

Chapter One

Random Variables and Probability Distributions

Definition:

Let (Ω, \mathcal{A}, P) be a probability measure space and let $X: \Omega \rightarrow \mathbb{R}$ be a function from Ω into the set \mathbb{R} of real numbers such that the pre-image of every interval of \mathbb{R} is an event of Ω . Then X is a random variable.

That is $X: \Omega \rightarrow \mathbb{R}$ is a random variable if:

$$X^{-1}((-\infty, x]) = \{\omega \in \Omega : X(\omega) \in (-\infty, x]\} \in \mathcal{A}, \forall x \in \mathbb{R}.$$

i.e., the inverse X^{-1} maps all of the sets $(-\infty, x] \subset \mathbb{R}$ to \mathcal{A} .

Remark: In general, if $\mathcal{A} = \mathcal{P}(\Omega)$, any function $X: \Omega \rightarrow \mathbb{R}$ is a random variable.

Example:

Consider the random experiment of flipping a fair coin twice. The sample space is $\Omega = \{HH, HT, TH, TT\}$. Since Ω is finite we take $\mathcal{A} = \mathcal{P}(\Omega)$. Suppose that we are primarily interested in the number of tails that have occurred, i.e, suppose that our interest is on the function $X: \Omega \rightarrow \mathbb{R}$ defined by:

$$X(\omega) = \begin{cases} 0 & \text{if } \omega = \{HH\} \\ 1 & \text{if } \omega = \{HT, TH\} \\ 2 & \text{if } \omega = \{TT\} \end{cases}$$

Verify that X is a random variable.

$\forall x \in \mathbb{R}$

$$X^{-1}((-\infty, x]) = \begin{cases} \emptyset \in \mathcal{A} & x < 0 \\ \{HH\} \in \mathcal{A} & 0 \leq x < 1 \\ \{HH, HT, TH\} \in \mathcal{A} & 1 \leq x < 2 \\ \Omega \in \mathcal{A} & 2 \leq x \end{cases}$$

It is clear that for each outcome ω of Ω , there corresponds a real number $X(\omega)$. Hence, X is a random variable.

Sometimes it is easier to work with the following equivalent condition:

$$\forall x \in \mathbb{R} \implies \{\omega \in \Omega : X(\omega) \leq x\} \in \mathcal{A}$$

This means that once we know the value $X(\omega)$, we know which of the events in \mathcal{A} have happened.

Note:

- If X is a random variable then $aX+b$ will also be a random variable where a, b are constants. Also X^2 and $\frac{1}{X}$ will be random variables.
- Sums and products of random variables are also random variables.

Discrete and Continuous Random Variables:

Def. A random variable is called discrete if its range includes finite number of values or countable infinite number of values x_1, x_2, \dots

Example: The number of members of a family, number of passengers on a plane, number of road accidents occurs in a day in a city ...etc.

Def. A random variable is called continuous if it can assume any value from a specified interval of the form $[a, b]$. That is; if for some $a < b$ any number x between a and b is possible.

Example: Heights of students in a class, temperature and barometric pressures of different cities, age of people or objects ... etc.

Distribution Function:

With every random variable is associated another function known as its distribution function.

Def. The cumulative distribution function of a random variable X is a function $F_x: \mathbb{R} \rightarrow [0, 1]$ which satisfies

$$F_X(x) = P[-\infty < X \leq x] = P[X \leq x] = P[\{\omega: X(\omega) \leq x\}], -\infty < x < \infty$$

Example:

Find the distribution function of the random variable X in the previous example.

$$F_X(x) = P(X^{-1}(-\infty, x]) = P[-\infty < X \leq x] = P[X \leq x]$$

$$= \begin{cases} P(\emptyset) = 0 & x < 0 \\ P(HH) = \frac{1}{4} & 0 \leq x < 1 \\ P(\{HH, HT, TH\}) = \frac{3}{4} & 1 \leq x < 2 \\ P(\Omega) = 1 & 2 \leq x \end{cases}$$

Lemma:

If F_x is the distribution function of the random variable X and if $a < b$, then

$$1. P(a < X \leq b) = F_X(b) - F_X(a)$$

Proof:

$$(-\infty, b] = (-\infty, a] \cup (a, b]$$

$$P(-\infty < X \leq b) = P(-\infty < X \leq a) + P(a < X \leq b)$$

$$\begin{aligned} P(a < X \leq b) &= P(-\infty < X \leq b) - P(-\infty < X \leq a) \\ &= F_X(b) - F_X(a) \end{aligned}$$

$$2. P(a < X < b) = F_X(b) - F_X(a) - P(X = b)$$

Proof:

$$(-\infty, b] = (-\infty, a] \cup (a, b) \cup \{b\}$$

$$P(-\infty < X \leq b) = P(-\infty < X \leq a) + P(a < X < b) + P(X = b)$$

$$P(a < X < b) = P(-\infty < X \leq b) - P(-\infty < X \leq a) - P(X = b)$$

$$P(a < X < b) = F_X(b) - F_X(a) - P(X = b)$$

$$3. P(a \leq X \leq b) = F_X(b) - F_X(a) + P(X = a)$$

Proof:

$$(-\infty, b] = (-\infty, a] \cup [a, b]$$

$$P(-\infty < X \leq b) = P(-\infty < X \leq a) + P(a \leq X \leq b) - P(X = a)$$

$$\begin{aligned} P(a \leq X \leq b) &= P(-\infty < X \leq b) - P(-\infty < X \leq a) + P(X = a) \\ &= F_X(b) - F_X(a) + P(X = a) \end{aligned}$$

$$4. P(a \leq X < b) = F_X(b) - F_X(a) - P(X = b) + p(X = a)$$

Proof:

$$(-\infty, b] = (-\infty, a] \cup [a, b) \cup \{b\}$$

$$P(-\infty < X \leq b)$$

$$= P(-\infty < X \leq a) + P(a \leq X < b) - P(X = a) + P(X = b)$$

$$P(a \leq X < b)$$

$$\begin{aligned} &= P(-\infty < X \leq b) - P(-\infty < X \leq a) + P(X = a) - P(X = b) \\ &= F_X(b) - F_X(a) + P(X = a) - P(X = b) \end{aligned}$$