



# PROBABILITY REFRESHER: +2 LEVEL CONCEPTS

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# CONTENTS

- Ref : NCERT Mathematics Textbooks for Std XI and XII
  - Latest versions can be found here : [NCERT](#)
- Sets (XI, Chapter 1)
- Permutations and Combinations (XI, Chapter 7)
- Probability (XI, Chapter 16)
- Probability Distributions (XII, Chapter 13 + additional concepts)

# Sets

- Foundation of relations and functions, sequences, geometry, probability theory, etc.
- A set is a well defined collection of objects.
  - Sets are usually denoted by upper case letters  $A, B, C$ , etc.
  - Elements of a set are represented by lower case letters  $a, b, c$ , etc.
  - If  $a$  is an element of a set  $A$ , we say that “ $a$  belongs to  $A$ ” the Greek symbol  $\in$  (epsilon) is used to denote the phrase ‘belongs to’. Thus, we write  $a \in A$ .
  - If ‘ $b$ ’ is not an element of a set  $A$ , we write  $b \notin A$  and read “ $b$  does not belong to  $A$ ”.
- A few examples of sets used particularly in mathematics, viz.
  - $\mathbf{N}$  : the set of all natural numbers
  - $\mathbf{Z}$  : the set of all integers ,  $\mathbf{Z}^+$  : the set of positive integers
  - $\mathbf{Q}$  : the set of all rational numbers ,  $\mathbf{Q}^+$  : the set of positive rational numbers
  - $\mathbf{T}$  : the set of all irrational numbers
  - $\mathbf{R}$  : the set of real numbers ,  $\mathbf{R}^+$  : the set of positive real numbers
  - $\mathbf{C}$  : the set of all complex numbers

# Sets

- Empty Set /Null Set/Void Set
  - A set which does not contain any elements.
  - The empty set is denoted by the symbol  $\phi$  or  $\{ \}$ .
- Finite and Infinite sets
  - A set which is empty or consists of a definite number of elements is called finite
  - Otherwise, the set is called infinite.
- Equal Sets
  - Two sets A and B are said to be equal if they have exactly the same elements,  $A = B$ .
  - Otherwise, the sets are said to be unequal and we write  $A \neq B$
- Equivalent Sets
  - Two finite sets A and B are said to be equivalent if they have the same number of elements,  $A \leftrightarrow B$
  - For Example  $A = \{1, 2, 3, 4, 5\}$ ,  $B = \{a, b, c, d, e\}$

# Sets

- Subsets

- A set  $A$  is said to be a subset of a set  $B$  if every element of  $A$  is also an element of  $B$ .
  - $A \subset B$  if whenever  $a \in A$ , then  $a \in B$
  - $A \subset B, a \in A \Rightarrow a \in B$ , for all  $a \in A$
- Every set  $A$  is a subset of itself, i.e.,  $A \subset A$  ( $A$  is called improper subset of  $A$ ).
- Since the empty set  $\phi$  has no elements,  $\phi$  is a subset of every set
- If  $A \subset B$  and  $A \neq B$ , then  $A$  is called a proper subset of  $B$  and  $B$  is called superset of  $A$
- Proper subsets of  $A$  include all subsets of  $A$  including  $\phi$  except the improper subset
- If a set has “ $n$ ” elements, then the number of subset of the given set is  $2^n$
- **$N \subset Z \subset Q, Q \subset R, T \subset R, N \not\subset T$**

- Power Set

- The set of all subsets of set  $A$  is called the power set.
- This includes the improper subset and all proper subsets.

# Sets

- Universal Set
  - Usually, in a particular context, we have to deal with the elements and subsets of a basic set which is relevant to that particular context. This basic set is called the “Universal Set”.
  - The universal set is usually denoted by  $U$ , and all its subsets by the letters  $A, B, C$ , etc.
  - For example, in human population studies, the universal set consists of all the people in the world
- Operations on sets
  - Union of Sets :  $A \cup B$ , elements present in both sets
  - Intersection of Sets :  $A \cap B$ , elements common to both sets
  - Set Difference :  $A - B$ , remove  $A \cap B$  from  $A$
  - Symmetric Difference :  $A \Delta B = (A - B) \cup (B - A)$

# Combinatorics

- Fundamental Principle of Counting (Multiplication Principle)
  - If an event can occur in  $m$  different ways, following which another event can occur in  $n$  different ways, then the total number of occurrence of the events in the given order is  $m \times n$ .
  - The above principle can be generalized for any finite number of events.
  - For example, for 3 events, the principle is as follows: If an event can occur in  $m$  different ways, following which another event can occur in  $n$  different ways, following which a third event can occur in  $p$  different ways, then the total number of occurrence to the events in the given order is  $m \times n \times p$ .
  - How many 3 letter code words are possible using the first 10 letter of English Alphabet, if
    - No letter can be repeated –  $10 \times 9 \times 8 = 720$
    - Letters are repeated –  $10 \times 10 \times 10 = 1000$
- Factorial notation
  - $n!$  represents the product of first  $n$  natural numbers
  - $n! = n \times (n - 1) \times (n - 2) \times \dots \times 2 \times 1 = n \times (n - 1)!$
  - $0! = 1$

# PERMUTATIONS

- A permutation is an arrangement in a definite order of a number of objects taken some or all at a time.
- Permutations when all objects are distinct
  - The number of permutations of  $n$  different objects, taken all at a time is given by  $P(n, n) = n!$
  - The number of permutations of  $n$  different objects taken  $r$  at a time, where  $0 < r \leq n$  and the objects do not repeat is  $n (n - 1) (n - 2) \dots (n - r + 1)$ , which is denoted by  $P(n, r)$ .
  - $P(n, r) = n! / (n - r)!$
- Permutations when all objects are not distinct
  - The number of permutations of  $n$  objects, where  $p_1$  objects are of one kind,  $p_2$  are of second kind, ...,  $p_k$  are of  $k$ th kind and the rest, if any, are of different kind is  $n! / (p_1! \times p_2! \times \dots \times p_k!)$
- Circular permutations
  - No of distinguishable circular permutations for  $n$  distinct objects arranged in a circle is  $(n - 1)!$



# COMBINATIONS

- A combination is a selection of some of all of a number of different objects. The order of selection of the objects is immaterial.
- The number of combinations of  $n$  distinct objects, taken  $r$  at a time, denoted by  $C(n, r)$ 
  - $C(n, r) = n! / (r! \times (n-r)!) = P(n, r) / r!$
  - $C(n, n) = C(n, 0) = 1$
  - $C(n, r) = C(n, n - r)$
- Pascal's rule
  - $C(n+1, r) = C(n, r) + C(n, r-1)$

# COMBINATORIAL PROBLEMS

- No of options to choose from :  $n$
- Ordered samples of size  $r$ , without replacement
  - Permutation, No of possible Outcomes =  $P(n, r)$
  - Example : No of 3 digit numbers that can be formed with  $1, 2, \dots, 9$  without repetition
  - Example : Choose 2 balls from a bag containing 5 numbered balls, order of numbers matters
- Unordered samples of size  $r$ , without replacement
  - Combination, No of possible Outcomes =  $C(n, r)$
  - Example : Choose 2 balls from a bag containing 5 numbered balls, order of numbers do not matter

# COMBINATORIAL PROBLEMS

- No of options to choose from :  $n$
- Ordered samples of size  $r$ , with replacement
  - No of possible Outcomes =  $n^r$
  - Example : Three dice are rolled together or a dice is rolled three times, order of numbers matters
  - Example : Choose 1 ball from a bag containing 5 numbered balls 2 times, order of numbers matters
- Unordered samples of size  $r$ , with replacement
  - Combination, No of possible Outcomes =  $C(n + r - 1, r)$
  - Example : Three dice are rolled together or a dice is rolled three times, order of numbers do not matter
  - Example : Choose 1 ball from a bag containing 5 numbered balls 2 times, order of numbers do not matter

# Basic Probability theory

- Probability is the measure of uncertainty of random experiments.
- Classical Definition : The probability of an event is the number of outcomes favorable to the event , divided by the total number of outcomes, where all outcomes are equally likely
  - It only considers experiments with a finite number of outcomes (and hence restrictive)
  - All outcomes are considered to be equally likely (again this is a circular definition since we are using the concept of probability to define probability itself)
  - Both these conditions are valid if we consider classical probability problems like tossing of an (**unbiased**) coin, rolling of (**fair**) die, or picking a card from a (**perfectly shuffled**) deck of cards
  - But this classical definition cannot be used in the construction of a mathematical theory of probability

# Basic Probability theory

- Frequentist / Statistical approach of probability :
  - Find the probability on the basis of observations and collected data
  - Frequentist statistics uses rigid frameworks, the type of frameworks that you learn in basic statistics, like p-values and confidence intervals.
  - Frequentist probability has more applicability than the classical model, but is still very limited
- Axiomatic Approach of Probability
  - It sets down a set of axioms (rules) that apply to all of types of probability, including frequentist probability and classical probability.
  - These rules, based on Kolmogorov's Three Axioms, set starting points for mathematical probability.
  - Kolmogorov's Three Axioms
    1. For any event  $A$ ,  $P(A) \geq 0$
    2.  $P(S) = 1$ , where  $S$  is the sample space of an experiment; i.e., the set of all possible outcomes
    3. If  $A$  and  $B$  are mutually exclusive events, then  $P(A \cup B) = P(A) + P(B)$ .

# Basic Probability theory

- Random Experiment : An experiment is called random experiment if it satisfies the following two conditions:
  1. It has more than one possible outcome.
  2. It is not possible to predict the outcome in advance.
- Outcomes and Sample Space ( $S$ )
  - A possible result of a random experiment is called its outcome. Each outcome of the random experiment is also called sample point.
  - The set of all possible outcomes of a random experiment is called the sample space associated with the experiment
  - A given performance of the experiment must produce a result corresponding to exactly one of the points of  $S$ .
- Event : An event is a subset of the sample space, i.e. a collection of points of the sample space.
- Occurrence of an event
  - The event  $E$  of a sample space  $S$  is said to have occurred if the outcome  $\omega$  of the experiment is such that  $\omega \in E$ .
  - If the outcome  $\omega$  is such that  $\omega \notin E$ , we say that the event  $E$  has not occurred

# Types of events

- Simple event: A simple event or an elementary event is an event containing only a single sample point.
- Compound events: Compound events or decomposable events are those events that are obtained by combining together two or more elementary events.
  - For instance, the event of drawing a heart from a deck of cards is the subset  $A = \{\text{heart}\}$  of the sample space  $S = \{\text{heart, spade, club, diamond}\}$ . Therefore,  $A$  is a simple event.
  - The event  $B$  of drawing a red card is a compound event since  $B = \{\text{heart} \cup \text{diamond}\} = \{\text{heart or diamond}\}$ .
- Mutually exclusive or disjoint events: Events are said to be mutually exclusive or disjoint or incompatible if the occurrence of any one of them prevents the occurrence of all the other events.
  - Getting a head and getting a tail in toss of a coin are mutually exclusive
- Mutually non-exclusive events: The events which are not mutually exclusive are known as compatible events or mutually non-exclusive events.
  - Getting a card of heart suite and getting a red card are not mutually exclusive.

# Types of events

- Independent events: Events are said to be independent, if the happening (or non-happening) of one event is not affected by the happening (or non-happening) of other events.
  - If I toss two coins simultaneously, the outcomes of both trials are independent of each other
- Dependent events: Two or more events are said to be dependent, if the happening of one event affects (partially or totally) the other event.
- Exhaustive Events: A Set of events is said to be exhaustive if the performance of random experiments always results in the occurrence of at least one of them.
  - For instance, consider an ordinary pack of cards. The events 'drawn card is heart', 'drawn card is diamond', 'drawn card is club' and 'drawn card is spade' is a set of events that is exhaustive.
  - In other words all sample points put together (i.e. sample space itself) would give us an exhaustive event.
  - If 'E' is an exhaustive event/sure event then  $P(E) = 1$ .



# Example

- Random Experiment : Toss a coin
- Sample Space ( $\Omega$ ) : In tossing of a coin, the sample space for the number that shows up on the top face would be :  $\Omega = \{H, T\}$
- For the coin tossing case,  $(2^2) = 4$  subsets can be formed out of that sample space.
  - These subsets are:  $\Phi$ ,  $\{H\}$ ,  $\{T\}$  and  $\{H, T\}$ .
- Here getting a head or tail are equally likely events.
- Event :
  - Getting a head in a single toss of a coin,  $E1 = \{H\} \subset \Omega$
  - Getting a tail in a single toss of a coin,  $E2 = \{T\} \subset \Omega$
  - Getting a head or a tail in a single toss of a coin,  $E3 \subseteq \Omega$
  - Not Getting either a head nor a tail in a single toss of a coin,  $E4 = \emptyset \subset \Omega$

# Example

- Random Experiment : Picking a card of a particular suite from a deck of cards
- Sample Space ( $\Omega$ ) : In picking a card of a particular suite from a deck of cards, the sample space will be :  $\Omega = \{\text{heart, spade, club, diamond}\}$
- For the suite picking case,  $(2^4) = 16$  subsets can be formed out of that sample space.
- Here picking any particular suite of card is equally likely
- Event :
  - Getting a Spade card from a deck of cards,  $E_2 = \{\text{Spade}\} \subset \Omega$
  - Getting a Diamond card from a deck of cards,  $E_2 = \{\text{Diamond}\} \subset \Omega$
  - Getting a red or black card from a deck of cards,  $E_3 \subseteq \Omega$
  - Getting a card with symbol circle from a deck of cards,  $E_4 = \emptyset \subset \Omega$

# Example

- Random Experiment : Rolling a die
- Sample Space ( $\Omega$ ) : In rolling a die, the sample space for the number that shows up on the top face would be :  $\Omega = \{1, 2, 3, 4, 5, 6\}$
- For the die rolling case,  $(2^6) = 64$  subsets can be formed out of that sample space.
- Here getting any number from 1 to 6 is equally likely
- Event :
  - Getting an odd no in a single throw of a die,  $E1 = \{1,3,5\} \subset \Omega$
  - Getting a prime no in a single throw of a die,  $E2 = \{2,3,5\} \subset \Omega$
  - Getting a no less than 7 in a single throw of a die,  $E3 \subseteq \Omega$
  - Getting a no greater than or equal to 7 in a single throw of a die,  $E4 = \emptyset \subset \Omega$

# Example

- Random Experiment : Picking a card from a deck of cards
- Sample Space ( $\Omega$ ) : In picking a card from a deck of cards, the sample space will consist of 52 entries corresponding to each card of the deck.
- For the card picking case,  $(2^{52})$  subsets can be formed out of that sample space.
- Here picking any particular card is equally likely
- Event :
  - Getting a King of Spades from a deck of cards,  $E1 = \{\text{King of Spades}\} \subset \Omega$
  - Getting a Diamond card from a deck of cards,  $E2 = \{K, Q, J, A, 10, 9, 8, 7, 6, 5, 4, 3, 2 \text{ of Diamond}\} \subset \Omega$
  - Getting a red or black card from a deck of cards,  $E3 \subseteq \Omega$
  - Getting a card with number 1 from a deck of cards,  $E4 = \emptyset \subset \Omega$

# CONDITIONAL PROBABILITY

- The probability of occurrence of an event A, given that B has already occurred is called the conditional probability of occurrence of A.
  - It is denoted by  $P(A | B)$ .
  - If the event B has already occurred, then the sample space reduces to B.
  - Now the outcome favorable to the occurrence of A (given that B has already occurred) are those that are common to both A and B, that is, those which belong to  $A \cap B$ .
  - $P(A | B) = P(A \cap B) / P(B)$ ,  $P(B) \neq 0$
- Similarly,  $P(B | A) = P(A \cap B) / P(A)$ ,  $P(A) \neq 0$
- If A and B are two independent events, then  $P(B | A) = P(B)$ ,  $P(A | B) = P(A)$

# A FEW THEOREMS ON PROBABILITY

- If A and B are two events,
  - Addition Rule of Probability
    - $P(A \cup B) = P(A) + P(B) - P(A \cap B)$
    - If A and B are two mutually exclusive events, then  $P(A \cup B) = P(A) + P(B)$ , since  $P(A \cap B) = 0$
  - Multiplication Rule of Probability
    - $P(A \cap B) = P(B | A) * P(A) = P(A | B) * P(B)$
    - If A and B are two independent events, then  $P(A \cap B) = P(A) * P(B)$
- Probability of Complement of an Event,  $P(A') = 1 - P(A)$
- If A, B and C are three events, then
  - $P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(B \cap C) - P(C \cap A) - P(A \cap B) + P(A \cap B \cap C)$
  - $P(E \cap F \cap G) = P(E) * P(F | E) * P(G | (E \cap F)) = P(E) * P(F | E) * P(G | EF)$

# Baye's Theorem

- Theorem of total probability
  - Let  $\{E_1, E_2, \dots, E_n\}$  be a partition of the sample space  $S$ , and suppose that  $E_1, E_2, \dots, E_n$  are mutually exclusive and exhaustive events
  - Let  $A$  be any event associated with  $S$  that occurs with  $E_1, E_2, \dots, E_n$ , then
  - $P(A) = P(E_1) P(A | E_1) + P(E_2) P(A | E_2) + \dots + P(E_n) P(A | E_n) = \sum (P(E_i) P(A | E_i)), i=1, \dots, n$
- Bayes' Theorem
  - Let  $\{E_1, E_2, \dots, E_n\}$  be a partition of the sample space  $S$ , and suppose that  $E_1, E_2, \dots, E_n$  are mutually exclusive and exhaustive events
  - Let  $A$  be any event associated with  $S$  that occurs with  $E_1, E_2, \dots, E_n$ , then
  - $P(E_i | A) = P(E_i) P(A | E_i) / P(A)$

# Application

	P(A is true)	P(A is not true)	
P(Test for A is false)	$P(A \cap TA') - \text{Type I Error } (\alpha)$	$P(A' \cap TA')$	$P(TA')$
P(Test for A is true)	$P(A \cap TA)$	$P(A' \cap TA) - \text{Type II Error } (\beta)$	$P(TA)$
	$P(A)$	$P(A')$	1

Example: Assume A is the Event that you are Covid +ve  
Is a Type I error or Type II error more problematic in this case?



Thank You !!!