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قسم التقنيات الاحيائية

LECTURE: (4)

Subject: Distribution of the Sample Mean and the Central Limit Theorem

Level: First

Lecturer: Dr. Mustafa Talal

Distribution of the Sample Mean and the Central Limit Theorem

Up to this point, the probabilities we have found have been based on individuals in a sample, but suppose we want to find probabilities based on the mean of a sample. In order for us to find these probabilities we need to know determine the sampling distribution of the sample mean. Knowing the sampling distribution of the sample mean will not only allow us to find probabilities, but it is the underlying concept that allows us to estimate the population mean and draw conclusions about the population mean which is what inferential statistics is all about. Sampling Error: The error resulting from using a sample to estimate a population characteristic. For a variable x and a given sample size n , the distribution of the variable \bar{x} (all possible sample means of size n) is called the sampling distribution of the mean. Note: The larger the sample size the smaller the sampling error tends to be in estimating a population mean, μ , by a sample mean \bar{x}

Mean of \bar{x} : denoted $\mu_{\bar{x}}$

For samples of size n , the mean of the variable \bar{x} equals the mean of the variable under consideration,

i.e. $\mu_{\bar{x}} = \mu$, where $\mu_{\bar{x}}$ is the mean of variable \bar{x} and μ is the population mean. In other words, the mean of all possible sample means of size n equals the population mean.

Example:

The following data represent the ages of the winners (age, in years, at time of award given) of the Academy Award for Best Actress for the years 2012 – 2017.

2012: Meryl Streep	62
2013: Jennifer Lawrence	22
2014: Cate Blanchet	44
2015: Julianne Moore	54
2016: Brie Larson	26
2017: Emma Stone	28

a) Calculate the population mean, μ . _____

b) The following table consisting of all possible samples with size $n = 2$ and calculate their corresponding means.

Sample	Mean	Sample	Mean	Sample	Mean	Sample	Mean	Sample	Mean	Sample	Mean
62, 62		22, 62		44, 62		54, 62		26, 62		28, 62	
62, 22		22, 22		44, 22		54, 22		26, 22		28, 22	
62, 44		22, 44		44, 44		54, 44		26, 44		28, 44	
62, 54		22, 54		44, 54		54, 54		26, 54		28, 54	
62, 26		22, 26		44, 26		54, 26		26, 26		28, 26	
62, 28		22, 28		44, 28		54, 28		26, 28		28, 28	

c) Calculate the mean of the sampling distribution of the mean, $\mu_{\bar{x}}$.
 _____ (i.e., calculate the mean of the sample means)

d) What do you conclude about μ and $\mu_{\bar{x}}$? _____

Note: The above example is exactly that an example, it is not a proof of
 $\mu_{\bar{x}} = \mu$.

Standard deviation of \bar{x} denoted $\sigma_{\bar{x}}$

For samples of size n , the standard deviation of the variable \bar{x} equals the standard deviation of the population under consideration divided by the square root of n . So,

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

\bar{x} is smaller than $\sigma_{\bar{x}}$ is sometimes called the **standard error** of the mean.

Example: Using the example above,

- a) Calculate the population standard deviation, σ . _____
- b) Calculate the standard deviation of the sampling distribution of the mean, $\sigma_{\bar{x}}$. _____ (i.e. calculate the standard deviation of the sample means)

$$n = \text{_____} \quad \sum \bar{x} = \text{_____} \quad \sum \bar{x}^2 = \text{_____}$$

- c) Use the formula $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$ to calculate $\sigma_{\bar{x}}$. _____
- d) What do you conclude from parts b and c? _____

Again, the example is an example not a proof.

The Sampling Distribution of the Sample Mean for a Normally Distributed Variable

Suppose that a variable x of a population is normally distributed with a mean μ and a standard deviation σ . Then, for samples of size n , the variable \bar{x} is also normally distributed and has mean and standard deviation

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

Of course, not all distributions are normal, but given certain conditions we can assume the variable \bar{x} is approximately normally distributed regardless of the distribution of x . This leads to one of the most important theorems in statistics; the central limit theorem.

Central Limit Theorem (CLT) For a relatively large sample size ($n > 30$), the variable \bar{x} is approximately normally distributed, regardless of the distribution of the variable under consideration. The approximation becomes better with increasing sample size.

