

Probability and Statistics

Random Experiments and Sample Spaces:-

↓
Experiments involving
Randomness
Coin toss, dice roll, etc.

→ Ω → Set of all possible outcomes
of a random experiment. It could be
finite or infinite.

$$\Omega_c = \{\text{Head, Tail}\}, \Omega_d = \{1, 2, 3, 4, 5, 6\}$$

$$2\text{-coins} \rightarrow \Omega_{2c} = \Omega_c \times \Omega_c = \{H, T\} \times \{H, T\}$$

Outcomes and Events

↓
(omega) $\rightarrow \omega \in \Omega$ is called a sample point
or a possible outcome

↓
A subset $A \subseteq \Omega$ is called an event.

* Events in the coin toss experiment $C_1 = \{T\} [\subseteq \Omega_c]$
Events in the die roll $D_1 = \{6\}, D_2 = \{1, 3, 5\} [\subseteq \Omega_d]$

Events are any subset of Ω , even null sets, but $P(\emptyset) = 0$

Probability of an event $A = P(A)$.

It may or may not be possible to measure / assign P for every subset A .

$$P_c(T) = 0.5$$

Probability measure P is a set function. It acts on sets and measures the probability of such sets.

Set Theory 101:- A^c = complement of A \emptyset = denotes empty set. [belongs to every set] $A \cup B$ = A union B $A \cap B$ = A intersection B ~~$A \setminus B$~~ $A \setminus B$ = A minus B = $A \cap B^c$ Symmetry $A \Delta B = (A \setminus B) \cup (B \setminus A)$

ME = Mutually Exclusive

~~HW \rightarrow Identity Complement~~ $|A|$ = no. of elements in A = cardinality of A.Inclusion - Exclusion Principle = $|A \cup B| = |A| + |B| - |A \cap B|$

Countable Sets and Uncountable Sets

Monotone Seq $\rightarrow A_1 \subseteq A_2 \subseteq A_3 \subseteq \dots$ [Increasing seq]
 $A_1 \supseteq A_2 \supseteq A_3 \supseteq \dots$ [Decreasing] $I_n = [0, 1 - \frac{1}{n}]$ $D_n = [0, \frac{1}{n}]$ Cartesian product :- $A \times B = \{(a, b) : a \in A, b \in B\}$ Powerset [$\mathcal{P}(A)$] :- Set of all possible subsets of A. $|\mathcal{P}(A)| = 2^{|A|}$ [Only for discrete elements] $\mathcal{P}([0, 1]) = \{(a, b) : a \leq b, a, b \in [0, 1]\}$ \hookrightarrow range, not just 2 elementsFunctions:-

Fns are maps from elements in the domain D to the range R

 $f: D \rightarrow R$ $f: \mathbb{R} \rightarrow \mathbb{R} \Rightarrow f(x) = x$

Set fns are those fns who act on sets; D is a collection

of sets.

IP is a set fn.

Axioms of probability:-

1) $P(\emptyset) = 0$, $P(\Omega) = 1$

2) For a set $A \subseteq \Omega$, $P(A) \in [0, 1]$

3) For a disjoint collection of events A_1, A_2, \dots, A_j where $A_i \subseteq \Omega$, then:

$$P\left(\bigcup_{i=1}^j A_i\right) = \sum_{i=1}^j P(A_i) \quad [j \text{ can be } \infty] \rightarrow \text{but must be countable}$$

where $A_i \cap A_j \forall i, j = \emptyset$

In general, the domain of P is $\mathcal{P}(\Omega)$

~~P. Probability of impossible event = 0.~~

But

Counter example for $\mathcal{P}(\Omega)$ being satisfactory

1) Pick a number randomly (uniformly) of x from the real line.

2) $\Omega = \mathbb{R}$, hence $IP(\mathbb{R}) = 1$.

3) Domain = $\mathcal{P}(\mathbb{R})$

4) $P: \mathcal{P}(\mathbb{R}) \rightarrow [0, 1]$

5) P has the property that sets of equal length have equal probability

6) We know that $\mathbb{R} = \bigcup_{n=-\infty}^{\infty} [n, n+1)$ where $[n, n+1) \in \mathcal{P}(\mathbb{R})$

[Countable union since n is an integer]

7) $IP(\mathbb{R}) = 1 = \sum IP([n, n+1)) = 0 / \infty$ depending

8) This is a contradiction! [Happens when picking segments of unit length]

Powerset is thus a bad choice

Not all set fns can be calibrated to measure all possible subset of the sample space.

Our choice of domain must have certain features.

Let domain = \mathcal{F}

then $\emptyset, \Omega \in \mathcal{F}$; if $B \in \mathcal{F} \rightarrow B^c \in \mathcal{F}$ also;

\uparrow Sets that meet these criteria are called sigma-algebras

Sigma algebra / Event space $[\mathcal{F}]$ associated with a set Ω is a collection of subsets of Ω that satisfy

$\rightarrow \emptyset \in \mathcal{F}$ and $\Omega \in \mathcal{F}$

$\rightarrow A \in \mathcal{F} \rightarrow A^c \in \mathcal{F}$

$\rightarrow A_1, A_2, \dots, A_n \in \mathcal{F} \rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{F}$

\rightarrow The σ -algebra is closed under formation of complements and countable unions and countable intersections (by De Morgan's)

\rightarrow When Ω is countable and finite, then $\mathcal{P}(\Omega)$ is a σ -algebra

If Ω is countable & finite, we will consider $\mathcal{P}(\Omega)$ as the domain.

Probability Space = $\{\Omega, \mathcal{F}, P\}$

Probability measure:- A probability measure P on the measurable space (Ω, \mathcal{F}) is a set fn $P: \mathcal{F} \rightarrow [0, 1]$ s.t.

1) $P(\emptyset) = 0$, $P(\Omega) = 1$

2) For disjoint collection of events A_1, A_2, \dots from \mathcal{F} , we have

$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} P(A_i)$$

P space for $U[0, 1]$

$\Omega = [0, 1]$

If we generate a σ -algebra, but we get a borel-sigma algebra $\mathcal{B}[0, 1]$

Borel σ -algebra $\mathcal{B}[0,1]$

Defined when $\Omega = [0,1]$, as the σ -algebra generated by closed sets of the form $[a,b]$ where $a \leq b$ & $a, b \in [0,1]$

$$(a,b) = \bigcup_{n=1}^{\infty} [a + \frac{1}{n}, b - \frac{1}{n}] \quad \text{--- } \left\{ \left[\overset{a+c}{a} \overset{b-c}{b} \right] \right\}$$

$$(a,b] = \bigcup_{n=1}^{\infty} (a, b + \frac{1}{n}]$$

$\mathcal{B}[0,1]$

Borel σ -algebra \mathcal{B} : is the σ -algebra generated by sets of the form $[a,b]$ or (a,b) or $(a,b]$ or even $[a,b)$ where $a \leq b$ & $a, b \in [0,1]$

If $\Omega = \mathbb{R}$, the $\mathcal{B}(\mathbb{R})$ is the σ -algebra generated by open sets of the form (a,b) where $a < b$ and $a, b \in \mathbb{R}$.

How to define $\mathcal{B}(\mathbb{R}^2)$

Consequences of the Probability Axioms:-

i) $P(A^c) = 1 - P(A)$

$$P(A \cup A^c) = P(A) + P(A^c) = 1$$

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

~~$P(A \cup B)$~~ → Please Prove it

iii) If $A \subseteq B$, prove that $P(A) \leq P(B)$ ($A \subseteq B$ can be interpreted as $A \rightarrow B$)

$$\text{If } A \subseteq B \rightarrow B = A + \lambda$$

iv) $P\left(\bigcup_{i=1}^{\infty} B_i\right) \leq \sum_{i=1}^{\infty} P(B_i)$ [Boole's ineq.]

Diff b/w Impossible event vs Zero probability event

\emptyset \leftarrow

\downarrow
finite sized set

In $U[0,1]$, $P(\omega = 0.5) = 0$

Every experiment outcome of this experiment is a zero probability event. This implies that zero outcome events could occur.

\emptyset on the other hand, can never occur, and are hence impossible events.

$$P(\omega \in [0, 0.25] \cap [0.75, 1]) = 0$$

Limits and Continuity:-

Limits:-

Let a_1, a_2, \dots, a_n be a sequence with limit L .

Then $\forall \epsilon, \exists N_\epsilon$ s.t. $\forall n > N_\epsilon, |a_n - L| \leq \epsilon$

For a fn $f(x)$, $\lim_{x \rightarrow c} f(x) = f(c)$

The limit exists only if $\lim_{x \rightarrow c^+} f(x) = \lim_{x \rightarrow c^-} f(x)$.

All points δ close to c are ϵ close to $f(c)$.

Continuity:-

LHL = RHL = $f(c)$. If the fn is cont at all pts, it is said to be continuous. As $x \rightarrow c$, $f(x) \rightarrow f(c)$

But what about P ?

For a continuous set fn S as $A_n \rightarrow A$, we have $S(A_n) \rightarrow S(A)$

Sequence of sets :-

Given (Ω, \mathcal{F}) , if A_1, A_2, A_3, \dots is an increasing sequence of events defined on \mathcal{F} and $\bigcup_{n=1}^{\infty} A_n = A \in \mathcal{F}$, then we say that the given seq. of sets A_n are increasing to A ($A_n \uparrow A$)

Similarly when $A_1 \supset A_2 \supset \dots$ is a decreasing seq. of events and $\bigcap_{n=1}^{\infty} A_n = A$, then we have $A_n \downarrow A$

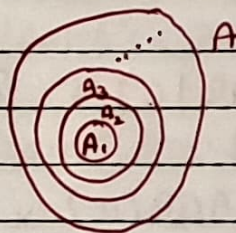
Alt notation :- increasing seq. of sets $A_n \Rightarrow \lim_{n \rightarrow \infty} A_n$ for $\bigcup_{n=1}^{\infty} A_n$

decreasing seq. of sets $A_n \Rightarrow \lim_{n \rightarrow \infty} A_n = \bigcap_{n=1}^{\infty} A_n$

Lemma:- For a sequence of events of the type $A_n \uparrow A$ or $A_n \downarrow A$, we have $\lim_{n \rightarrow \infty} P(A_n) = P(A)$

Proof:-

$$A = \bigcup_{i=1}^{\infty} A_i$$



(i) Consider an increasing seq.

$$\begin{aligned} \text{define } F_n &= A_n - A_{n-1} \\ \bigcup_{n=1}^{\infty} A_n &= \bigcup_{n=1}^{\infty} F_n = \sum_{n=1}^{\infty} F_n = \lim_{n \rightarrow \infty} \sum_{i=1}^n P(F_i) \\ &= \lim_{n \rightarrow \infty} P\left(\bigcup_{i=1}^n F_i\right) = \lim_{n \rightarrow \infty} P(A_n) \end{aligned}$$

Hence Proved.

(ii) Decreasing seq. [H.W.]

Conditional Probability :-

Ex:- Given the die outcome is odd, what is $P(1)$? [$1/3$]
Given $\bar{\omega} \in [0, 0.5]$, what is $P(\bar{\omega} \in [0, 0.25])$

Defn:- The conditional probability of event A is defined as

$$P(B|A) = \frac{P(A \cap B)}{P(A)} \text{ when } P(A) > 0$$

Theorem:- $P(A|B) \cdot P(B) = P(B|A) \cdot P(A)$

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

Bayes' Rule:- $P(B|A) = \frac{P(A|B) \times P(B)}{P(A)}$

(i) What is $P(A|B \cap C)$

(ii) $P(A \cap B) = P(A|B) \times P(B)$

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

$$\downarrow$$

$$\frac{P(A \cap B \cap C)}{P(C)} \times \frac{P(C)}{P(B \cap C)} \times \frac{P(B \cap C)}{P(C)}$$

$$\downarrow$$

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

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

$$\downarrow \qquad \qquad \qquad \uparrow$$

$$\frac{P(A \cap B \cap C)}{P(A \cap B)} \times \frac{P(A \cap B)}{P(A)} \times P(A)$$

iv) $P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1) \times \dots \times P(A_n | A_1 \cap \dots \cap A_{n-1})$

Q) Draw 4 cards without replacement. What is the $P(\text{9 club, 8 dia, K sp, K cl})$

$$P(9_c) \times P(8_d | 9_c) \times P(K_{sp} | 9_c \cap 8_d) \times P(K_{cl} | 9_c \cap 8_d \cap K_{sp})$$

$$= \frac{1}{52} \times \frac{1}{51} \times \frac{1}{50} \times \frac{1}{49} = \frac{1}{52!}$$

When every outcome is equally likely in a finite sample space Ω , $P(A) = \frac{|A|}{|\Omega|}$

Law of Total Probability:-

$$A = (A \cap B) \cup (A \cap B^c), \quad P(A) = P(A \cap B) + P(A \cap B^c)$$

$$\text{Same as } P(A) = P(A|B)P(B) + P(A|B^c)P(B^c)$$

Let B_1, B_2, \dots, B_n be the partition of the sample space Ω . Then for any event A we have

$$P(A) = \sum_{i=1}^n P(A \cap B_i) = \sum_{i=1}^n P(A|B_i)P(B_i)$$

Q) 3 bags with M marbles total. Bag i has R_i red and B_i blue for $i \in [1, 3]$. Find $P(\text{marble} = \text{red from rand bag})$

$$\frac{1}{3} \times \sum_{i=1}^3 \frac{R_i}{R_i + B_i} = \frac{1}{3} \times \sum_{i=1}^3 P(\text{Red} | B_i) \times P(B_i)$$