

## Lecture - 04

→ IQR (Python Code)

→ Probability .

→ Permutation and Combination .

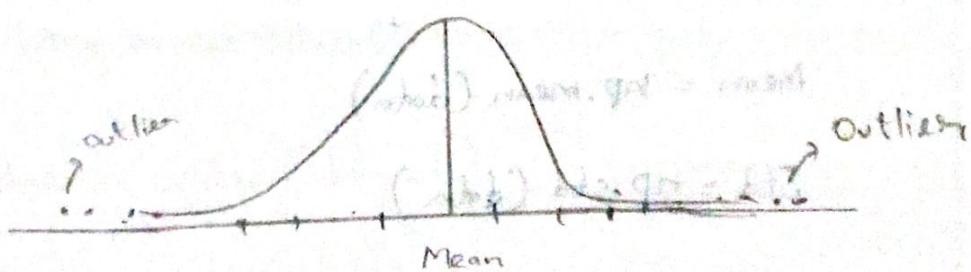
→ Confidence Interval .

→ P-Value

→ Hypothesis testing .

### Outlier:

The outlier lies after the left and right of 3rd Standard deviation.



Python Code:

# Example for Outlier detection program

(i) by np . median()

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
%matplotlib inline
```

```
# Define data
```

```
dataset = [11, 10, 12, 14, 12, 15, 14, 13, 15, 102, 12, 14, 17, 19,  
107, 10, 13, 12, 14, 12, 108, 12, 11, 14, 13, 15, 10, 15,  
12, 10, 14, 13, 15, 10]
```

```
Outliers = [] # initialize empty list
```

~~def detect\_outlier(data):~~

```
def detect_outlier(data):
```

```
threshold = 3 # 3rd std away should be call outliers.
```

```
Mean = np.mean(data)
```

```
Std = np.std(data)
```

```
for i in data:
```

```
z-score = (i - mean) / Std
```

```
if np.abs(z-score) > threshold:
```

```
    Outliers.append(i)
```

## Identify Outliers

detect\_outlier (dataset)

$$(Q_3 + 1.5 \times IQR) + 3 \times \text{std deviation}$$
$$(Q_1 - 1.5 \times IQR) - 3 \times \text{std deviation}$$

Output:

[102, 107, 108]

IQR: [finding IQR]

1. Sort the data.

2. Calculate  $Q_1$  and  $Q_3$ .

3.  $IQR = Q_3 - Q_1$

4. finding the lower fence ( $Q_1 - 1.5 \times IQR$ )

5. finding the upper fence ( $Q_3 + 1.5 \times IQR$ )

Program:

# Using above dataset

dataset = sorted(dataset)

# finding Quartiles:  $Q_1$  and  $Q_3$

$Q_1, Q_3 = np.\text{Percentile}(\text{dataset}, [25, 75])$

# finding IQR

$IQR = (Q_3 - Q_1)$

# find lower and higher fence

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

$$\text{Higher-fence} = Q_3 + (1.5 * \text{IQR})$$

print (lower-fence, Higher-fence)

7.5 , 19.5

# Condition for finding Outlier and removing

for i in dataset:

if  $i < 7.5$  and  $i > 19.5$ :

dataset.remove(i)

~~return~~

return "No outlier present"

## PROBABILITY

Probability is a measure of the likelihood of

an event.

Eg: Rolling a dice  $\{1, 2, 3, 4, 5, 6\}$

To find probability of getting 6 in dice,

$$P(6) = \frac{\text{Number of way an event occur}}{\text{Number of possible Outcomes}}$$

$$P(6) = \frac{1}{6}$$

Another Example,

Toss a Coin {H, T}

$$P(H) = \frac{1}{2}$$

$$P(T) = \frac{1}{2}$$

Addition Rule (probability, "or")

We need to understand addition rule, first

know about mutual exclusive and non-mutual exclusive events.

Mutual - Exclusive events:

Two events are mutual exclusive if they cannot occur at the same time.

Eg: Rolling a dice  $\{1, 2, 3, 4, 5, 6\}$

Outcome -  $\{\{1\}, \{2\}, \{3\}, \{4\}, \{5\}, \{6\}\}$

Cannot come -  $\{\{1, 2\}, \{3, 4\}\}$  [Events Cannot occur at  
Two or more  
Some time]

Another Example: Tossing a coin - either head (or) tail.

Non-Mutually Exclusive Events:

Multiple events can occur at the same  
time.

Eg: Deck of Cards  $\{\underline{Q}, \heartsuit\}$  [Queen card with heart card]  
↑  
Two events occur.

Addition rule:

This rule is used to find probability of  
two events based on the type of event occur.

Problem:

Problem based on mutual exclusive events.

1. If I toss a coin, what is the probability of the coin landing on heads or tails?

Soln:

The above events are Mutual Exclusive.

$P(A)$  - probability of head event     $P(B)$  - probability of tail event.

$P(A \text{ or } B)$  ?

$$P(A \text{ or } B) = P(A) + P(B) \quad [\text{Addition rule}]$$

$$= \frac{1}{2} + \frac{1}{2}$$

$$= 1$$

$$\boxed{P(A \text{ or } B) = 1}$$

Another Examples

Rolling a dice.

$P(1 \text{ or } 3 \text{ or } 6)$  ?

$$P(1 \text{ or } 3 \text{ or } 6) = P(1) + P(3) + P(6)$$

$$= \frac{1}{6} + \frac{1}{6} + \frac{1}{6}$$

$$P(1 \text{ or } 3 \text{ or } 6) = \frac{3}{6} \Rightarrow \frac{1}{2}$$

Problem:

Problem based on non-mutual exclusive events.

2. You are picking a card randomly from a deck. What is the probability of choosing a card that is Queen or a heart?

The above events are non-mutual exclusive events.

$P(A)$  - Probability of card that are Queen.

$P(B)$  - Probability of card that are a heart.

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

$$= \frac{4}{52} + \frac{13}{52} - \frac{1}{52}$$

$$= \frac{16}{52}$$

$$P(A) = \frac{4}{52} \quad P(B) = \frac{13}{52}$$

↑  
Queen

↑  
Heart

$$P(A \text{ and } B) = \frac{1}{52}$$

Card with Queen and heart

$$P(A \text{ or } B) = \frac{16}{52}$$

## Multiplication Rule:

We need to understand multiplication rule, first understand about independent and dependent events.

### Independent Events:

It is the event, it does not depend upon the other event.

Eg: Rolling a dice  $\{1, 2, 3, 4, 5, 6\}$

↑  
Each and every are independent

Outcome like 1, 1, 2, 2, 6, 5...

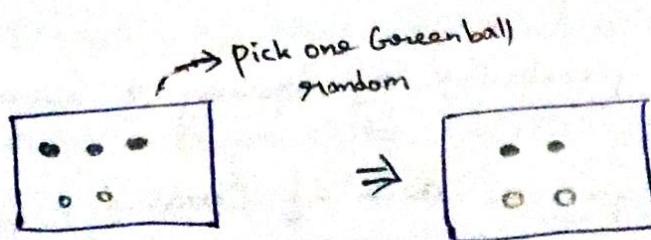
### Dependent Event:

One event are depend upon the other events.

These are non-independent.

- → Red ball
- → Green ball

Eg:



$$P(\text{Green}) = \frac{3}{5}$$

$$P(\text{Red}) = \frac{2}{4}$$

dependent

↓ we apply something in ML called as,

Naive Bayes (Conditional Probability)

Problem:

1. Problem based on Independent events.

1. What is the probability of rolling a "5" and then a "4" in a dice?

The above event are independent event.

$$P(A \text{ and } B) = P(A) * P(B) \quad \text{[Multiplication rule]}$$

$$P(5 \text{ and } 4) = \frac{1}{6} * \frac{1}{6} = \frac{1}{36}$$

$$\boxed{P(5 \text{ and } 4) = \frac{1}{36}}$$

Problem:

Problem based on dependent events

1. What is the probability of drawing a Queen and then a Ace from a deck of cards?

The above events are dependent event.

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

conditional probability  
 Bayes theorem

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

$$= \frac{4}{52} * \frac{4}{51}$$

$$P(Q \text{ and } A) = \frac{16}{2652}$$

### Permutation and Combination:

#### Permutation:

→ Permutation is the arrangement of items in which order matters.

→ Number of ways of selection and arrangement of items in which order matters.

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

### Combination:

$$(a)(b)q + (a)q = (a+b)q$$

→ Combination is the selection of items in which Order does not matter.

$$(a)(b)q + (a)q = (a+b)q$$

→ Number of ways of Selection of items in which Order does not matters.

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

### Permutation Example:

I went to school trip to chocolate factory. In Chocolate factory, six type of brands are available are dairy milk, 5 star, mars, kitkat, milkybar and snickers. Assign a task for Student, ~~to take~~ whichever chocolate you see first three chocolate need to be noted in diary.

Soln:

Student →

1st	2nd	3rd
—	—	—
noted	noted	noted

$$= 6 \times 5 \times 4 = \underline{120 \text{ ways}}.$$

↙ Permutation

Dairymilk, Milkybar, Kitkat

Kitkat, Milkybar, Dairymilk.

Kitkat, Dairymilk, Milkybar.

— — —

Up to 120 way permutation

So,

$${}^n P_r = \frac{n!}{(n-r)!} = \frac{6!}{(6-3)!}$$

$$= \frac{6 \times 5 \times 4 \times 3!}{3!} = 120 \text{ ways.}$$

Permutation return all possible ways of items.

Combination Example:

Combination return Only Unique possible ways of items.

$${}^n C_r = \frac{n!}{r!(n-r)!} = \frac{6!}{3!(6-3)!} = \frac{6 \times 5 \times 4 \times 3!}{3! \times 3!}$$

$$= \frac{6^2}{1} \times 5 \times 4^2$$

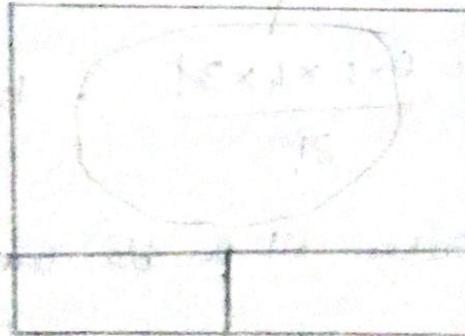
$$1 \times 2 \times 3$$

$$= 2 \times 5 \times 2 = 20 \text{ ways [unique ways]}$$

### Concept of P-value:

P-value refers to the probability value of data fall in distribution.

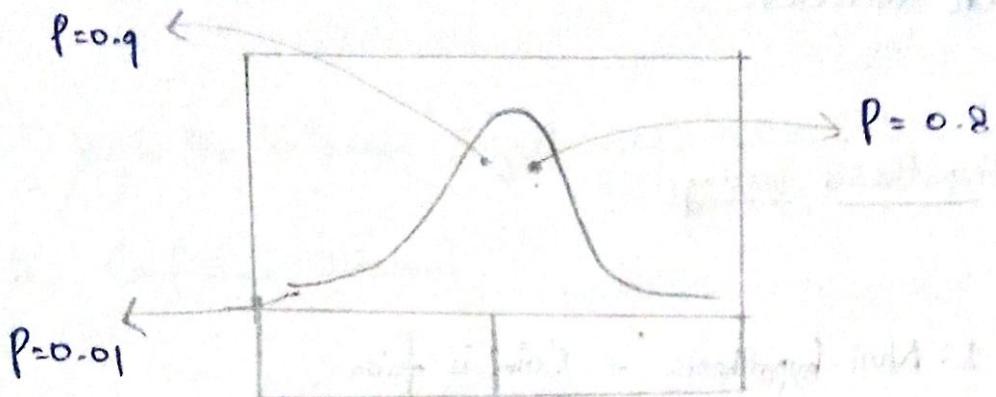
### Example:



MOUSE PAD IN LAPTOP

Every 100 time I touch the mouse pad 80 times I touch this specific region.

(Consider Same as,



The above one represents the how much data fall in the region in the terms of Probability.

Combine topic Hypothesis testing, Confidence Interval, Significance

Value.

Coin  $\rightarrow$  Test whether this coin is a fair Coin or not by performing 100 tosses.

Coin is fair,  $P(H) = 0.5$ ,  $P(T) = 0.5$

50 times head (or) tail we consider the Coin is fair.

Consider 50 times head, we Consider the Coin is fair.

To test the coin, we use the hypothesis testing on statistics.

### Hypothesis Testing:

1. Null hypothesis - Coin is fair.
2. Alternate hypothesis - Coin is not fair.
3. Experiment.
4. Reject or accept the null hypothesis.

$$CI = 95\%$$

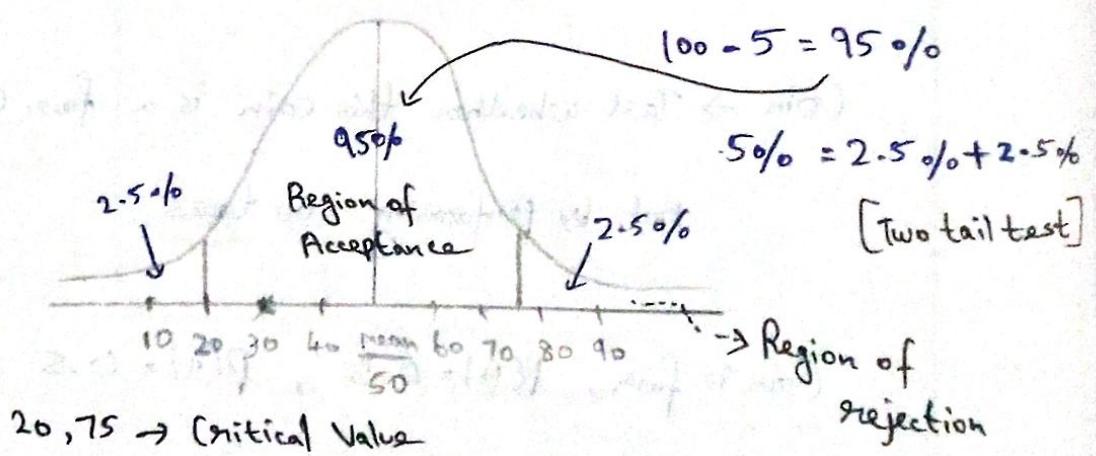
$$\text{Significance Value } (\alpha) = 0.05$$

$$1 - 0.05 = 0.95$$

$$100 - 5 = 95\%$$

$$5\% = 2.5\% + 2.5\%$$

[Two tail test]



50 times → head [we definitely say it is fair coin]

We cannot perform 100 experiments because it is difficult.

So, In domain export define Something Called as Significance Value for normal distribution.

With the help of Significance Value, We define the Confidence Interval.

With the help of Confidence Interval, We Conclude the Coin is fair Coin ( $\sigma$ ) not.

Then, I toss a coin 30 times. Then Say the Coin is fair Coin  $\sigma$  or not.

The Confidence interval with 20 - 75. So, obviously. The Coin is fair.

Without <sup>m</sup> exponent we say that the Coin is fair or not based on the Confidence Interval.

The data fall outside the Confidence Interval. We say the Coin is not fair.

Significance value  $\alpha = 0.05$

15 heads - Coin is not fair

80 heads - Coin is not fair

40 heads - Coin is fair