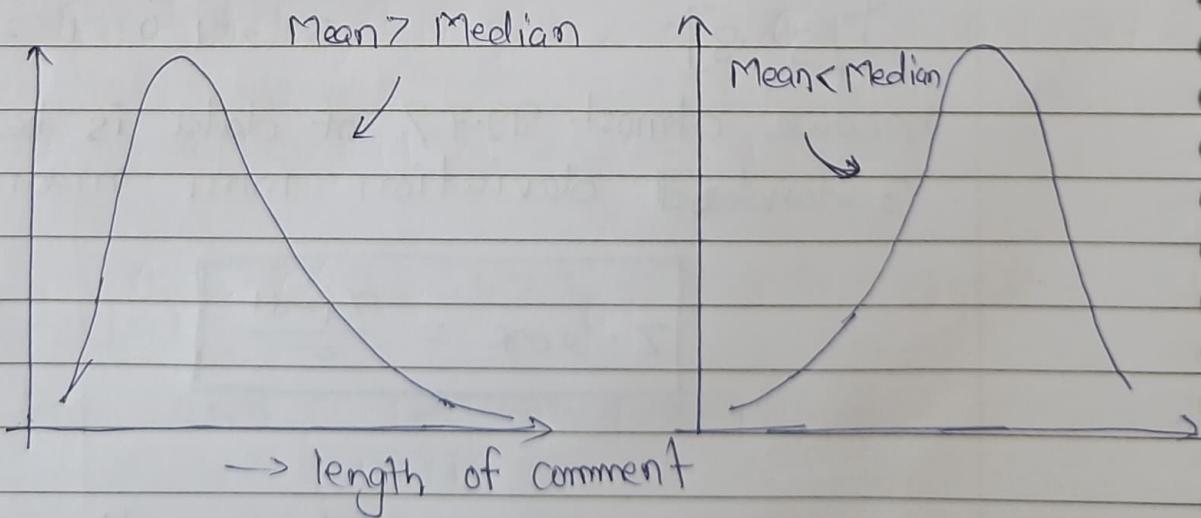
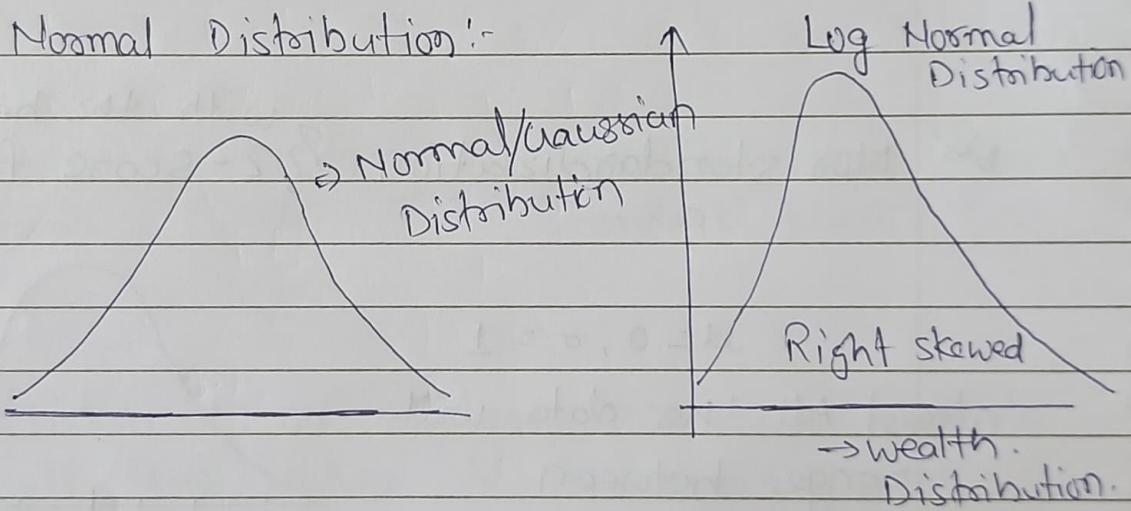
[lower scale \leftrightarrow Higher scale.]

- Min-Max Scales $[0 - 1]$

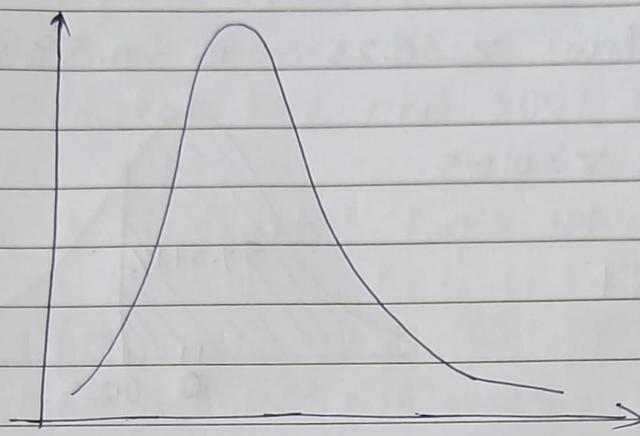
$$x_{\text{scaled}} = \frac{x - x_{\text{min}}}{x_{\text{max}} - x_{\text{min}}}$$

↳ Application: pixel values that lie b/w 0-255 can be brought between 0-1 scale.

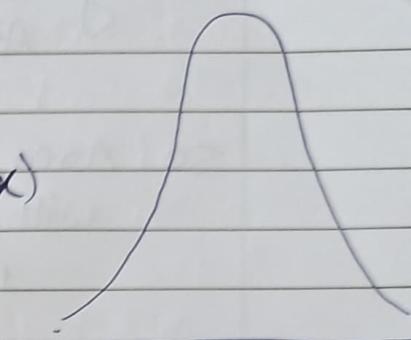
W Log Normal Distribution:-



from Ascending order given the relation of mean, median & mode.

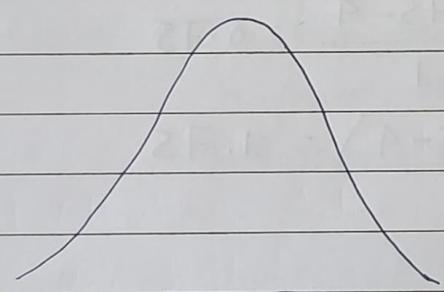


$$y = \ln(x)$$



$y \approx$ Normal Distribution

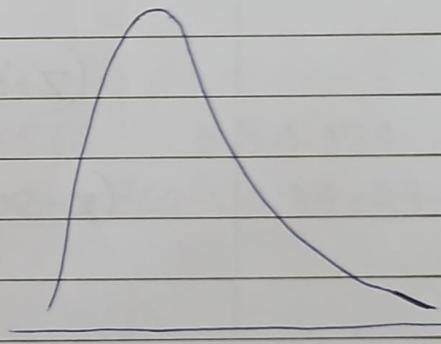
$x \approx$ log Normal Distribution



$$\Rightarrow \exp(x)$$

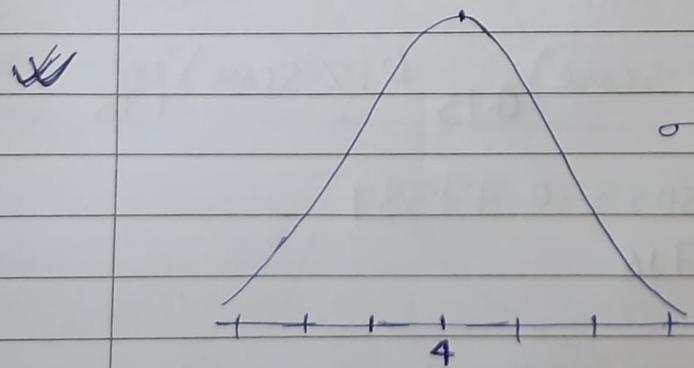
σ

e^x



$x \approx$ Normal distribution (μ, σ)

log Normal Dist



$$\sigma = 1$$

What is the percentage
of data that falls
above 4.25?

$$Z\text{-Score} = \frac{x - \mu}{\sigma} = \frac{4.25 - 4}{1} = 0.25$$

Using z-table we can find the area left of
 $z = 0.25$.

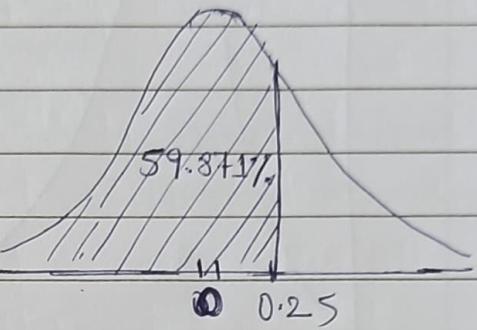
Using Z-table.

Area below $Z = 0.25$ is 0.59871

so Area above, $Z = 0.25$
will be!

$$1 - 0.5987$$

$$= 0.4013\%$$



v Percentage of data lies b/w 4.75 & 5.75 if mean is 4 & median is 1.

$$(Z\text{-score})_{4.75} = \frac{4.75 - 4}{1} = 0.75$$

$$(Z\text{-score})_{5.75} = \frac{(5.75 - 4)}{1} = 1.75$$

Area below/left of Z-score 0.75 = 0.77337

Area below/left of Z-score 1.75 = 0.95053

So, Area b/w $(Z\text{-score})_{0.75}$ & $(Z\text{-score})_{1.75}$

$$= 0.95053 - 0.77337$$

$$= 0.17716$$

v In India the average IQ is 100 with a standard Deviation of 15. what is the percentage of population would you expect to have ~~an~~ an IQ.

- lower than 85.
- Higher than 85.
- Between 85 and 100.

$$\mu = 100 \quad \sigma = 15$$

$$(z\text{-score})_{85} = \frac{85 - 100}{15} = -1$$

Using z-table:-

lower than 85 : ~~0.84134~~

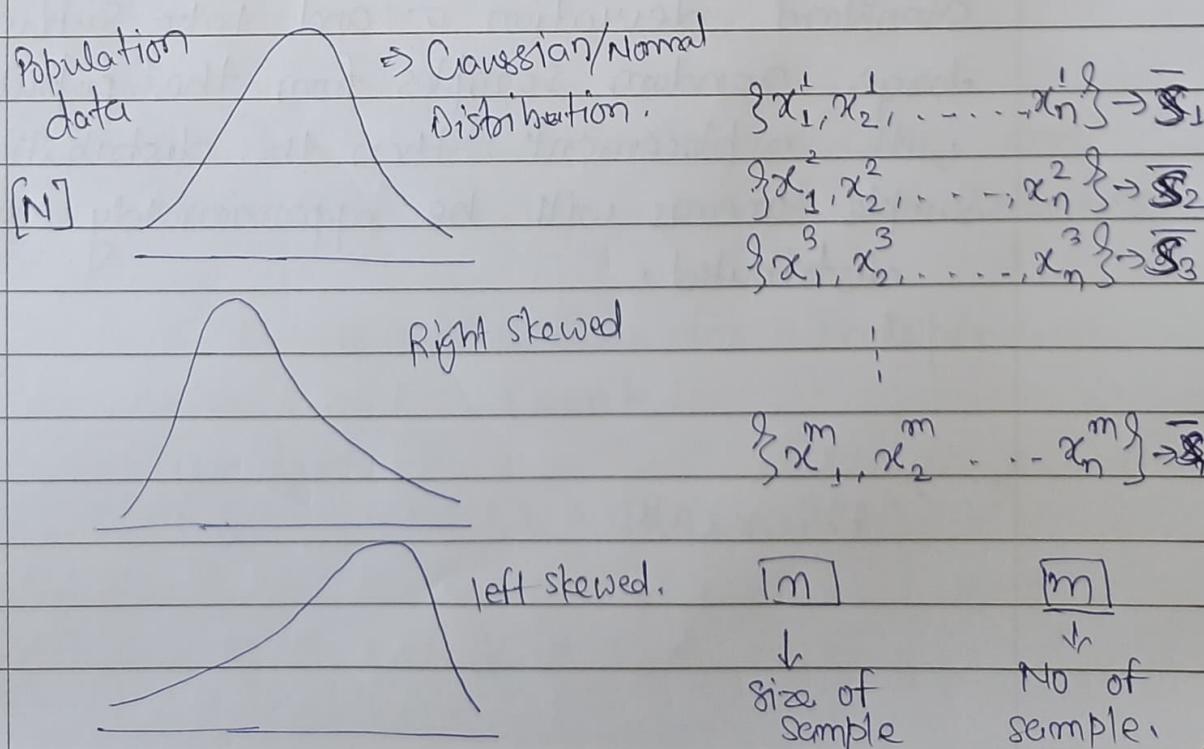
Higher than 85

lower than 85 : 0.15866

higher than 85 : $1 - 0.15866 = 0.84134$

Between 85 & 100 : $0.5 - 0.15866 = 0.34134$

Central Limit Theorem :-



Central limit theorem states that Given distribution may or may not be Gaussian/ Normal distribution of population data. If we are taking samples of size $n \geq 30$ from the population and calculate the mean of different ~~different~~ ^{then} in the distribution of the sample means will be approximately normally distributed.

W Size of shock through out the world?

10 different Region $n \geq 30$

↳ Many Assumptions can be made.

The central limit theorem states that if you have a population with mean μ and standard deviation σ and take sufficiently large random samples from the population with replacement, then the distribution of sample means will be approximately normally distributed.

Probability: Probability is a measure of the likelihood of an event.

Mutual Exclusive Event

Two events are mutually exclusive if they can't occur at the same time.

- Tossing a coin • Rolling a dice

$$P(A \text{ or } B) = P(A) + P(B)$$

Addition Rule
for ME

Non mutual Exclusive events

Two events can occur at the same time.

- picking randomly a cards , two "hearts" and "king" can be selected.

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

Addition Rule for Non mutual exclusive events.

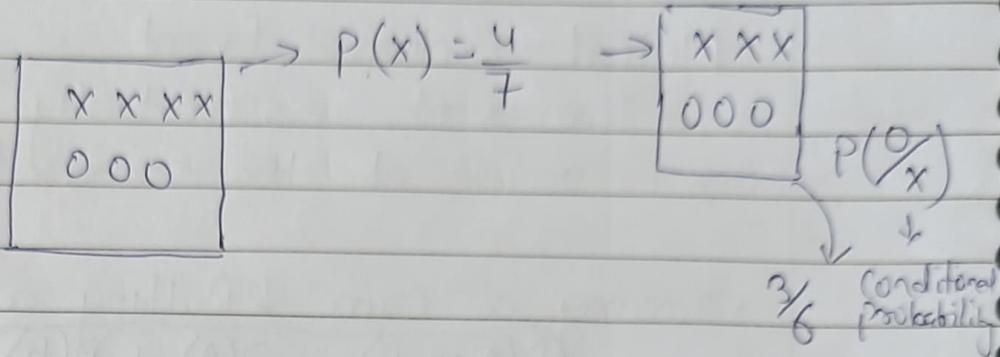
Multiplication Rule:-

Dependent events: Two events are dependent if they affect one another.

Multiplication Rule for Independent Events

$$P(A \text{ and } B) = P(A) \times P(B)$$

v) Probability of drawing a "orange" and drawing a "yellow" marble from the bag?



$$P(x \text{ and } o) = P(x) * P(o/x)$$

$$P(A \text{ and } B) = P(A) * P(B/A)$$

$$P(A \text{ and } B) = P(B) * P(A/B)$$

v) Permutation & Combination

↳ with permutation, order matters.

means all the possible arrangement matters.

$$n_{P_r} = \frac{n!}{(n-r)!}$$

n = Total no. of object

r = No. of selection.

↳ with combination order doesn't matter. Only unique combination is considered.

$$n_{C_r} = \frac{n!}{(n-r)!r!}$$

n = Total no. of object

r = No. of selection.

W CO-Variance **

↳ One of Feature selection is done with the help of covariance and co-relation

Age weight

12 40

13 45

15 48

17 60

18 62

Age ↑ weight ↑

Age ↓ weight ↓

weight

x x

x

x

Age

Can we Quantify the relationship x & y using mathematical equation.

$$\text{Cov}(x, y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

$$\sigma^2 = \frac{\sum (x_i - \bar{x})^2}{n-1} = \frac{\sum (x_i - \bar{x})(x_i - \bar{x})}{n-1}$$

$$\Rightarrow \text{Cov}(x, x) = \sigma^2, = \text{Variance}(x)$$

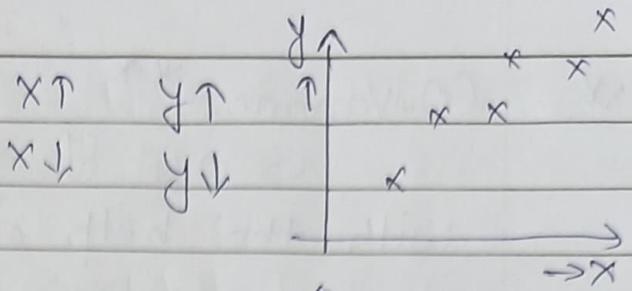
$$\bar{x} = 15$$

$$\bar{y} = 51$$

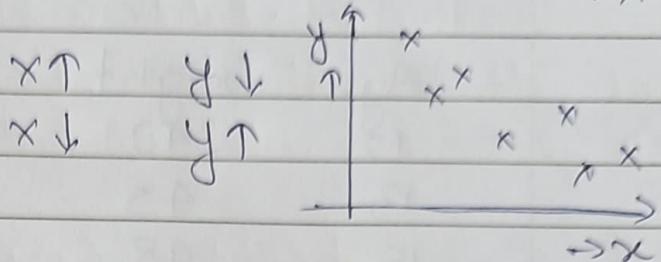
$$\begin{aligned} \text{Cov}(x, y) &= (12-15)(40-51) + (13-15)(45-51) + (15-15)(48-51) \\ &\quad + (17-15)(60-51) + (18-15)(62-51) \end{aligned}$$

$$\begin{aligned} &= -3 \times 8 + -2 \times 6 + 0 + 2 \times 9 + 3 \times 1 \\ &= -24 + 12 + 18 + 27 \\ &= 21.75 \end{aligned}$$

+ve Covariance

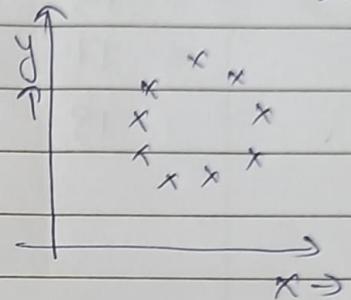


-ve Covariance



0 covariance

No Relation



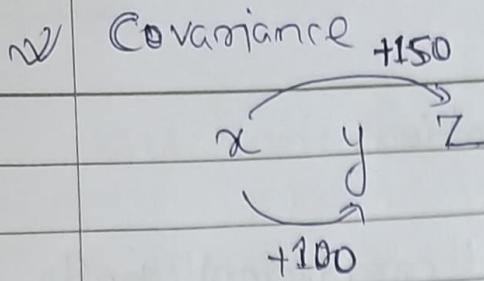
Q Pearson correlation coefficient (-1 to 1)

$$r(x,y) = \frac{\text{cov}(x,y)}{\sigma_x \cdot \sigma_y}$$

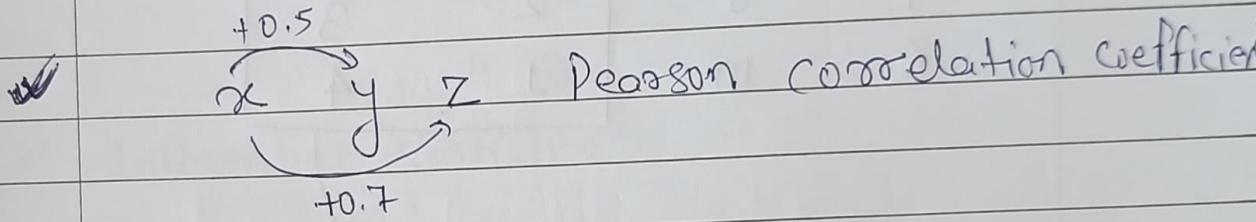
Why pearson correlation?

It is because covariance of two variable can be any positive value or negative value. That's why we would be able to compare which variable more correlated to a variable.

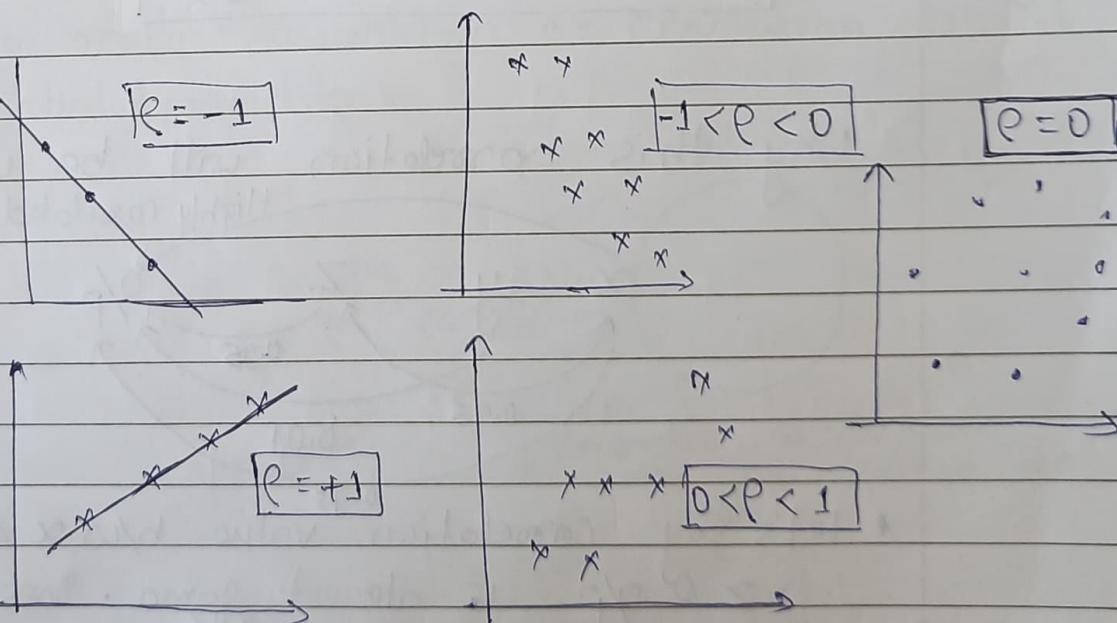
But in case of pearson correlation coefficient ~~the~~ correlation values are always between [-1, 1] so we would easily able to correlate the variables that which are more correlated to one variable.



As we see covariance b/w x & z is greater than covariance b/w x & y even tho. we would ^{not} be able to comment if x & z is more correlated than x & y or not. Because covariance b/w x & z and x & y might not be in same scale.



Here we can visualize which variables are more correlated to each other because pearson correlation coefficient is always in same scale [-1 1]



X

Spearman Rank Correlation:-

Pearson correlation coefficient holds good only for linear data.

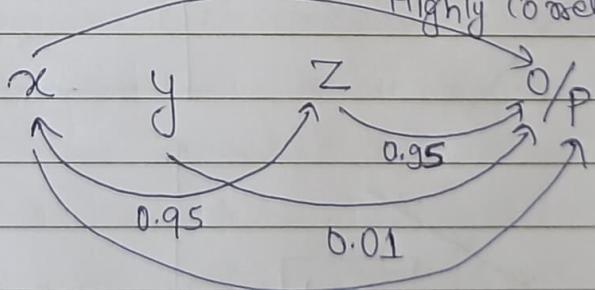
Spearman Rank correlation holds good for Non-linear data.

X	Y	R(X)	R(Y)
10	4	4	1
8	6	3	2
7	8	2	3
6	10	1	4

$$\rho_s = \frac{\text{Cov}(R(x), R(y))}{\sigma(R(x)) \sigma(R(y))}$$

Why this correlation will be used?

Highly correlated.



- Let's say correlation value b/w x & $\hat{o/p}$ and z & $\hat{o/p}$ is almost same. Doesn't it mean that both features are same? Yes we can remove one feature because both

are same features.

- let's assume x is highly correlated with y/p and y is very less correlated with y/p then we can neglect the feature y .
- correlation coefficients helps us in feature engineering.

✓ Inferential statistics :-

Hypothesis Testing :-

We have population data and sample data. Using sample data we are going to make assumption or conclusion about population data.

