

Discrete Structures (Monsoon 2021)

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Topic: **Probability Theory**

Definition

If the sample space S of an experiment E consists of finitely many outcomes (points) that are equally likely, the probability of an event A connected with the experiment E is

$$\begin{aligned} P(A) \text{ or } Pr(A) &= \frac{\text{no. of points in } A}{\text{no. of points in } S} \\ &= \frac{m(A)}{n}, \text{ say} \end{aligned}$$

Axioms of Probability

- 1 For any event A in S , $P(A) \geq 0$.
- 2 The probability of a certain event S is $P(S) = 1$.
- 3 For mutually exclusive events A and B , that is $A \cap B = \emptyset$:
 $P(A \cup B) = P(A) + P(B)$.
If S is infinite (has infinitely many points), then for mutually exclusive events A_1, A_2, \dots :

$$P(A_1 \cup A_2 \cup \dots) = P(A_1) + P(A_2) + \dots$$

$$[A_i \cap A_j = \emptyset, i \neq j; i, j = 1, 2, 3, \dots]$$

- **Problem:** If A and B be two events, then show that $P(\bar{A} \cup \bar{B}) = 1 - P(A \cap B)$, where \bar{X} denotes the complement of an event X .
- **Problem (Boole's Inequality):** For any n events A_1, A_2, \dots, A_n , $P(A_1 \cup A_2 \cup \dots \cup A_n) \leq P(A_1) + P(A_2) + \dots + P(A_n)$.
- **Problem:** Find the probability of occurrence of only one of the events A and B .

Conditional Probability

Definition

The conditional probability of an event B on the hypothesis that another event A has occurred will be denoted by $P(B|A)$ and defined by

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

provided that $P(A) \neq 0$.

In a similar way,

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

provided that $P(B) \neq 0$.

Conditional Probability

Theorem

For any two events A and B , $P(A \cap B) = P(A)P(B|A) = P(B)P(A|B)$, if $P(A) \neq 0$ and $P(B) \neq 0$.

Proof.

We have, $P(B|A) = \frac{P(A \cap B)}{P(A)} \Rightarrow P(A \cap B) = P(A)P(B|A)$

Again, $P(A|B) = \frac{P(A \cap B)}{P(B)} \Rightarrow P(A \cap B) = P(B)P(A|B)$.

Thus, $P(A \cap B) = P(A)P(B|A) = P(B)P(A|B)$.



Conditional Probability

Problem: For n events A_1, A_2, \dots, A_n ,

$$P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1)P(A_2|A_1)P(A_3|A_1 \cap A_2) \\ \dots P(A_n|A_1 \cap A_2 \cap \dots \cap A_{n-1})$$

Conditional Probability

Problem: A die is rolled. Let A be the event that the result is an even face and B the event that the result is multiple of 3. Then, compute $P(B|A)$ and $P(A|B)$.

Solution: Here, $n = 6$ points in the sample space $S = \{1, 2, 3, 4, 5, 6\}$.

$A = \{2, 4, 6\}$, $m(A) = 3$ and $P(A) = \frac{m(A)}{n} = \frac{3}{6} = \frac{1}{2}$

$B = \{3, 6\}$, $m(B) = 2$ and $P(B) = \frac{m(B)}{n} = \frac{2}{6} = \frac{1}{3}$

$A \cap B = \{6\}$, $m(A \cap B) = 1$ and $P(A \cap B) = \frac{m(A \cap B)}{n} = \frac{1}{6}$.

Hence, $P(B|A) = \frac{P(A \cap B)}{P(A)} = \frac{1/6}{1/2} = \frac{1}{3}$.

$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{1/6}{1/3} = \frac{1}{2}$.

Conditional Probability

Theorem (Baye's Theorem)

If A_1, A_2, \dots, A_n be a given set of n pairwise mutually exclusive events, one of which certainly occurs, and if X be an arbitrary event such that $P(X) \neq 0$, then

$$P(A_i|X) = \frac{P(A_i)P(X|A_i)}{\sum_{i=1}^n P(A_i)P(X|A_i)}$$

($i = 1, 2, \dots, n$).

Stochastic Independence

Let A and B be events connected with a random experiment E .

The event B is said to be *independent* of A or *stochastically independent* of A , if the probability of B is noway depended on the occurrence of A , that is, if

$$P(B|A) = P(B).$$

We have,

$$\begin{aligned} P(B|A) &= \frac{P(A \cap B)}{P(A)}, P(A) \neq 0 \\ P(A \cap B) &= P(A)P(B|A) \\ &= P(A).P(B). \end{aligned}$$

In other words, two events A and B are said to be independent if

$$P(A \cap B) = P(A).P(B)$$

Random Variable

Definition

Let S be an event space of a random experiment E and R be the set of real numbers. A mapping $X : S \rightarrow R$ is called a random variable or a stochastic variable or simply a variate.

- The range of the mapping X is called the spectrum of X .
- If this spectrum is a discrete set, X is called a *discrete* random variable.
- If this spectrum is a continuous set, X is called a *continuous* random variable.

Distribution Function

Definition

Let X be a random variable defined on S , the event space of a random experiment E , and $x \in R$ be a real number. The distribution function of X is denoted by $F_x(x)$ or $F(x)$ defined in $R = (-\infty, \infty)$ by

$$F(x) = P(X \leq x) = P(-\infty < X \leq x).$$

Note that $F(x)$ is a function of the real variable x .

Properties of Distribution Function $F(x)$

- ① $F(x)$ is a monotonic non-decreasing function of x . That is, if $b > a$, $F(b) \geq F(a)$.
- ② $F(-\infty) = 0$.
- ③ $F(\infty) = 1$.
- ④ $F(a) - F(a - 0) = P(X = a)$, where $F(a - 0)$ denotes the left-handed limit of $F(x)$ at $x = a$.
- ⑤ $F(a + 0) = F(a)$, where $F(a + 0)$ denotes the right-handed limit of $F(x)$ at $x = a$.

Application of Distribution Function $F(x)$

If the distribution function $F(x)$ of a random variable X is given, we can find the probability that X lies in any arbitrary interval $(a, b]$.

In other words, $P(a < X \leq b) = F(b) - F(a)$.

Probability mass function (p.m.f)

- If X is a discrete random variable, its distribution function can be calculated from its probability mass function (p.m.f) f_i defined for all the reals x_i by

$$f_i = P(X = x_i).$$

- If X is a discrete random variable, then

$$\begin{aligned} F(x) &= P(-\infty < X \leq x) = P\left[\sum_{a=-\infty}^i (X = x_a)\right] \\ &= \sum_{a=-\infty}^i P(X = x_a) = \sum_{a=-\infty}^i f_a \end{aligned}$$

$$\text{Again, } F(\infty) = 1 \Rightarrow \sum_{i=-\infty}^{\infty} f_i = 1.$$

Probability density function (p.d.f)

- If X is a continuous random variable, its distribution function can be calculated from its probability density function (p.d.f) $f(\cdot)$ which is characterized by the following properties:

- ▶ (i) $f(x) \geq 0$ for all real x
- ▶ (ii) $\int_{-\infty}^{\infty} f(x)dx = 1$
- ▶ $P(a \leq X \leq b) = \int_a^b f(x)dx$, for all reals a, b with $a < b$

where $f(x) = F'(x) = \frac{d}{dx} F(x)$.

- Then, $dF(x) = f(x)dx \Rightarrow F(x) = \int_{-\infty}^x f(t)dt = P(-\infty < X \leq x)$.

Probability Differential

- We have,

$$\begin{aligned}P(x < X \leq x + dx) &= F(x + dx) - F(x) \\&= dF(x) \\&= \frac{d}{dx}F(x).dx \\&= f(x)dx.\end{aligned}$$

- $P(x < X \leq x + dx) = f(x)dx$.
- $dF(x)$ is called the probability differential.