

1-Jul-2022

Term 2 / Week 10

Central Limit Theorem - Examples

Ex.1 Mobile phone bills for residents in a city have a mean (μ) of \$65 and a std. dev. (σ) of \$10. The data is known to have "Normal Distribution". Calculate the probability that phone bill of an individual is greater than \$70.

We have $\mu = 65$, $\sigma = 10$
 & $x = 70$

To Find $P[x > 70]$

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- Let us say that the data distribution in Ex.1 is unknown or known to be NOT normally distributed.
- However, the population Mean (μ) and Std. Dev (σ) can still be evaluated from the population data.
- But we cannot use "Standard Normal Table" for calculating the probability for a given data value (x)
- This is when the Central Limit Theorem becomes useful.

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$$\text{Calculate } Z = \frac{x - \mu}{\sigma} = \frac{70 - 65}{10} = \frac{5}{10} = 0.5$$

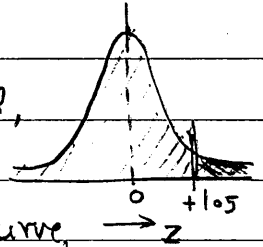
Since the data is

Normally distributed,

we can use "Standard

Normal Probability" curve,

using the z value.



$$P[z > 0.5] = 1 - P[z \leq 0.5]$$

$$= 1 - 0.6915$$

$$= 0.3085 \text{ (30.85\%)}$$

$$\therefore P[\text{Bill} > \$70] = \underline{\underline{30.85\%}}$$

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- Central Limit Theorem enables calculation of the probability of the Mean Value of a given sample (\bar{x}) instead of probability of a given data value (x).
- As per CLT, in general, the sample size (n) should be ≥ 30 . However, it can be less depending on the "Normalness" of the data distribution.
- Lower the sample size when the distribution is closer to "Normal Distribution"
- For perfect Normal Distribution sample size (n) can be 1 as in Ex.1!!

Ex.2 The population in Ex.1 is given to be NOT normally distributed.

Calculate the probability of obtaining a Mean value (\bar{x}) of \$70 for a given sample of 36 bills.

We have: $\mu = 65$; $\sigma = 10$
 $n = 36$; $\bar{x} = 70$

From Central Limit Theorem, sample means are normally distributed with :

Mean of Sample Means ($\mu_{\bar{x}}$) = 65
and Std. dev. of Sample Means ($\sigma_{\bar{x}}$) = $\frac{\sigma}{\sqrt{n}}$

Ex.3 (From Week 8)

The nicotine content in a cigarette of a particular brand is found to have a population mean (μ) of 0.8 mg and a popln std dev (σ) of 0.1 mg. If an individual smokes 100 cigarettes in a week, what is the probability the total amount of nicotine consumed is ≥ 82 mg / week.

We have: (for population)

$\mu = 0.8$ mg ; $\sigma = 0.1$ mg

Sample size (n) = 100

Sample Mean (\bar{x}) = $\frac{82}{100} = 0.82$ mg

(i.e., 82 mg consumption for 100 cigarettes)

$$\therefore \sigma_{\bar{x}} = \frac{10}{\sqrt{36}} = \frac{10}{6} \approx 1.6667$$

We need to find $P[\bar{x} > 70]$

we have

$$Z_{\bar{x}} = \frac{\bar{x} - \mu_{\bar{x}}}{\sigma_{\bar{x}}} = \frac{70 - 65}{1.6667} = 2.9999 \ (\approx 3.0)$$

$$P[Z_{\bar{x}} > 3.0] = 1 - P[Z_{\bar{x}} \leq 3.0] = 1 - 0.9987 = 0.0013 \ (0.13\%)$$

$\therefore P[\text{Mean Bill for Sample} \geq \$70] = 0.13\%$

Note: The probability is much lower, since Sample Mean "averages out" individual bills!

From Central Limit Theorem:

$$\mu_{\bar{x}} = 0.8 \text{ mg} \quad \sigma_{\bar{x}} = \frac{0.1}{\sqrt{100}} = 0.01 \text{ mg}$$

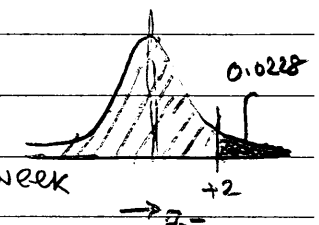
To Find: $P[\bar{x} \geq 0.82 \text{ mg}]$

$$\text{We have } Z_{\bar{x}} = \frac{\bar{x} - \mu_{\bar{x}}}{\sigma_{\bar{x}}} = \frac{0.82 - 0.80}{0.01} = +2.0$$

$$P[Z_{\bar{x}} > 2] = 1 - P[Z_{\bar{x}} \leq 2] = 1 - 0.9772 = 0.0228 \ (2.28\%)$$

\therefore Probability of

Consuming ≥ 82 mg / week



$$= 2.28\%$$