

Central Limit Theorem - Examples  
CCLT)

- Central Limit Theorem enables analysis of population with unknown distribution.
- It is done by considering "Sample Mean" ( $\bar{x}$ ) rather than individual values
- As per CLT, Sample Means are normally distributed with:

Mean ( $\mu_{\bar{x}}$ ) =  $\mu$

Std Dev ( $\sigma_{\bar{x}}$ ) =  $\sigma / \sqrt{n}$

where:  $\mu$  - Population Mean  
 $\sigma$  - Population Std. Dev.  
 $n$  - Sample size

-3-

Ex. 1 Mobile phone bills for residents in a city have a mean ( $\mu$ ) of \$63 and a Std. Dev ( $\sigma$ ) of \$10. The data distribution is unknown. Calculate and plot the distribution of Sample Means for following sample sizes.

(a)  $n = 100$  & (b)  $n = 36$

(a) We have  $\mu = 63$  &  $\sigma = 10$   
∴ for Sample Means,

Mean of Sample Means  $\mu_{\bar{x}} = \mu = 63$

Std. Dev. of Sample Means  $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}} = \frac{10}{\sqrt{100}}$

(Also called Standard Error of Means) = 1

• For Non-normal distribution, the sample size ( $n$ ) is  $\geq 30$ .

(Greater the sample  $\Rightarrow$  Better the approx)

• CLT can also be used for population with normal distribution! In such a case, there is no restriction on sample size!

• Lastly, note that we can always calculate the population Mean ( $\mu$ ) and population Std. Dev ( $\sigma$ ) for any given population.

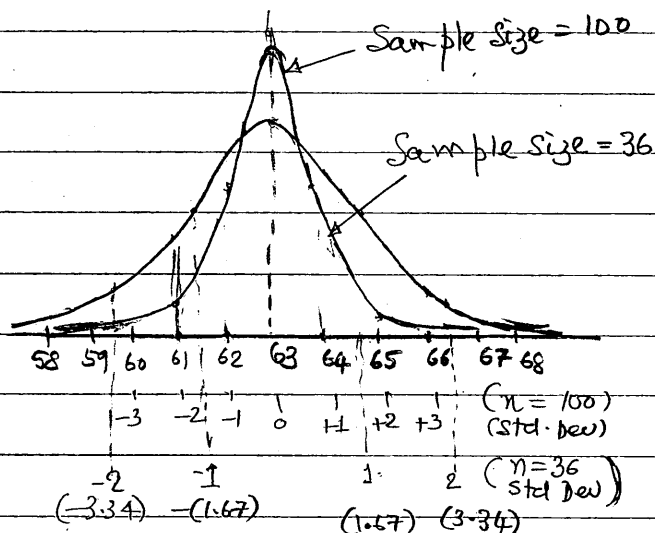
• Quite often population  $\mu$  &  $\sigma$  are obtained from historical data.

-4-

(b) Similarly, for  $n = 36$

Mean of Means:  $\mu_{\bar{x}} = \mu = 63$

Std. Dev of Means  $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}} = \frac{10}{\sqrt{36}} = \underline{\underline{1.67}}$



• Note that larger sample size gives smaller std. dev (less std. error!)

- In the above Example, a smaller sample size has its own its own advantages! It gives better information on the variability of the population data!
- A large sample size provides a more accurate population Mean, but hides the variability of the data!!

Sample size (n) = 50

For sample means ( $\bar{x}$ ):

$$\mu_{\bar{x}} = 25 \text{ min}$$

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}} = \frac{1.5}{\sqrt{50}} = 0.212 \text{ min.}$$

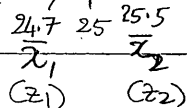
Given values for calculations

$$\bar{x}_1 = 24.7 \text{ min}$$

$$\bar{x}_2 = 25.5 \text{ min}$$



$$z_1 = \frac{\bar{x}_1 - \mu_{\bar{x}}}{\sigma_{\bar{x}}}$$



$$= \frac{24.7 - 25}{0.212} = -1.415$$

Similarly  $z_2 = \frac{25.5 - 25}{0.212} = +2.358$

$$P[x_1 : x_2] = P[\bar{x}_2] - P[\bar{x}_1]$$

$$= 0.9909 - 0.0786$$

$$= 0.9123 \text{ [91.2\%]}$$

Ex.2

In a regional town, it is given that the time spent driving has a mean ( $\mu$ ) of 25 min and a std. Dev ( $\sigma$ ) of 1.5 min. The distribution of data is unknown.

If a sample of 50 drivers are randomly selected, what is the probability that the mean driving time is between 24.7 and 25.5 min.

We have:

Population Mean ( $\mu$ ) = 25 min  
 Population std. Dev ( $\sigma$ ) = 1.5 min

Home work

The mean boarding & room expenses per year at four-year colleges is \$7540.

If a 9 four-year colleges are randomly selected, what is the probability that The mean board & room expenses is less than \$7800.

It is also given that board & room expenses are normally distributed with a std. dev. of \$1245.