

PROBABILITY

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Ch 0: Mathematical background

Set theory

A set is a collection of distinct elements

example :-

$$S_1 = \{1, 2, 3, 4, 5, 6\}$$

$$S_2 = \{\text{Head, Tail}\}$$

Let $S = \{x_1, x_2, x_3, \dots, x_n\}$
 $x_1 \in S$: x_1 is an element of the set S
 $x_n \in S$: x_n is an element of the set S
 $x_j, j > n$

$$x_j \notin S$$

x_j is not an element of the set S

- The empty (null) set is the set that contains no elements and is denoted by \emptyset
- Universal set (denoted by Ω): it contains all objects that could be conceivably be of interest in a particular context.

Set classification:-

① Finite set

Ex

$$A = \{1, 2, 3\}$$

$$B = \{2, 4, 6\}$$

② countably infinite set $A = \{1, 2, \dots\}$

A set that satisfies certain property P

$$A = \{x \mid x \text{ satisfies } P\}$$

* the set of even integers

$$S = \{k \mid k/2 \text{ is integer}\}$$

* the set of all real numbers between 0 and 1

$$S = \{x \mid 0 < x < 1\}$$

↳ continuous, uncountable

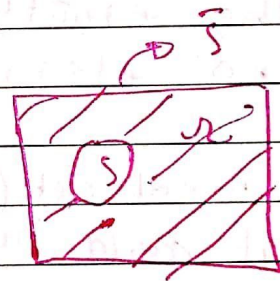
→ Set operations:

1 - The complement of a set S with respect to a universal set Ω is the set

$$\bar{S} = \{x \in \Omega \mid x \notin S\}$$

$$\Delta \Omega - S$$

$$\bar{\Omega} = \emptyset$$



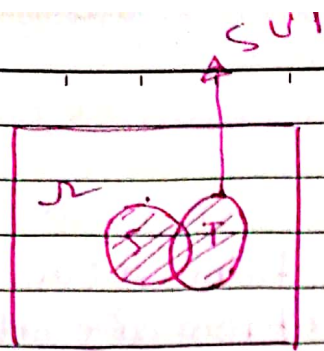
2 - Union of sets: the union of two sets S and T is the set of all elements that belong to S or T (or both) and denoted by $S \cup T$

$$S \cup T = \{x \mid x \in S \text{ or } x \in T\}$$

$$S = \{1, 2\}$$

$$S \cup T = \{1, 2, 4\}$$

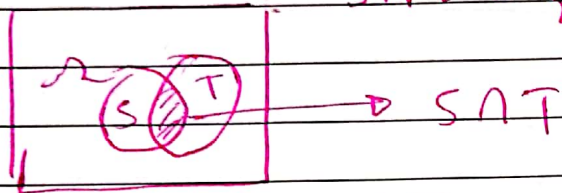
$$T = \{2, 4\}$$



3 - Intersection of sets

the intersection of two sets S and T ($S \cap T$) is the set of all elements that belongs to both S and T

$$S \cap T = \{x \mid x \in S \text{ and } x \in T\} \therefore$$



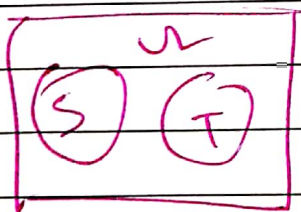
* For infinitely many sets S_1, S_2, \dots

$$S_1 \cup S_2 \cup S_3 \cup \dots = \bigcup_{n=1}^{\infty} S_n = \{x \mid x \in S_n \text{ for some } n\}$$

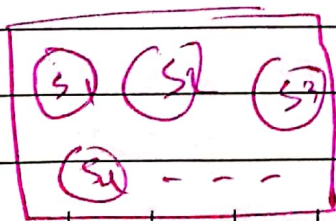
$$S_1 \cap S_2 \cap S_3 \cap \dots = \bigcap_{n=1}^{\infty} S_n = \{x \mid x \in S_n \text{ for all } n\}$$

4 - Two sets are said to be disjoint if their intersection is empty

S and T are disjoint if $S \cap T = \emptyset$



* several sets are said to be disjoint if no two of them have a common element



S_1, S_2, \dots, S_n are disjoint if

$$S_i \cap S_j = \emptyset \quad \text{if } i \neq j$$

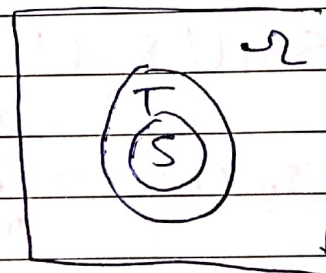
S-Partitions

A collection of sets is said to be a partition of set S if the sets in the collection are disjoint and their union is S .



S is a subset of T if every element of S is also an element of T .

$$S \subset T \iff T \supset S$$



1- Equality: if $S \subset T$ and $T \subset S$, then $S = T$

2- Transitivity: if $S \subset T$ and $T \subset V$, then $S \subset V$

* Algebra of sets:

1- $S \cup T = T \cup S$

2- $S \cap (T \cup A) = (S \cap T) \cup (S \cap A)$

3- $\overline{(\overline{S})} = S$

4- $S \cup \Omega = \Omega$

5- $S \cup (T \cap A) = (S \cup T) \cap (S \cup A)$

6- $S \cup (T \cap A) = (S \cup T) \cap (S \cup A)$

7- $S \cap \overline{S} = \emptyset$

8- $S \cap \Omega = S$

Ex: consider an experiment involving rolling a 4-faced die

* $\Omega = \{1, 2, 3, 4\}$ \rightarrow universal set

* A is the set of even outcomes

$$A = \{2, 4\}$$

* B is the set of all outcomes that are less than 3

$$B = \{1, 2\}$$

$$* A \cap B = \{2\}$$

$$* A \cup B = \{1, 2, 4\}$$

$$* A \cap \bar{B} = \{4\}$$

$$1 - \bar{B} = \{3, 4\}$$

$$A \cap \bar{B} = \{4\}$$

or we can say

(another method)

$$2 - A \cap \bar{B} = A - B = \{4\}$$

$$* B \cap \bar{A} = B - A = \{1\}$$

→ DE Morgan's law :-

$$1 - \overline{(A \cap B)} = \bar{A} \cup \bar{B}$$

$$2 - \overline{(A \cup B)} = \bar{A} \cap \bar{B}$$

1- Proof

$$x \in \overline{(A \cap B)} \iff x \notin (A \cap B) \iff \left\{ \begin{array}{l} x \notin A \\ \text{or} \\ x \notin B \end{array} \right.$$

$$\iff \left\{ \begin{array}{l} x \in \bar{A} \\ \text{or} \\ x \in \bar{B} \end{array} \right. \iff \bar{A} \cup \bar{B} \iff x \in (\bar{A} \cup \bar{B})$$

2 - proof

$$x \in \overline{(A \cup B)} \iff x \notin (A \cup B) \iff \left\{ \begin{array}{l} x \notin A \\ \text{and} \\ x \notin B \end{array} \right\}$$

$$\iff \left\{ \begin{array}{l} x \in \bar{A} \\ \text{and} \\ x \in \bar{B} \end{array} \right\} \iff x \in (\bar{A} \cap \bar{B})$$

Generally

$$*(A_1 \cup A_2 \cup A_3 \dots) = \left(\bigcup_{n=1}^{\infty} A_n \right) = \overline{\bigcap_{n=1}^{\infty} \bar{A}_n}$$

$$*(A_1 \cap A_2 \cap A_3 \dots) = \left(\bigcap_{n=1}^{\infty} A_n \right) = \overline{\bigcup_{n=1}^{\infty} \bar{A}_n}$$

show that

Ex: 1- $(A \cap (B \cup C)) = (\bar{A} \cup \bar{B}) \cap (\bar{A} \cup \bar{C})$

2- $\overline{(A \cap B) \cup (A \cap C)} = \bar{A} \cup (\bar{B} \cap \bar{C})$

Probabilistic model (PM)

1- Sample space " Ω "

- the set of all possible outcomes of an experiment

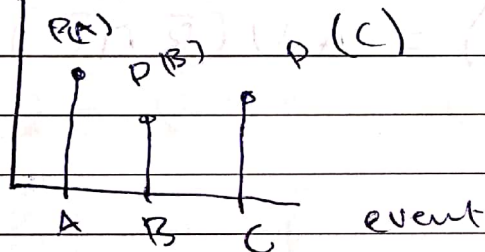
* Tossing a coin twice

<u>I</u>	<u>II</u>
H	H
H	T
T	H
T	T

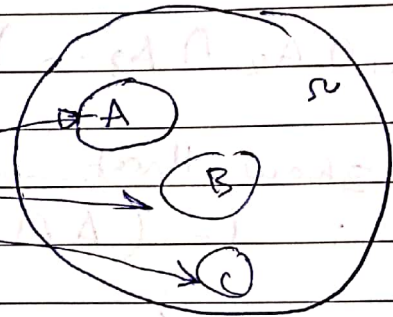
$$\Omega = \{ (H, H), (H, T), (T, H), (T, T) \}$$

2- Probability law

prob



Experiment

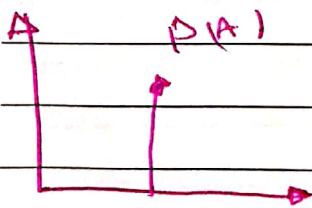


$$P(\Omega) = 1$$

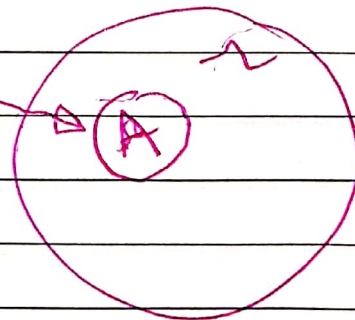
$$(\forall, P(A) \geq 0$$

* assigns to a set "A" of possible outcomes a non negative number "P(A)". P(A) encodes our beliefs about the likelihood element of "A"

Probability



experiment



Probability ~~axioms~~ axioms

1- non-negativity

$$P(A) \geq 0$$

2- normalization

$$P(\Omega) = 1$$

3- if A and B are disjoint ($A \cap B = \emptyset$), then

$$P(A \cup B) = P(A) + P(B)$$

→ * example: given a 4-sided die

$$\Omega = \{1, 2, 3, 4\}$$

$$P(\Omega) = P(\{1, 2, 3, 4\}) = 1$$

$$P(\{1\}) = \frac{1}{4}$$

$$P(\{1, 2\}) = P(\{1\}) + P(\{2\}) = \frac{2}{4} = \frac{1}{2}$$

let A_1, A_2, \dots, A_n be disjoint then

$$P(A_1 \cup A_2 \cup \dots \cup A_n) = \sum_{i=1}^n P(A_i)$$

Consequences of the Axioms

$$1 - P(A) \leq 1$$

$$\Omega = A \cup \bar{A}$$

$$P(\Omega) = 1 \longrightarrow P(A \cup \bar{A}) \quad \text{since they are disjoint}$$

$$= P(A) + P(\bar{A}) = 1$$

$$P(A) = 1 - P(\bar{A}) \longrightarrow P(A) \leq 1 \quad \# \text{ non-negativity} \\ \text{probability axioms}$$

$$2 - P(\emptyset) = 0$$

$$\Omega = \Omega \cup \emptyset$$

~~$$\emptyset = \Omega - \Omega$$~~

$$P(\Omega) = P(\Omega \cup \emptyset)$$

~~$$P(\emptyset) = P(\Omega - \Omega)$$~~

\hookrightarrow disjoint

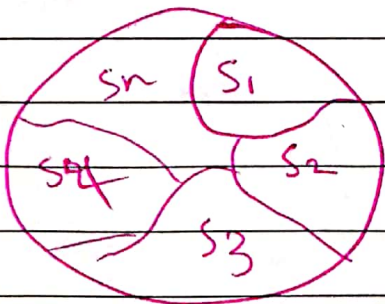
$$P(\Omega) = P(A) + P(\emptyset)$$

$$P(\emptyset) = 0 !!$$

Discrete probability law

$$\text{let } S = \{s_1, s_2, \dots, s_n\}$$

$$P(\{s_1, s_2, s_3, \dots, s_n\}) = P(\{s_1\}) + P(\{s_2\}) + \dots + P(\{s_n\})$$

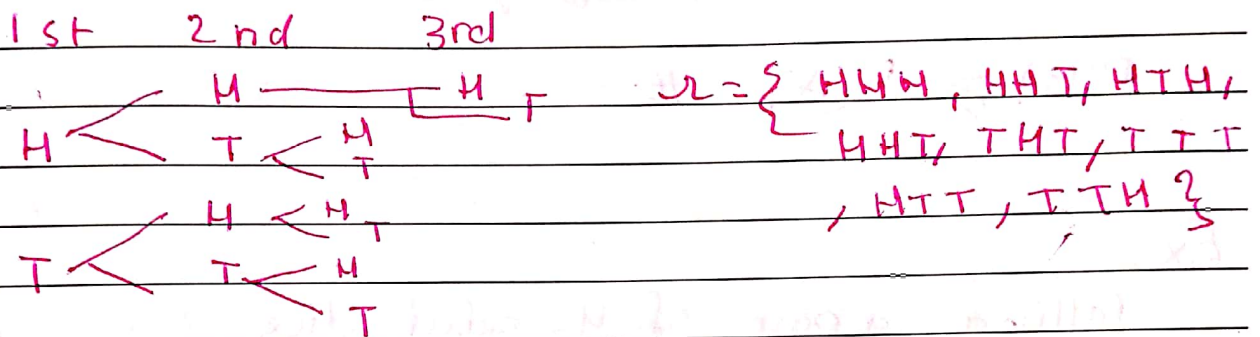


Discrete uniform probability law

- As a special case when $P(s_1) = P(s_2) = \dots = P(s_n)$
then the probability of any event "A" is
given by :- $P(A) = \frac{\# \text{ of elements in "A"}}{n}$

example

Tossing a coin 3 times



let "A" be defined as $A = \{ \text{exactly 2 Heds occur} \}$

$$\begin{aligned} P(A) &= P(\{ \text{HHT}, \text{HTH}, \text{THH} \}) \\ &= P(\{ \text{HHT} \}) + P(\{ \text{HTH} \}) + P(\{ \text{THH} \}) \\ &= \frac{1}{8} + \frac{1}{8} + \frac{1}{8} = \frac{3}{8} \end{aligned}$$

- since all outcomes occur equally likely with probability $\frac{1}{8}$

A has 3 elements / Ω has 8 elements

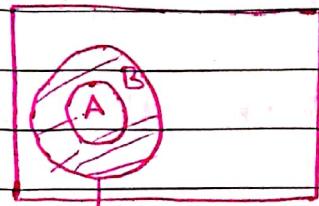
Discrete uniform prob law:-

$$P(A) = \frac{3}{8}$$

Additional Probability Law 5

1 - if $A \subset B$, $P(A) \leq P(B)$

$$B = A \cup (B \cap \bar{A})$$



$$P(B) = P(A \cup (B \cap \bar{A}))$$

disjoint

$$= P(A) + P(B \cap \bar{A})$$

$$\geq 0$$

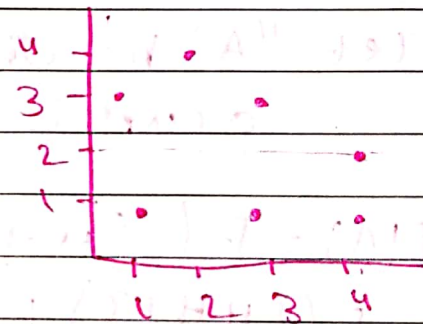
$$P(B) \geq P(A) \quad \#$$

Ex:

rolling a pair of 4-sided dice

$$P(\{\text{sum is even}\}) = \frac{8}{16} = \frac{1}{2}$$

$$P(\{\text{sum is odd}\}) = 1 - \frac{1}{2} = \frac{1}{2}$$



$$P(\{\text{1st roll} = \text{2nd roll}\}) = \frac{6}{18}$$

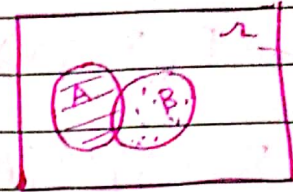
$$P(\{\text{at least one roll} = 4\}) = \frac{7}{16}$$

→ Additional Probability laws

$$2 - P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Proof:

$$A \cup B = A \cup (B \cap \bar{A})$$



$$P(A \cup B) = P(A \cup (B \cap \bar{A}))$$

$$= P(A) + P(B \cap \bar{A}) \quad \text{--- (1)}$$

$$P(B) = P((A \cap B) \cup (B \cap \bar{A})) = P(A \cap B) + P(B \cap \bar{A}) \quad \text{--- (2)}$$

From (1) and (2)

$$P(A \cup B) - P(A) = P(B) - P(A \cap B)$$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

$$P(A \cup B) \leq P(A) + P(B)$$

In general

$$P(A_1 \cup A_2 \cup \dots \cup A_n) \leq \sum_{k=1}^n P(A_k)$$

$$* P(A \cup B \cup C) = P(A) + \underbrace{P(B \cup C)}_I + \underbrace{P((A \cap C) \cup (B \cap \bar{A}))}_{II}$$

$$I = P(B) + P(C) - P(B \cap C)$$

$$II = P(A \cap C) - P(B \cap A) + P(A \cap B \cap C)$$

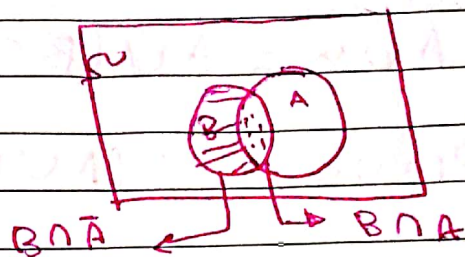
$$P(A \cup B \cup C) = P(A) + P(B) - P(A \cap B) - P(A \cap C) - P(B \cap C) + P(A \cap B \cap C)$$

$$4- P(B \cap \bar{A})$$

$$B = (B \cap \bar{A}) \cup (A \cap B)$$

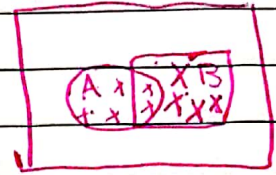
$$P(B) = P(B \cap \bar{A}) + P(A \cap B)$$

$$P(B \cap \bar{A}) = P(B) - P(A \cap B)$$



Conditional Probability \rightarrow the main idea of conditioning

- assuming that all elements from both events are equally likely.



Probability of B

$$P(B) = \frac{6}{9}$$

* What is the probability of "A" given "B" has occurred?

$$P(A/B) = \frac{P(A \cap B)}{P(B)} = \frac{2}{6}$$

* Definition =

$$P(A/B) = \frac{P(A \cap B)}{P(B)} \quad ; \quad P(B) > 0 \rightarrow P(B) \text{ must be greater than } 0 \text{ because we can't condition on an empty set.}$$

\rightarrow Results from the definition: - set.

$$\boxed{1} \quad P(A/B) \geq 0$$

$$\boxed{2} \quad P(B/B) = 1$$

$$\boxed{3} \quad P(\Omega/B) = 1$$

$$\boxed{4} \quad P(B/\Omega) = P(B)$$

$\boxed{5}$ If A_1, A_2 are disjoint events, then: -

$$P((A_1 \cup A_2)/B) = P(A_1/B) + P(A_2/B)$$

proof [5]

- From the definition: $P((A_1 \cup A_2)/B) = \frac{P((A_1 \cup A_2) \cap B)}{P(B)}$

$$\frac{P((A_1 \cup A_2) \cap B)}{P(B)}$$

$$= \frac{P(A_1 \cap B) + P(A_2 \cap B)}{P(B)} = \frac{P(A_1 \cap B)}{P(B)} + \frac{P(A_2 \cap B)}{P(B)}$$

$$= P(A_1/B) + P(A_2/B)$$

// In general.

→ If A_1, A_2, \dots, A_n are disjoint events then

$$P(A_1 \cup A_2 \cup \dots \cup A_n / B) = \sum_{i=1}^n P(A_i / B)$$

→ If A_1, A_2, \dots, A_n are not disjoint events then

$$P(A_1 \cup A_2 \cup \dots \cup A_n / B) \leq \sum_{i=1}^n P(A_i / B)$$

example $n=2$ then

$$P((A_1 \cup A_2)/B) \leq P(A_1/B) + P(A_2/B)$$

example:

A conservative design team, call it "c", and an innovative design team call it "n", are asked to separately design a new product within a month. From a past experience we know that:

(a) the probability that team "c" is successful $2/3$

(b) " " " " "n" " " " " $1/2$

(c) " " " " " at least one team is successful equals $3/4$

- Assuming that exactly one design is produced, what is the probability that it was done by "n"?

Sol :-

- The sample space for the success/fail for team C/n respectively is $\Omega = \{ss, sf, fs, ff\}$

- The desired probability is

$$P(Fs / \{sf, fs\}) = \frac{P(Fs \cap \{sf, fs\})}{P(\{sf, fs\})} = \frac{P(Fs \cap \{sf, fs\})}{P(sf) + P(fs)}$$

● First: Probability that "c" is successful :-

$$P(\{ss\}) + P(\{sf\}) = 2/3 \rightarrow \boxed{1}$$

● Second: Probability that N is successful

$$P(\{ss\}) + P(\{fs\}) = 1/2 \rightarrow \boxed{2}$$

● Third: Probability that at least one is successful

$$P(\{ss\}) + P(\{sf\}) + P(\{fs\}) = \frac{3}{4} \rightarrow \boxed{3}$$

● Forth: $P(\Omega) = 1$

$$P(\{ss\}) + P(\{sf\}) + P(\{fs\}) + P(\{ff\}) = 1$$

~~Substituting~~

→ substituting [3] in [4]

$$\frac{3}{4} + P(\{FF\}) = 1 \rightarrow P(\{FF\}) = \frac{1}{4}$$

→ substituting [2] in [3]

$$P(\{SF\}) + \frac{1}{2} = \frac{3}{4} \rightarrow P(\{SF\}) = \frac{1}{4}$$

→ substituting [1] in [4]

$$P(\{FS\}) + \frac{2}{3} + \frac{1}{4} = 1 \rightarrow P(\{FS\}) = \frac{1}{12}$$

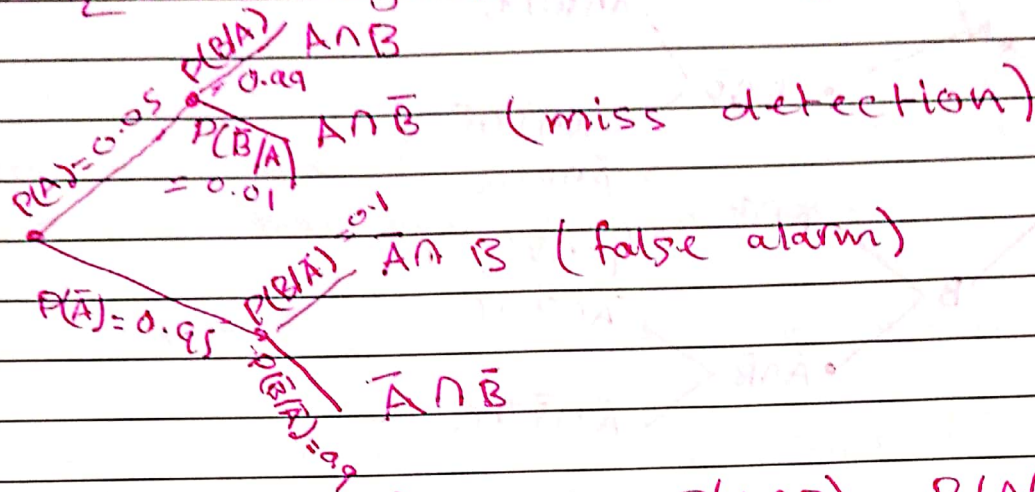
Therefore the desired probability :-

$$\frac{P(\{FS\})}{P(\{SF\}) + P(\{FS\})} = \frac{\frac{1}{12}}{\frac{1}{4} + \frac{1}{12}} = \frac{\frac{1}{12}}{\frac{4}{12}} = \frac{1}{4}$$

Radar detection

$A = \{ \text{an Aircraft is flying above} \}$

$B = \{ \text{something has been registered of radar screen?} \}$



$$P(A|B) = \frac{P(A \cap B)}{P(B)} \rightarrow P(A \cap B) = P(A|B) \cdot P(B) = P(B|A) \cdot P(A)$$

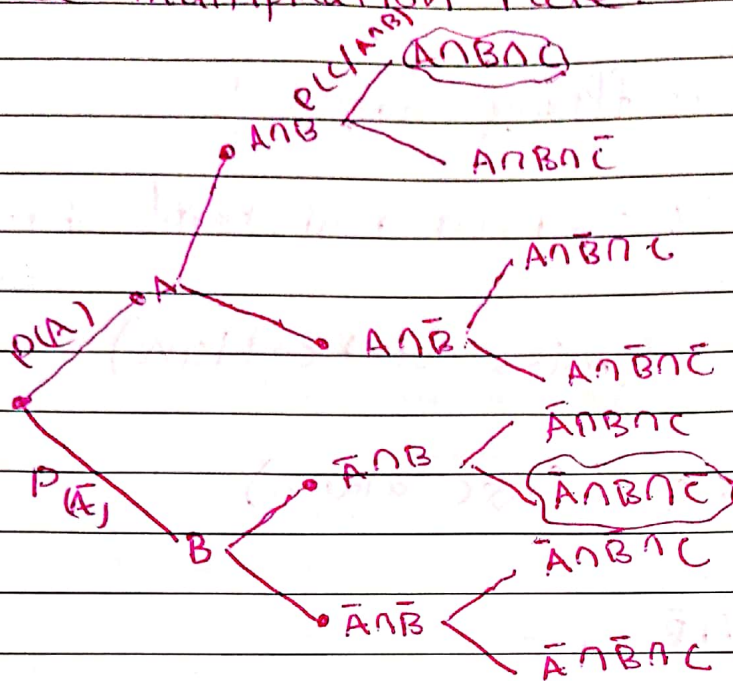
$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

$$1 - P(A \cap B) = P(B|A) \cdot P(A) = 0.99 \times 0.05 = \dots$$

$$2 - P(\text{missed detection}) = P(A \cap \bar{B}) = P(\bar{B}|A) \cdot P(A) = 0.01 \times 0.05$$

$$3 - P(A|B) = \frac{P(A \cap B)}{P(B)} \quad (\text{Next lectures we will answer this})$$

The multiplication rule:-



$$P(A \cap B \cap C) = P(A) \cdot P(B|A) \cdot P(C|A \cap B)$$

$$P(\bar{A} \cap B \cap \bar{C}) = P(\bar{A}) \cdot P(B|\bar{A}) \cdot P(\bar{C}|\bar{A} \cap B)$$

using the definition

$$P(A \cap B \cap \bar{C}) = P(\bar{C}|\bar{A} \cap B) \cdot P(\bar{A} \cap B)$$

$$\hookrightarrow P(B|\bar{A}) \cdot P(\bar{A})$$

$$= P(\bar{C}|\bar{A} \cap B) \cdot P(B|\bar{A}) \cdot P(\bar{A})$$

The general form

$$P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1) \cdot \prod_{i=2}^n P(A_i / A_1 \cap A_2 \cap \dots \cap A_{i-1})$$

$n=3$

$$P(A_1 \cap A_2 \cap A_3) = P(A_1) \times P(A_2 / A_1) \times P(A_3 / A_1 \cap A_2)$$

example:-

Three cards are drawn from 52-card ^{deck} without replacement
Probability that none of these cards is a heart?

let $A_i = \{ \text{The } i\text{th card is not a heart} \}$

$$P(A_1 \cap A_2 \cap A_3) = P(A_1) \cdot P(A_2 / A_1) \cdot P(A_3 / A_1 \cap A_2)$$

$$= \frac{39}{52} \cdot \frac{38}{51} \cdot \frac{37}{50}$$

Example:

- Three cards are drawn from 52 card deck without replacement. What is the probability that none of them is a heart.

Solution

Let A_i be defined as $A_i = \{ \text{The } i^{\text{th}} \text{ card is not a heart} \}$

Probability that none of them is not heart is

$$P(A_1 \cap A_2 \cap A_3) = P(A_1)P(A_2/A_1)P(A_3/A_1 \cap A_2)$$

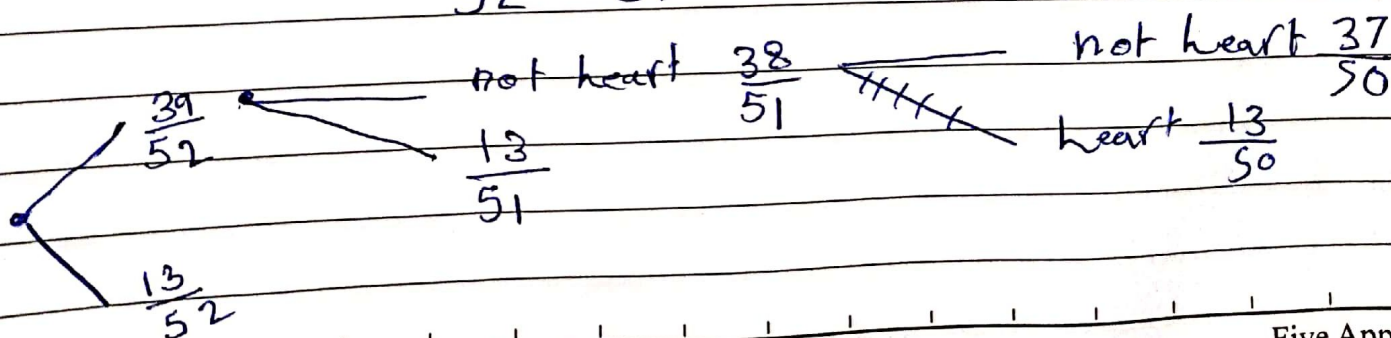
$$= \frac{39}{52} \times \frac{38}{51} \times \frac{37}{50}$$

- What is the probability that the first one is not a heart and the second one is a heart?

$$P(A_1 \cap \bar{A}_2) = P(\bar{A}_2/A_1) \cdot P(A_1) = \frac{13}{51} \times \frac{39}{52}$$

- What is the pro. that the first two cards are not hearts and the third is heart

$$P(A_1 \cap A_2 \cap \bar{A}_3) = \frac{39}{52} \times \frac{38}{51} \times \frac{13}{50} \quad (\text{Tree diagram})$$

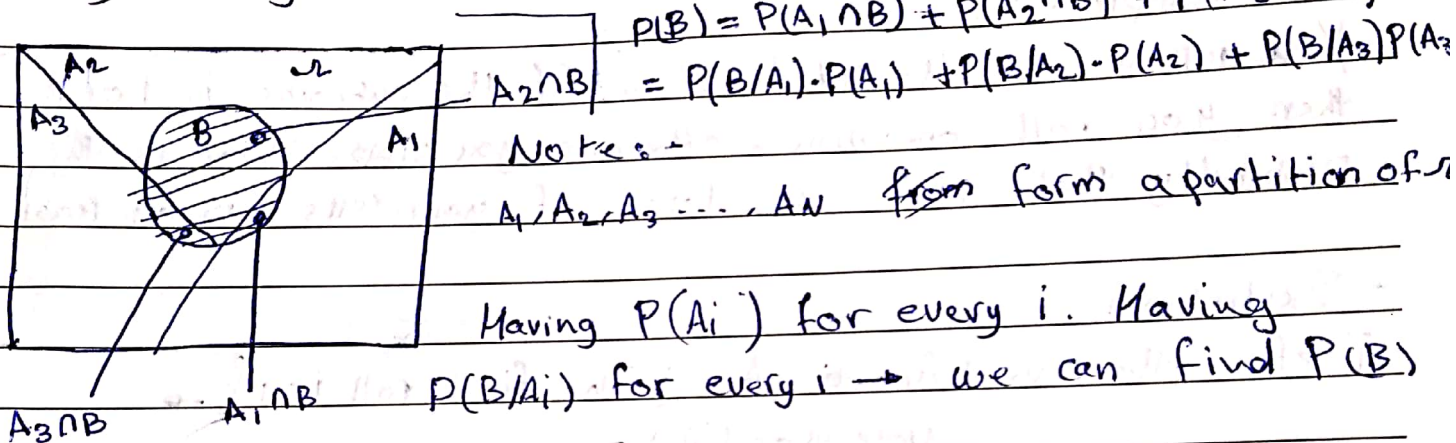


Total Probability Theorem:

Let A_1, A_2, \dots are disjoint events that form a partition of Ω . Then for any event B . We have:

$$P(B) = P(A_1 \cap B) + P(A_2 \cap B) + \dots \rightarrow P(B/A_1) \cdot P(A_1) + P(B/A_2) \cdot P(A_2) + \dots \rightarrow \sum_{j=1}^N P(B/A_j) \cdot P(A_j)$$

By using the total probability theorem \rightarrow we find $P(B)$



* * * From the radar detection example

$$P(B) = P(\{\text{something has been registered}\}) = P(B \cap A) + P(B \cap \bar{A})$$

$$= P(B/A) \cdot P(A) + P(B/\bar{A}) \cdot P(\bar{A})$$

$$= 0.99 \times 0.05 + 0.1 \times 0.95 = 0.1445$$

Baye's Rule: $P(A_i/B) = \frac{P(A_i \cap B)}{P(B)}$

$P(A_i \cap B) = P(B/A_i) \cdot P(A_i)$ (Multiplication rule)

$P(B) = \sum_{i=1}^n P(B/A_i) \cdot P(A_i)$
by total probability theorem

→ From the radar detection example

$$P(A/B) = \frac{P(B/A) \cdot P(A)}{P(B)} = \frac{0.99 \times 0.05}{0.1445}$$

→ Example:

You roll a fair 4-sided die. If the outcome is 1 or 2 then you roll one more - otherwise you stop. What is the probability that the sum total of your rolls is at least 4.

- Solution

Define the event $A_i \rightarrow A_i = \{ \text{The first roll is } i \} \rightarrow$

Note that $P(A_i) = \frac{1}{4}$

Define the event B : The sum total is at least 4

$P(B/A_3) = 0$ - A_3 means the roll is 3

$P(B/A_4) = 1$

$$P(B/A_2) = \frac{1}{4} + \frac{1}{4} + \frac{1}{4} = \frac{3}{4}$$

$$P(B/A_1) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$$

$$P(B) = P(B/A_1) \cdot P(A_1) + P(B/A_2) \cdot P(A_2) + P(B/A_3) \cdot P(A_3) + P(B/A_4) \cdot P(A_4)$$

→ If the sum is at least 4, what is the probability that the first roll was 1?

$$P(A_1/B) = \frac{P(B/A_1) \cdot P(A_1)}{P(B)} = \frac{\frac{1}{2} \cdot \frac{1}{4}}{\frac{9}{16}} = \frac{2}{9}$$

Independence: Intuitively $\neq P(B/A) = P(B) \rightarrow [1]$

↓
the occurrence of event A provides no information about the occurrence of event B.

- From the definition of conditional Probability

$$P(B/A) = \frac{P(B \cap A)}{P(A)} \rightarrow P(B \cap A) = P(A) \cdot P(B/A)$$

But $P(B/A) = P(B)$

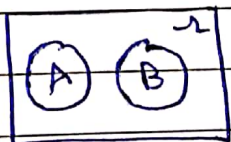
Therefore: $P(B \cap A) = P(A) \cdot P(B)$ (Definition of independence)

→ From [1] and [2] $P(A/B) = P(A)$

→ Two events are independent if their intersection is equal to their individual probability.

- Disjoint events cannot be independent: Disjoint events totally dependent.

- When talking about non-zero events.



$$A \cap B = \emptyset \quad P(A) \times P(B) > 0$$
$$P(A \cap B) \neq P(A) \times P(B)$$

A and B are not independent

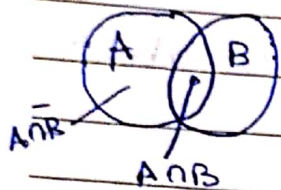
- If A and B are independent then:

- (1) \bar{A} and \bar{B} are also independent
- (2) \bar{A} and B are " "
- (3) A and \bar{B} " "

• Prove that if A and B are independent then A and \bar{B} are also independent

We want to show that $P(A \cap \bar{B}) = P(A) \cdot P(\bar{B})$

$$A = (A \cap \bar{B}) \cup (A \cap B)$$



$$P(A) = P(A \cap \bar{B}) + P(A \cap B)$$

$$P(A) = P(A \cap \bar{B}) + P(A) \cdot P(B)$$

$$P(A)(1 - P(B)) = P(A \cap \bar{B})$$

$$P(A \cap \bar{B}) = P(A) \cdot P(\bar{B})$$

Note: Independent event cannot be visualised using Venn diagram.

Conditional independence

$$P(A \cap B | C) = P(A | C) \cdot P(B | C) \quad \text{--- [1] (conditionally independence)}$$

From definition of conditional probability

$$P(A \cap B | C) = \frac{P(A \cap B \cap C)}{P(C)} \quad \text{(Multiplication Rule)}$$

$$P(A \cap B | C) = \frac{P(C) \cdot P(B | C) \cdot P(A | B \cap C)}{P(C)}$$

$$P(A \cap B | C) = P(B | C) \cdot P(A | B \cap C) \quad \text{--- [2]}$$

$$\text{[1]} = \text{[2]}$$

$$P(A|C) \cdot P(B|C) = P(B|C) \cdot P(A|B \cap C) \quad (\text{assume } P(B|C) > 0)$$

$$P(A|B \cap C) = P(A|C)$$

The occurrence of B doesn't change our beliefs about P(A|C)

- Independence of several events

A_1, A_2, \dots, A_n are independent if $P(\bigcap_{i \in S} A_i) = \prod_{i \in S} P(A_i)$ for every subset of $S = \{1, 2, \dots, n\}$

- If $S = \{1, 2\} \rightarrow P(A_1 \cap A_2) = P(A_1) P(A_2)$

- If $S = \{1, 2, 3\} \rightarrow$ To (prove) that all A_1, A_2 and A_3 are independent then:

$$P(A_1 \cap A_2 \cap A_3) = P(A_1) P(A_2) P(A_3) \rightarrow [1]$$

$$P(A_1 \cap A_2) = P(A_1) P(A_2) \rightarrow [2]$$

$$P(A_1 \cap A_3) = P(A_1) P(A_3) \rightarrow [3]$$

$$P(A_2 \cap A_3) = P(A_2) P(A_3) \rightarrow [4]$$

If all ~~stat~~ true we say that A_1, A_2 and A_3 are pair-wise independent

Pair-wise independent does not imply independence

example 1:- Two successive rolls of a 4-sided die (fair)

let $A_i = \{ \text{First roll is } i \}$

$B_j = \{ \text{second roll is } j \}$

1) Are A_i and B_j independent

$$P(A_i) = \frac{1}{4}$$

$$P(A_i) P(B_j) = \frac{1}{4} \cdot \frac{1}{4} = \frac{1}{16}$$

$$P(B_j) = \frac{1}{4}$$

$$P(A_i \cap B_j) = \frac{1}{16}$$

Yes, since $P(A_i \cap B_j) = P(A_i) P(B_j)$

[2] Let $A = \{ \text{First roll is 1} \}$ $P(A) = \frac{1}{4}$

$B = \{ \text{sum of the two rolls is 5} \}$

$B = \{ (1, 4), (4, 1), (2, 3), (3, 2) \}$

$P(A) \times P(B) = \frac{1}{4} \times \frac{1}{4} = \frac{1}{16}$

$P(B) = \frac{4}{16}$

~~$P(A) \cap B$~~

$P(A \cap B) = 1/16$

Yes, since $P(A \cap B) = P(A) \cdot P(B)$

[3] $A = \{ \text{The maximum of two rolls is 2} \}$

$B = \{ \text{minimum is 2} \}$

$A = \{ (1, 2), (2, 1), (2, 2) \}$ $P(A) = \frac{3}{16}$

$B = \{ (2, 2), (2, 3), (3, 2), (4, 2), (2, 4) \}$ $P(B) = \frac{5}{16}$

$P(A) \times P(B) = \frac{3 \times 5}{16 \times 16} = \frac{15}{256}$

$P(A \cap B) = \frac{1}{16}$

$P(A \cap B) \neq P(B) P(A)$ (dependent)

Example 2: Two independent fair coin Tosses

$H_1 = \{ \text{first toss is H} \}$ $H_2 = \{ \text{second toss is H} \}$

$D = \{ \text{both tosses have different results} \}$

- check if H_1 and H_2 are conditionally independent given D .

- We need to check if

$$P(H_1 \cap H_2 / D) = P(H_1 / D) P(H_2 / D)$$

$$P(H_1) = \frac{1}{2} \quad P(H_2) = \frac{1}{2}$$

$$P(H_1 \cap H_2) = \frac{1}{2} \times \frac{1}{2} = P(H_1) P(H_2) \rightarrow \text{independent}$$

$$P(H_1 / D) = \frac{1}{2}$$

$$P(H_2 / D) = \frac{1}{2}$$

H	H	H	T
T	H	T	T

$$P(H_1 \cap H_2 / D) = 0 \neq P(H_1 / D) \cdot P(H_2 / D)$$

Pair-wise independence does not imply independence
 H_1 and H_2 indep.

H_1, D indep. $\rightarrow P(H_1 / D) = P(H_1) = \frac{1}{2}$ (from Definition)

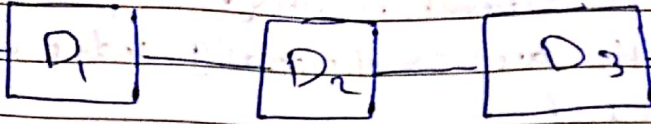
H_2, D indep. $\rightarrow P(H_2 / D) = P(H_2) = \frac{1}{2}$ (" ")

$H_1, H_2, D \rightarrow P(H_1 \cap H_2 \cap D) = 0 \neq P(H_1) P(H_2) P(D)$

H_1, H_2, D are not indep. \rightarrow but they are pair-wise indep.

Applications of independence

• Reliability



- let P_i denote the probability that D_i is up
- What is the probability that the system is up?

$U_i = D_i$ is up

$$P(U_1 \cap U_2 \cap U_3) = P(U_1) \times P(U_2) \times P(U_3)$$

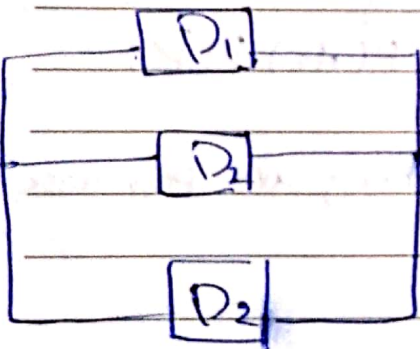
(since devices fail independently)

For this system to be up all D_1, D_2, D_3 must be up.

$$\rightarrow P_1 \times P_2 \times P_3$$

- What is the probability that the system is down

$$= 1 - (P_1 \times P_2 \times P_3)$$



$$P(U_1 \cap U_2 \cap U_3) = P(\bar{U}_1 \cap \bar{U}_2 \cap \bar{U}_3)$$

$$= 1 - P(\bar{U}_1 \cap \bar{U}_2 \cap \bar{U}_3)$$

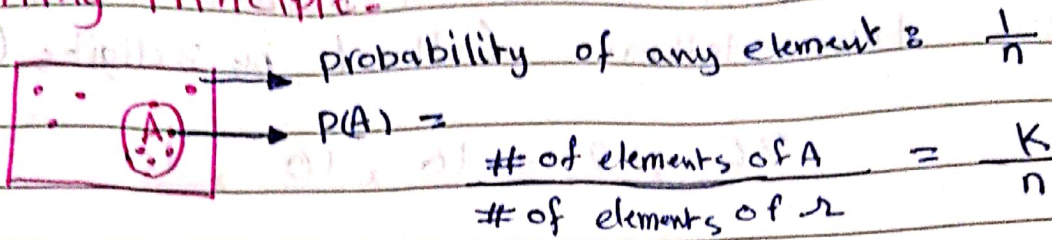
$$= 1 - P(\bar{U}_1) P(\bar{U}_2) P(\bar{U}_3)$$

$$= 1 - (1 - P(U_1)) (1 - P(U_2)) (1 - P(U_2))$$

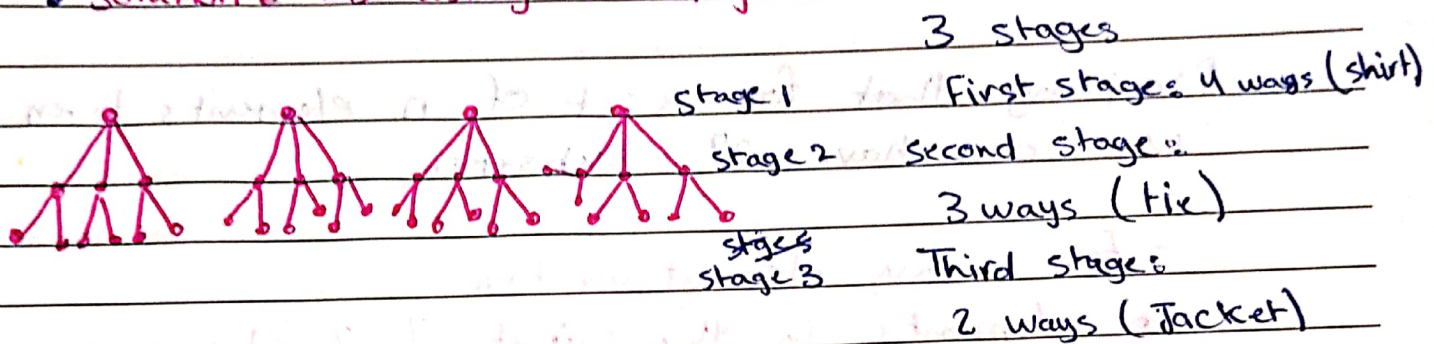
$$= 1 - P_1 \times P_2 \times P_3$$

Probability that the system is up

Counting Principle:



- Example: choosing from: 4 shirts, 3 ties, 2 jackets
- solution: By using tree diagram



let us define:

n_i : The number of choices at stage i

Total # of choices = $n_1 \times n_2 \times \dots \times n_n$

for n stages

r stage $\rightarrow n$ is the # of choices at the " i " th

stage \rightarrow total # of choices = $\prod_{i=1}^r n_i$

- Example: The # of license plates with two letters followed by 3 digits (repetition allowed)

$$26 \times 26 \times 10 \times 10 \times 10$$

- Same about (but repetition not allowed)

$$26 \times 25 \times 10 \times 9 \times 8$$

Providing that for a set of n elements then we can have 2^n subsets.

- For any element we have

- elements in the subset } 2 choice \rightarrow
- element not in the subset } Repeated for n stages.

$$2 \times 2 \times 2 \dots \times 2 = 2^n$$

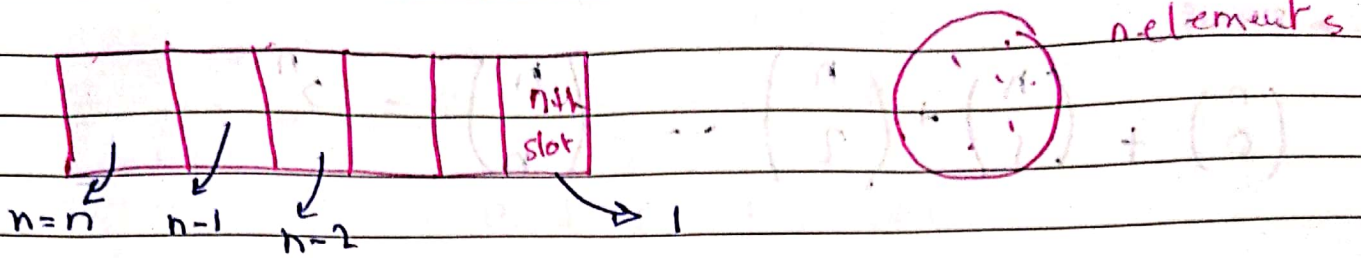
first stage nth stage

example:

for $S = \{1, 2, 3\} \rightarrow 2^3 = 8$

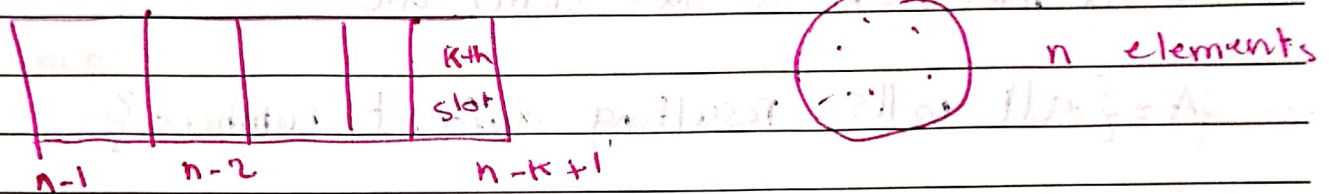
$$\{1\}, \{2\}, \{3\}, \{1, 2, 3\}, \{1, 2\}, \{2, 3\}, \{1, 3\}$$

* Permutations: the number of ways to order n elements



$$\# \text{ of ways} = n(n-1)(n-2) \dots = n!$$

K-permutation



$$\# \text{ of way} = n(n-1)(n-2) \dots (n-k+1)$$

$$= \frac{n!}{(n-k)!}$$

* Combination:

$\binom{n}{k}$ = # of k -elements subsets of an n -element set

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

$$1 - \binom{n}{0} = 1 \quad 3 - \binom{n}{n-1} = n$$

$$2 - \binom{n}{1} = n \quad 4 - \sum_{k=0}^n \binom{n}{k} = 2^n$$

$$\sum_{k=0}^n \binom{n}{k} = 2^n$$

→ proof

$$\binom{n}{0} + \binom{n}{1} + \binom{n}{2} + \dots + \binom{n}{n} = 2^n$$

Example

6 six rolls of a six-sided die

$A = \{ \text{all rolls resulting different numbers} \}$

$$P(A) = \frac{\text{\# of elements of } A}{\text{\# of elements of } \Omega} = \frac{6!}{6^6} \quad \rightarrow \text{\# of ways to order 6 numbers}$$

Example

Tossing a coin "n" times - let $P(H) = P$

$$P(T) = 1 - P$$

$$P(\text{HHTHHT}) = P^4 \times (1-P)^2 = P \times P \times (1-P) \times P \times P \times (1-P)$$

$$P(k \text{ heads } (n-k) \text{ Tails}) = P^k (1-P)^{n-k}$$

particular seq sequence

$$P(k \text{ heads}) = P^k (1-P)^{n-k} \binom{n}{k}$$

Tossing a coin 3 times

$$\begin{array}{l}
 HHH \\
 \textcircled{HHT} \rightarrow P_1 = P \cdot P \cdot (1-P) = P^2(1-P) \\
 \textcircled{HTH} \rightarrow P_2 = P \cdot (1-P) \cdot P = P^2(1-P) \\
 HTT \\
 TTT \\
 \textcircled{THT} \rightarrow P_3 = P \cdot P \cdot (1-P) = P^2(1-P) \\
 THT \\
 THT \\
 TTT
 \end{array}$$

$$P(2 \text{ heads}) = 3 \times P \cdot P \cdot (1-P)$$

* Example :-

- n-coin Tosses \rightarrow P(k heads in a particular sequence)

$$= P^k (1-P)^{n-k}$$

$$P(k \text{ heads}) = P^k (1-P)^{n-k} \binom{n}{k}$$

* Example :

- Given that are 3 heads in 10 tosses. Find the probability that the first tosses are heads. Provided that $P(H) = P$

• Solution:-

$$P(T) = 1-P$$

Define event $A = \{ \text{The first two tosses are heads} \}$.

Define event $B = \{ \text{There are 3 heads in 10 tosses} \}$.

\rightarrow we need to find

$$P(A/B) = \frac{P(A \cap B)}{P(B)} \quad \text{--- [1]}$$

$$P(B) = P(\text{Three heads in 10 tosses}) = P^3 (1-P)^7 \binom{10}{3} \quad \text{--- [2]}$$

$$P(A \cap B) = P(H_1, H_2 \text{ one H in 8 tosses}) = P \cdot P \binom{8}{1} \cdot P (1-P)^7 \quad \text{--- [3]}$$

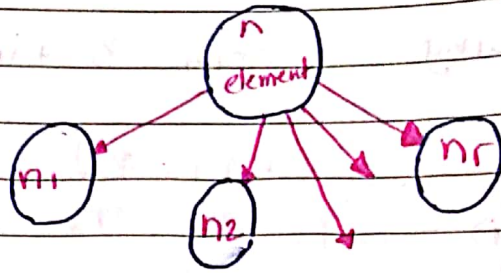
By substituting [2], [3] in [1]

$$P(A/B) = \frac{P^3 (1-P)^7 \binom{8}{1}}{P^3 (1-P)^7 \binom{10}{3}} = \frac{8}{\binom{10}{3}}$$

* Partitions :-

[1] $\binom{n}{n_1}$

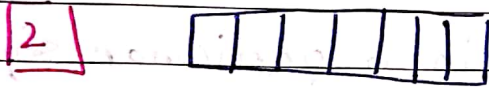
$$\binom{n - n_1 - n_2 \dots - n_{r-1}}{n_r}$$



→ note that: $n = n_1 + n_2 + n_3 + \dots$

$$C = \binom{n}{n_1} \cdot \binom{n - n_1}{n_2} \cdot \binom{n - n_1 - n_2}{n_3} \dots \binom{n - n_1 - n_2 \dots - n_{r-1}}{n_r}$$

$$C = \frac{n!}{n_1! \cdot n_2! \cdot n_3! \dots \cdot n_r!}$$



$$n! = C \cdot n_1! \cdot n_2! \cdot n_3! \dots \cdot n_r!$$

$$C = \frac{n!}{n_1! \cdot n_2! \cdot n_3! \dots \cdot n_r!}$$

* Example :-

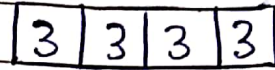
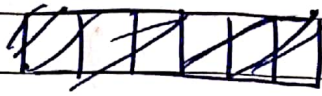
A class contains 12 undergraduate and 4 grad students, we randomly distribute the students on 4 groups of 4 students. What is the probability that each group has a grad student.

* Solution :-

Define event $A = \{ \text{each group has a grad. student} \}$.

$$P(A) = \frac{\# \text{ of elements of } A}{\# \text{ of elements of } \Omega} = \frac{|A|}{|\Omega|}$$

$$|\Omega| = \frac{16!}{4! \cdot 4! \cdot 4! \cdot 4!} \quad \text{--- [2]}$$



12 ~~set~~ undergraduate students

Take 4 grad students
4! --- [3]

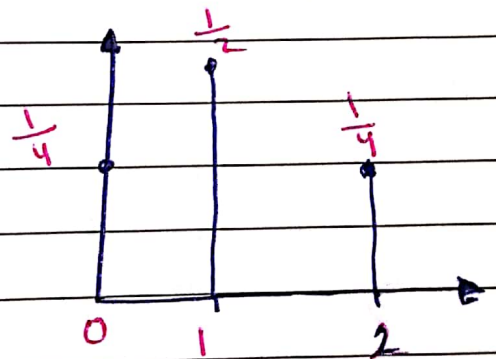
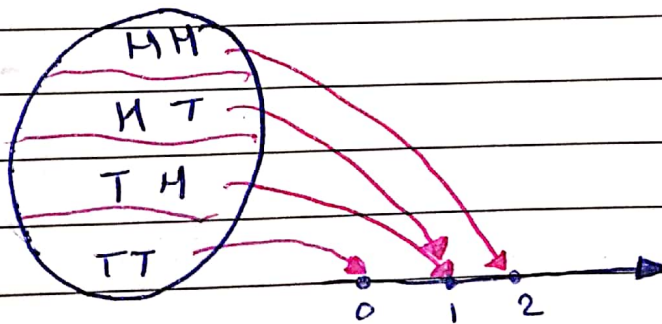
$$\rightarrow \frac{12!}{3! \cdot 3! \cdot 3!} \quad \text{--- [4]}$$

$$* |A| = [3] \cdot [4] = \frac{4! \cdot 12!}{3! \cdot 3! \cdot 3!} \quad \text{--- [5]} \rightarrow \text{By substituting [2] and [5] in [1]}$$

$$* P(A) = \frac{4! \cdot 12!}{3! \cdot 3! \cdot 3! \cdot 16!} = \frac{4! \cdot 4! \cdot 4! \cdot 4!}{4! \cdot 4! \cdot 4! \cdot 4!}$$

* Example :-

Tossing a coin twice (fair coin)



of heads.

Random Variable
 $X(\omega)$