

# Lecture Notes 4: Probability and Statistics

So far we have talked about random experiment, sample space,  $\sigma$ -field and probability function. Now we are going to introduce an interesting and important concept in probability theory, and it is known as random variables. Suppose, we have a random experiment, and based on the random experiment we have a sample space  $\Omega$ . Now a random variable  $X$  is a real valued function defined on  $\Omega$ . Note that  $\Omega$  can have finite, countable or uncountable number of elements. Let us look at some example.

**Example:** Suppose we throw a dice twice and observe the faces which appear on the top. Clearly in this case the sample space  $\Omega$  has 36 points, namely

$$\Omega = \{(i, j); i, j = 1, 2, \dots, 6\}.$$

Suppose we define a function  $X$  on  $\Omega$  as follows:  $X(i, j) = i + j$ . Clearly,  $X$  is a random variable and it can take values 2, 3, ..., 12. If it is assumed that probability of appearing any face on the top is equally likely and the two dice are independently thrown, then

$$\begin{aligned} P(X = 2) &= P(X = 12) = \frac{1}{36}, & P(X = 3) = P(X = 11) = \frac{2}{36}, \\ P(X = 4) &= P(X = 10) = \frac{3}{36}, & P(X = 5) = P(X = 9) = \frac{4}{36}, \\ P(X = 6) &= P(X = 8) = \frac{5}{36}, & P(X = 7) = \frac{6}{36}. \end{aligned}$$

**Example:** Now let us look at the continuous case when the random variable  $X$  can take continuous values. Suppose we choose a number at random from  $[0, 1]$ , and let us denote that number as  $X$ . In this case  $\Omega = [0, 1]$ , and  $X$  is a random variable taking any values in  $[0, 1]$ . Since, it is assumed that the number is chosen at random, it means for any  $a, b \in [0, 1]$ ,

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

A discrete random variable will be always characterized by the probability mass function (PMF). For a continuous random variable  $X$  if there exists a

function  $f(x)$ , such that  $f(x) \geq 0$ , and for all  $-\infty < a < b < \infty$ ,

$$P(a < X < b) = \int_a^b f(x)dx,$$

then  $f(x)$  is called the probability density function (PDF) of  $X$ . It is immediate that

$$\int_{-\infty}^{\infty} f(x)dx = P(-\infty < X < \infty) = 1.$$

From now on it is assumed that if  $f(x)$  is a function satisfy

$$f(x) \geq 0, \quad \text{for all } x \quad \text{and} \quad \int_{-\infty}^{\infty} f(x)dx = 1,$$

then there exists a random variable  $X$ , such that for all  $-\infty < a < b < \infty$ ,

$$P(a < X < b) = \int_a^b f(x)dx,$$

and  $f(x)$  is the PDF of  $X$ .

Now we will introduce another important function, and it is known as the cumulative distribution function (CDF) or a distribution function (DF) of a random variable  $X$ . It is defined as

$$F_X(x) = P(X \leq x) = P(-\infty < X \leq x); \quad \text{for all } x.$$

Note that the CDF can be defined both for the discrete and continuous random variables. If  $X$  is a discrete random variable with the following PMF

$$P(X = a_i) = p_i; \quad i = 1, 2, 3 \dots,$$

here  $a_i$ 's are arbitrary real numbers, and  $p_i$ 's satisfy the following continuous;

$$p_i \geq 0, \quad \text{for all } i, \quad \text{and} \quad \sum_{i=1}^{\infty} p_i = 1.$$

Then the CDF of  $X$  becomes

$$F_X(x) = \sum_{i:a_i \leq x} p_i.$$

It is clear that the distribution function of a discrete random variable is a step function, and it has jumps at  $a_1, a_2, \dots$

**Example:** Suppose  $X$  has the following PMF

$$P(X = -1) = \frac{1}{4}, \quad P(X = 0) = \frac{1}{2} \quad P(X = 1) = \frac{1}{4}.$$

Then the CDF of  $X$  becomes

$$F_X(x) = \begin{cases} 0 & \text{if } x < -1 \\ \frac{1}{4} & \text{if } -1 \leq x < 0 \\ \frac{3}{4} & \text{if } 0 \leq x < 1 \\ 1 & \text{if } x \geq 1 \end{cases}$$

It can be seen that  $F_X(x)$  is a step function, and it has jumps at  $-1, 0, 1$ .

If  $X$  is a continuous random variable with PDF  $f(x)$ , then the CDF of  $X$  becomes

$$F_X(x) = \int_{-\infty}^x f(u)du.$$

It is immediate that  $F_X(x)$  is everywhere continuous.

**Example:** Suppose  $X$  is a continuous random variable with PDF

$$f(x) = e^{-2|x|}; \quad -\infty < x < \infty,$$

then the CDF of  $X$  becomes

$$\begin{aligned} F_X(x) &= \int_{-\infty}^x e^{-2|u|}du = \begin{cases} \int_{-\infty}^x e^{2u}du & \text{if } x < 0 \\ \frac{1}{2} + \int_0^x e^{-2u}du & \text{if } x \geq 0 \end{cases} \\ &= \begin{cases} \frac{1}{2}e^{2x} & \text{if } x < 0 \\ \frac{1}{2} + \frac{1}{2}(1 - e^{-2x}) & \text{if } x \geq 0 \end{cases} \end{aligned}$$

Now let us look at some of the properties of a CDF. If  $F(x)$  is a distribution function of a random variable, then it is clear

1.  $0 \leq F(x) \leq 1$ , for all  $x$ .
2.  $\lim_{x \rightarrow -\infty} F(x) = 0$  and  $\lim_{x \rightarrow \infty} F(x) = 1$ .

Now we will prove that  $F_X(x)$  is right continuous for all  $x$ . Let  $a_n$  be a sequence of real numbers and  $a_n \downarrow a_0$ . Then let us consider the sets  $A_n = \{X \in (-\infty, a_n]\}$ , for  $n = 1, 2, \dots$ . Note that  $\bigcap_{n=1}^{\infty} A_n = A_0 = \{X \in (-\infty, a_0]\}$ .

Further,

$$\lim_{n \rightarrow \infty} F(a_n) = \lim_{n \rightarrow \infty} P(A_n) = P\left(\bigcap_{n=1}^{\infty} A_n\right) = P(A_0) = F(a_0).$$

On the other hand let  $b_n$  be a sequence such that  $b_n \uparrow a_0$ , and if  $B_n = \{X \in (-\infty, b_n]\}$ , then  $\bigcup_{n=1}^{\infty} B_n = B_0 = \{X \in (-\infty, a_0)\}$ . Hence,

$$\lim_{n \rightarrow \infty} F(b_n) = \lim_{n \rightarrow \infty} P(B_n) = P\left(\bigcup_{n=1}^{\infty} B_n\right) = P(B_0) = F(a_0-).$$

Therefore,

$$P(X = a_0) = P(-\infty < X \leq a_0) - P(-\infty < X < a_0) = F(a_0) - F(a_0-).$$

From now on if a function  $F(x)$  defined on the whole real line and satisfy the following properties

1.  $0 \leq F(x) \leq 1$ , for all  $x$ .

2.  $\lim_{x \rightarrow -\infty} F(x) = 0$  and  $\lim_{x \rightarrow \infty} F(x) = 1$ .

3.  $F(x)$  is right continuous,

will be called a distribution function.

It easily follows from the definition of a distribution function:

$$P(a < X \leq b) = F(b) - F(a), \quad P(a \leq X \leq b) = F(b) - F(a-)$$

$$P(a \leq X < b) = F(b-) - F(a-), \quad P(a < X < b) = F(b-) - F(a).$$

**Example** Now consider the following function  $F(x)$

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 - \frac{1}{2}e^{-x} & \text{if } 0 \leq x < 2 \\ 1 & \text{if } x \geq 2 \end{cases}$$

It is immediate that  $F(x)$  satisfies all the above properties, hence, it is a proper CDF. It can be seen that it has jumps at 0 and 2. Here

$$P(X = 0) = F(0) - F(0-) = \frac{1}{2}, \quad P(X = 2) = F(2) - F(2-) = \frac{1}{2}e^{-2}.$$

It is clear that  $F(x)$  is not a step function it is also not a continuous function. This kind of CDF is called a mixture distribution. We would like to write  $F(x)$  in the following form

$$F(x) = \alpha F_c(x) + (1 - \alpha)F_d(x).$$

Here  $F_c(x)$  is a proper continuous CDF,  $F_d(x)$  is a proper discrete distribution CDF and  $\alpha$  is the mixing proportion. Let  $g(x) = \frac{d}{dx}F(x)$ . Therefore,

$$g(x) = \begin{cases} 0 & \text{if } x < 0 \\ \frac{1}{2}e^{-x} & \text{if } 0 \leq x < 2 \\ 0 & \text{if } x \geq 2 \end{cases}$$

Therefore,  $\alpha F_c(x) = \int_{-\infty}^x g(u)du$ , and

$$\alpha F_c(x) = \begin{cases} 0 & \text{if } x < 0 \\ \frac{1}{2}(1 - e^{-x}) & \text{if } 0 \leq x < 2 \\ \frac{1}{2}(1 - e^{-2}) & \text{if } x \geq 2. \end{cases}$$

Hence,  $\alpha = \frac{1}{2}(1 - e^{-2})$ ,  $(1 - \alpha) = \frac{1}{2}(1 + e^{-2})$  and

$$F_c(x) = \begin{cases} 0 & \text{if } x < 0 \\ \frac{1-e^{-x}}{1-e^{-2}} & \text{if } 0 \leq x < 2 \\ 1 & \text{if } x \geq 2. \end{cases}$$

Therefore,  $(1 - \alpha)F_d(x) = F(x) - \alpha F_c(x)$ , and

$$(1 - \alpha)F_d(x) = \begin{cases} 0 & \text{if } x < 0 \\ \frac{1}{2} & \text{if } 0 \leq x < 2 \\ \frac{1}{2}(1 + e^{-2}) & \text{if } x \geq 2. \end{cases}$$

and

$$F_d(x) = \begin{cases} 0 & \text{if } x < 0 \\ \frac{1}{1+e^{-2}} & \text{if } 0 \leq x < 2 \\ 1 & \text{if } x \geq 2. \end{cases}$$