



# Chapter 4

## Random Sample

### Central Limit Theorem



## Learning Outcomes:

At the end of the lesson student should be able to

- Describe the terms random sample, statistics and sampling distribution.
- Understand the concept of central limit theorem.
- Use central limit theorem to approximate sampling distribution of the sample mean.



## DEFINITIONS

- **Parameters**: a quantity of interest that is a property of an unknown probability distribution. Example it may be the mean or variance of the probability distribution.
- Parameters are unknown but could be estimated.

Example:

Let  $p$  be the probability that a machine breakdown is due to operator misuse.

$p$  depends on the probability distribution that represents the causes of the machine breakdown.

$p$  is a parameter and unknown but can be estimated from records of machine breakdown.

## DEFINITIONS

- A **statistic** is a property of a sample from the population. A function of a set of data observations. Example are the sample mean or sample variance.
- Statistics can be calculated from a set of observed data.
- Statistics can be used to estimate unknown parameters.

Example:

$$\bar{X} = f(X_1, \dots, X_n) = \frac{1}{n} \sum_{i=1}^n X_i$$

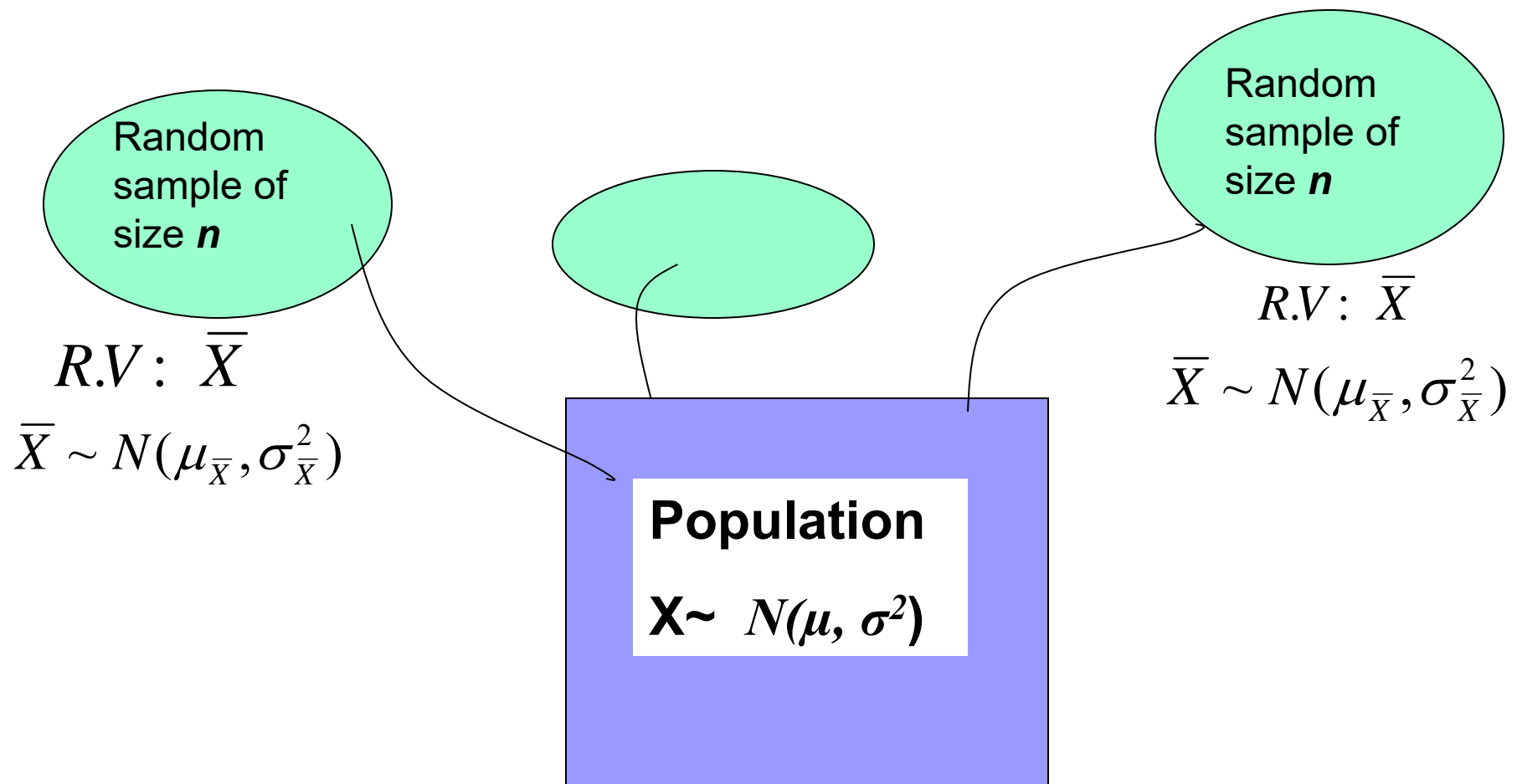
$$g(X_1, \dots, X_n) = \sum_{i=1}^n X_i^2$$

## DEFINITIONS

- **Random Sample**: Independent random variables  $X_1, X_2, \dots, X_n$  with the same distribution and each has the same chance of being selected.
- **Sampling Distribution**: the probability distribution of a statistic.

Example:  $\bar{X} \sim N(\mu_{\bar{X}}, \sigma_{\bar{X}}^2)$

# Random Sample; Sampling Distribution



# Distribution of Sample Mean

- Let  $X_1, X_2, \dots, X_n$  be a random sample from a population with mean  $\mu$  and variance  $\sigma^2$
- Definition and Properties (sampling distribution of  $\bar{X}$ ):

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$E(\bar{X}) = \mu, \quad V(\bar{X}) = \frac{\sigma^2}{n}$$

# Central Limit Theorem

- Let  $X_1, X_2, \dots, X_n$  be a random sample from a population with mean  $\mu$  and variance  $\sigma^2$

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$E(\bar{X}) = \mu, \quad V(\bar{X}) = \frac{\sigma^2}{n}$$

- If  $n$  is sufficiently large

$$\bar{X} \approx N\left(\mu, \frac{\sigma^2}{n}\right)$$

**Apply safely**  
**if  $n \geq 30$**

$$Z = \frac{\bar{X} - \mu}{\sigma / \sqrt{n}} \sim \mathcal{N}(0,1),$$



## Example 1

At chemical engineering department, UTP, the mean age of the students is 20.6 years old, and the variance is 20 years. A random sample of 80 students is drawn from 250 students. What is the probability that the average age of these students is greater than 22 years old?

## Solution

## Example 1

At chemical engineering department, UTP, the mean age of the students is 20.6 years old, and the variance is 20 years. A random sample of 80 students is drawn from 250 students. What is the probability that the average age of these students is greater than 22 years old?

### Solution

The mean of  $X = E(X) = 20.6$  and the variance of  $X = V(X) = 20$

For  $n = 80$ , the mean of  $\bar{X} = E(\bar{X}) = 20.6$  and  $V(\bar{X}) = \frac{\sigma^2}{n} = \frac{20}{80} = 0.25$

Hence,  $\bar{X} \sim N(20.6, 0.25)$

$$\begin{aligned}\text{So, } P(\bar{X} > 22) &= P\left(Z > \frac{22 - 20.6}{\sqrt{0.25}}\right) = P(Z > 2.8) = 1 - P(Z \leq 2.8) \\ &= 1 - \Phi(2.8) = 1 - 0.9974 = 0.0026\end{aligned}$$

**Example 2:** Let  $X$  denote the number of flaws in a 1 in length of copper wire. The pmf of  $X$  is given in the following table

$X = x$	0	1	2	3
$P(X = x)$	0.48	0.39	0.12	0.01

100 wires are sampled from this population. What is the probability that the average number of flaws per wire in this sample is less than 0.5?

**Solution:**  $\mu = 0.66, \sigma^2 = 0.5244, n = 100$

Let  $X_1, \dots, X_{100}$  denote the number of flaws in the 100 wires sampled from the population.

From CLT  $\bar{X} \approx N(0.66, 0.005244)$

$$\bar{X} \approx N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$\begin{aligned}
 P(\bar{X} < 0.5) &= P\left(Z < \frac{0.5 - 0.66}{0.7242 / \sqrt{100}}\right) \\
 &= P(Z < -2.21) = 0.0136
 \end{aligned}$$

















### Example 3:

The flexural strength (in MPa) of certain concrete beams is  $X \sim N(8, 2.25)$ . Find the probability that the sample mean of strength of 36 concrete beams will belong to  $(7.55, 8.75)$

### Solution:

$$\mu = 8, \quad \sigma^2 = 2.25$$

$$n = 36, \quad \sigma = 1.5$$

$$\begin{aligned} P(7.55 < \bar{X} < 8.75) &= P\left(\frac{7.55 - 8}{1.5/6} < \frac{\bar{X} - 8}{1.5/6} < \frac{8.75 - 8}{1.5/6}\right) \\ &= P\left(-1.8 < \frac{\bar{X} - 8}{1.5/6} < 3\right) = P(-1.8 < Z < 3) \\ &= \Phi(3) - \Phi(-1.8) = 0.9987 - 0.0359 = 0.9628 \end{aligned}$$





### Exercise:

At a large university, the mean age of the students is 22.3 years, and the standard deviation is 4 years. A random sample of 64 students is drawn. What is the probability that the average age of these students is greater than 23 years?













## Normal Approximation

The binomial and Poisson distributions are discrete random variables, whereas the normal distribution is continuous.

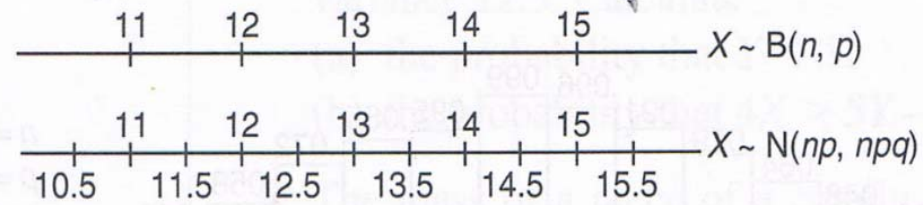
The Normal approximation can be used to find the probability of  $X$  for Binomial or Poisson by add the continuity correction value 0.5.

The continuity correction,  $\pm 0.5$  for probability of  $X$  is depend on the inequality sign,  $<$ ,  $\leq$ ,  $>$ ,  $\geq$

For example  **$P(X < a) = P(X - 0.5 < a - 0.5)$**  and for

$$P(X \leq a) = P(X + 0.5 \leq a + 0.5)$$





▲ Figure 66

Hence, we can approximate the probabilities as follows:

**Discrete  $B(n, p)$**

$$P(X = 12)$$

$$P(X < 12)$$

$$P(X \leq 12)$$

$$P(X > 12)$$

$$P(X \geq 12)$$

$$P(11 \leq X \leq 14)$$

$$P(11 < X < 14)$$

$$P(11 < X \leq 14)$$

$$P(11 \leq X < 14)$$

**Continuous  $N(np, npq)$**

$$P(11.5 < X < 12.5)$$

$$P(X < 11.5)$$

$$P(X < 12.5)$$

$$P(X > 12.5)$$

$$P(X > 11.5)$$

$$P(10.5 < X < 14.5)$$

$$P(11.5 < X < 13.5)$$

$$P(11.5 < X < 14.5)$$

$$P(10.5 < X < 13.5)$$



## Normal approximation to Binomial

A random variable  $X \sim \text{Bin}(n, p)$  and  $n$  is large and  $p$  is small such that  $np > 5$ , then  $X$  can be calculated approximately using the Normal distribution.

It means that the random variable  $X$  will be normally distributed with mean  $\mu = np$  and variance,  $\sigma^2 = np(1 - p)$  so,  $X \sim N(np, npq)$ .

### Example :

Suppose in experiment of tossing a fair coin for 200 times. What is the probability of getting between 90 and 110 heads?

### Solution:

Let  $X$  be the random variable representing the number of heads thrown.

$$X \sim \text{Bin}(200, 0.5)$$

Since  $n$  is large and  $np > 5$ , then we can use normal approximation to find the probability. It means that now,  $X$  is normally distributed with mean  $np = 100$  and variance  $50$ .



So,  $X \sim N(100, 50)$ .

Hence,

$$\begin{aligned} P(90 \leq X \leq 110) &= P\left(\frac{(90 - 0.5) - 100}{\sqrt{50}} \leq Z \leq \frac{(110 + 0.5) - 100}{\sqrt{50}}\right) \\ &= P\left(\frac{89.5 - 100}{7.07} \leq Z \leq \frac{110.5 - 100}{7.07}\right) = P(-1.48 \leq Z \leq 1.48) \\ &= \Phi(1.48) - \Phi(-1.48) = 0.9306 - 0.0694 = 0.8612 \end{aligned}$$



## Normal approximation to Poisson

The normal distribution can also be used to approximate the Poisson distribution for large values of  $\lambda$  (the mean of the Poisson distribution).

If we have a random variable  $X \sim \text{Poisson}(\lambda)$  and  $\lambda$  is large then  $X$  can be calculated approximately using the Normal distribution. It means that the random variable  $X$  will be normally distributed with mean  $\mu = \lambda$  and variance,  $\sigma^2 = \lambda$  so,  $X \sim N(\lambda, \lambda)$

### Example:

A car hire firm has 20 cars to hire. The number of demands for a car is hired per day is a Poisson distribution with mean of 5. Calculate the probability that at most ten cars will be hired in one day.



### Solution:

Let a random variable ***X*** denotes the number of demands for a car.  
The given mean value is 5. By the Poisson distribution

$$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}, \quad x = 0, 1, 2, 3, 4, \dots, 20$$

Since  $\lambda$  is large, then the probability can be calculated using a normal approximation with mean  $\lambda=5$  and variance is also  $\lambda=5$ .  
so, ***X***  $\sim$  ***N***(5, 5).

Hence,

$$\begin{aligned} P(X \leq 10) &= P\left(\frac{(X + 0.5) - 5}{\sqrt{5}} \leq \frac{(10 + 0.5) - 5}{\sqrt{5}}\right) \\ &= P(Z \leq 2.46) = \Phi(2.46) = 0.9931 \end{aligned}$$











