

# Stat 113 (Introduction to Mathematical Statistics)

## Lesson 2.1: Discrete Random Variables and Their Probability Distributions

### Learning Outcomes

At the end of the lesson, students must be able to

1. Explain what is a random variable,
2. Differentiate discrete and continuous random variables, and
3. Construct the probability mass function and cumulative distribution function of a discrete random variable.

### Introduction

In practice it is convenient to represent random outcomes numerically. If the outcome is numerical and one-dimensional we call the outcome a random variable. If the outcome is multi-dimensional we call it a random vector. Random variables are one of the most important and core concepts in probability and statistics. It provides a way to quantify the outcomes of random experiments.

As an example consider a coin flip which has the possible outcome H or T. We can write the outcome numerically by setting the result as  $X = 1$  if the coin is heads and  $X = 0$  if the coin is tails. The object  $X$  is random since its value depends on the outcome of the coin flip.

#### Definition:

A *random variable* is a function which assigns a numerical value to every outcome of a random experiment.

#### Example 2.1:

Consider tossing 2 coins. The possible outcomes are

$$\Omega = \{HH, HT, TH, TT\}$$

Let us count the number of heads and label it as  $Y$ . Then

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

Remarks:

1. In the introductory example, the values of  $X$  are random since being equal to 0 or 1 depends on the random outcomes in tossing a coin. Hence,  $X$  is a random variable. The same is true in Example 2.1
2. By convention we use uppercase letters to denote random variables, say  $X$ ,  $Y$ , or  $Z$ , and the corresponding lowercase letters for the realized value.

Example 2.2:

Consider rolling a pair of dice. Let  $Z$  be the sum of the number of dots on each die. What are the possible values of  $Z$ ?

### Types of random variables

A random variable which assumes finite or countably infinite number of possible values is called a *discrete* random variable.

Random variables  $X$ ,  $Y$ , and  $Z$  are examples of discrete random variables. Other examples are

1.  $X$  = number of vehicles owned by a random sample of politicians
2.  $Y$  = number of houses affected by a typhoon
3.  $Z$  = number of mobile phones owned by a random sample of students

In the above examples, it is clear that a discrete random variable assumes values corresponding to the natural numbers or counting numbers, otherwise called the whole numbers.

Meanwhile, a random variable which assumes values in an interval of values is referred to as *continuous* random variable. Examples are

1.  $X$  = height of a random tree in a forest reserve
2.  $Y$  = salary of a randomly selected VSU employee
3.  $Z$  = length of time a random student finishes an exam

## Probability distribution of a discrete random variable

Every random variable has an associated *probability distribution*, or simply *distribution*. The distribution of the random variable, say  $X$ , refers to the assignment of probabilities to all events defined in terms of this random variable.

### Definition:

Suppose that  $X$  is a discrete random variable. The function  $p_X(x) = P(X = x)$  is called the *probability mass function* (pmf) for  $X$ .

Remarks:

1. The pmf of  $X$  has two parts:
  - a. the domain of  $X$ , and
  - b. the probability assignment  $P(X = x)$  for every  $x \in \mathbb{R}$
2. For any discrete probability distribution, the following must be true:
  - a.  $p_X(x) = P(X = x) \geq 0$ , and
  - b.  $\sum p_X(x) = 1, \forall x \in \mathbb{R}$
3. The pmf of a discrete random variable can be presented as a table, a graph, or a formula.
4. Given the pmf of a discrete random variable we can derive its cumulative distribution function (CDF), and vice versa.

### Definition:

The cumulative distribution function (CDF) of a random variable, say  $X$ , is given by  $F_X(x) = P(X \leq x)$ .

### Example 2.3:

In Example 2.1 we defined  $Y$  as the number of heads recorded in tossing two coins and the possible values of  $Y$  are 0, 1, or 2. Then the pmf of  $Y$  is given by

$Y$	0	1	2
$P(Y = y)$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$

while the CDF of Y is given by

$$F_Y(y) = \begin{cases} 0, & \text{if } y < 0 \\ \frac{1}{4}, & \text{if } 0 \leq y < 1 \\ \frac{3}{4}, & \text{if } 1 \leq y < 2 \\ 1, & \text{if } y \geq 2 \end{cases}$$

QUESTIONS:

1. How did I come up with this pmf?
2. How did I come up with the CDF?

Example 2.4:

Consider the random experiment of tossing 3 coins. Define X as the number of heads recorded in this experiment.

- a. Derive the pmf of X.
- b. Derive the CDF of X.