# STANDARD NORMAL DISTRIBUTION

INFERENTIAL STATISTICS

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# TOPIC OUTLINE

**Standard Normal Distribution** 

**Central Limit Theorem** 



# STANDARD NORMAL DISTRIBUTION



# **STANDARDIZATION**

**Standardization** is the process of converting the distribution of a variable  $X \sim (\mu, \sigma^2)$  to a normal distribution  $Z \sim N(0, 1)$ .

#### **Formula**

$$Z = \frac{X - u}{\sigma}$$

#### where

Z = z-score

X = random variable

## <u>example</u>

Data	$X-\mu$ Z-score	
1	-4	-1.46
2	-3	-1.095
3	-2 -0.73	
4	-1 -0.365	
5	0 0	
6	1	0.365
7	2	0.73
8	3	1.095
9	4	1.46

$$\bar{x} = 5.0$$
  $s = 2.74$ 



# **STANDARDIZATION**

**Standardization** is the process of converting the distribution of a variable  $X \sim (\mu, \sigma^2)$  to a normal distribution  $Z \sim N(0, 1)$ .

#### **Formula**

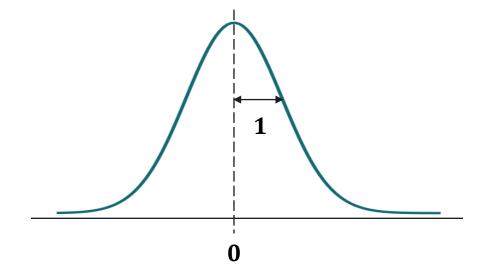
$$Z = \frac{X - u}{\sigma}$$

#### <u>where</u>

$$Z = z$$
-score

X = random variable

#### **Normal Distribution**





# STANDARD NORMAL DISTRIBUTION

### When we standardize the **normal distribution**

 $X \sim N(\mu, \sigma^2)$ , the result is a **standard normal distribution**  $Z \sim N(0, 1)$ .

#### **Formula**

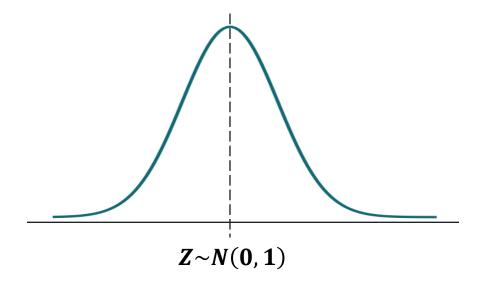
$$Z = \frac{X - u}{\sigma}$$

#### <u>where</u>

$$Z = z$$
-score

X = random variable

#### **Standard Normal Distribution**





# **EXERCISE**

Convert the given dataset into a <u>standard normal</u> <u>distribution</u> N(0, 1) by computing the **z-score** for each data point.

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Dala
1
2
2
3
3
3
4
1 2 2 3 3 3 4 4 4
5

### solution

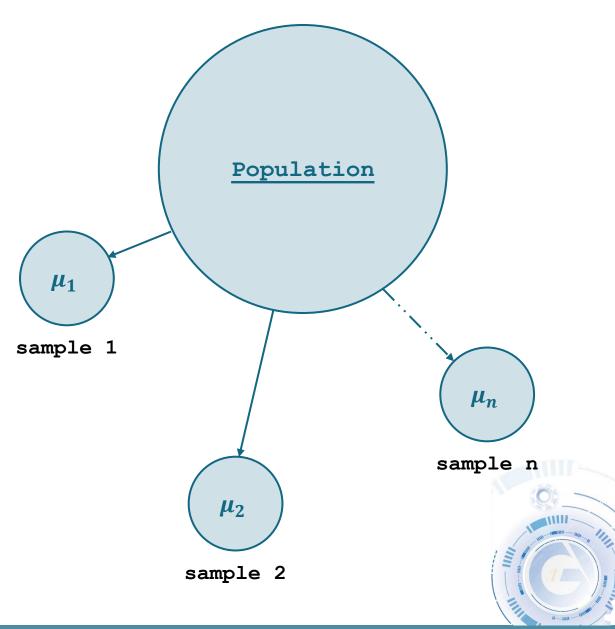


# **CENTRAL LIMIT THEOREM**



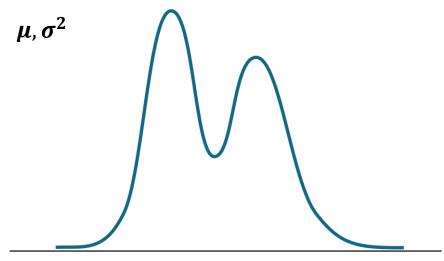
## **CENTRAL LIMIT THEOREM**

The <u>Central Limit Theorem</u> (CLT) states that the sampling distribution of the <u>sample mean</u> will be normally distributed, regardless of the shape of the original population distribution.



# **CENTRAL LIMIT THEOREM**

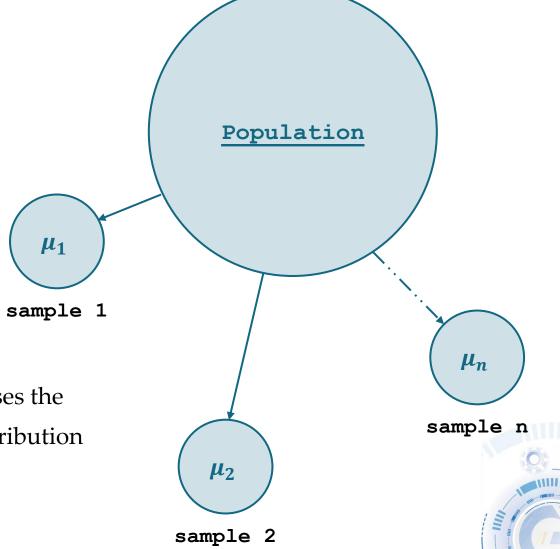
Original Population Distribution



<u>Sampling Distribution</u>

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

As the sample size n increases the variance  $\frac{\sigma^2}{n}$  of sampling distribution decreases.



## **SAMPLING DISTRIBUTION**

A <u>sampling distribution</u> is the probability distribution of a <u>statistic</u> (e.g.,  $\mu$ ,  $\sigma^2$ ) obtained from a large number of samples drawn from a specific population.

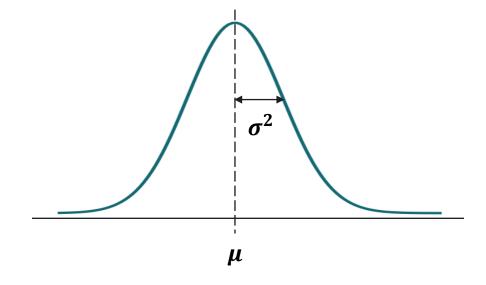
#### Denoted by

$$N\left(\mu, \frac{\sigma^2}{n}\right)$$
 ,  $n > 30$ 

#### where

 $\frac{\sigma^2}{n}$  = variance of the sampling distribution

## **Sampling Distribution**





## STANDARD ERROR

**Standard error** is the **standard deviation** of the distribution formed by the **sample means**.

#### **Formula**

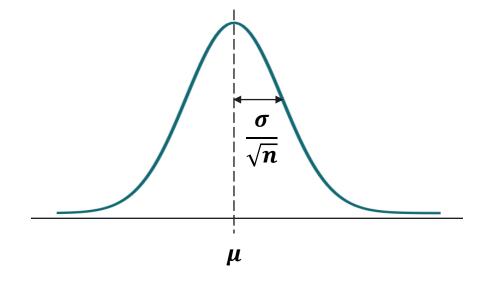
$$SE = \frac{\sigma}{\sqrt{n}}$$

#### where

 $\sigma$  = sampling standard deviation

n = number of observations

## **Sampling Distribution**





# **LABORATORY**

