

CP Statistics – Chapter 6: Discrete Random Variables

6.1: Discrete and Continuous Random Variables

RANDOM VARIABLE

A **random variable** is a variable whose value is a numerical outcome of a random phenomenon.

DISCRETE RANDOM VARIABLE

A **discrete random variable** X has a *countable* number of possible values. Generally, these values are limited to integers (whole numbers). The probability distribution of X lists the values and their probabilities.

Value of X	x₁	x₂	x₃	...	x_k
Probability	p₁	p₂	p₃	...	p_k

The probabilities p_i must satisfy two requirements:

1. Every probability p_i is a number between 0 and 1.
2. $p_1 + p_2 + \dots + p_k = 1$

Find the probability of any event by adding the probabilities p_i of the particular values x_i that make up the event.

CONTINUOUS RANDOM VARIABLE

A **continuous random variable** X takes all values in an interval of numbers and is *measurable*.

For example: a person's height in inches can take on any value and is not limited to a whole number of inches.

MEAN OF A DISCRETE RANDOM VARIABLE

Suppose that X is a discrete random variable whose distribution is

Value of X	x₁	x₂	x₃	...	x_k
Probability	p₁	p₂	p₃	...	p_k

To find the **mean** of X , multiply each possible value by its probability, then add all the products:

$$\mu_x = E(x) = \sum x_i \cdot p_i = x_1 \cdot p_1 + x_2 \cdot p_2 + \dots + x_k \cdot p_k$$

For example, for the distribution below:

X	1	2	3	4	5
P(X)	0.1	0.2	0.3	0.3	.1

The mean would be computed as

$$\mu_x = 1(.1) + 2(.2) + 3(.3) + 4(.3) + 5(.1) = 3.1$$

LAW OF LARGE NUMBERS

Draw independent observations at random from any population with finite mean μ . As the number of observations drawn increases, the mean of the observed values eventually approaches the mean μ .

For example, if a coin has probability .5 of being flipped tails, the proportion of tails will be very close to .5 *in the long run*. We cannot say what will happen in the short term, but after *many, many* flips, this proportion will be approximately .5.

6.3 – Binomial Random Variables

A **binomial probability distribution** occurs when the following requirements are met.

- B** – Each outcome is *binary*, having only two possibilities – call them “success” or “failure.”
- I** – The observations must be *independent* – result of one does not affect another.
- N** – The procedure has a fixed *number* of trials – we call this value n .
- S** – The probability of *success* – call it p - remains the same for each observation.

Notation for binomial probability distribution

- n = the number of fixed trials
- k = the number of successes in the n trials
- p = the probability of success
- $1 - p$ = the probability of failure

Binomial Probability Formula

$$P(X = k) = \frac{n!}{k!(n-k)!} (p)^k (1-p)^{n-k}$$

Mean (expected value) of a Binomial Random Variable

Formula: $\mu = np$ Meaning: Expected number of successes in n trials (think *average*)

Example: Suppose you are a 80% free throw shooter. You are going to shoot 4 free throws.

For $n = 4$, $p = .8$, $\mu = (4)(.8) = 3.2$, which means we expect 3.2 makes out of 4 shots, on average

6.3 – Geometric Random Variables

A **geometric probability distribution** occurs when the following requirements are met.

- B** – Each outcome is *binary*, having only two possibilities – call them “success” or “failure.”
- I** – The observations must be *independent* – result of one does not affect another.
- T** – The variable of interest is the number of *trials* required to obtain the first success.*
- S** – The probability of *success* – call it p - remains the same for each observation.

* As such, the geometric setting is also called a “waiting-time” distribution

Notation for geometric probability distribution

- n = the number of trials required to obtain the first success
- p = the probability of success
- $1 - p$ = the probability of failure

Geometric Probability Formula

$$P(X = n) = (1-p)^{n-1} (p)$$

Mean (expected value) of a Geometric Random Variable

Formula: $\mu = \frac{1}{p}$ Meaning: Expected number of n trials to achieve first success (*average*)

Example: Suppose you are a 80% free throw shooter. You are going to shoot until you make.

For $p = .8$, $\mu = \frac{1}{.8} = 1.25$, which means we expect to take 1.25 shots, on average, to make first