

- Statistical Inference is divided into two major areas:
 - parameter estimation
 - [Statistical Inference, Hypothesis Testing](#)

Parameter Estimation:

- A statistic (mean etc.) that is gained from a sample of a population is called a **point estimate** of that parameter/statistic.
- A statistic is also a random variable, so it has a probability distribution.
- The probability distribution of a statistic is called a **sampling distribution**.
- θ will be used to represent any parameter, such as mean μ or variance σ^2 etc.
- The objective of point estimation is to select a single number, based on sample data, that is the most plausible value for θ .

A point estimate of some population parameter θ is a single numerical value $\hat{\theta}$ of a statistic $\hat{\Theta}$. The statistic $\hat{\Theta}$ is called the point estimator.

- We will make point estimates of the following parameters:
 1. The sample mean $\hat{\mu}$
 2. The sample variance $\hat{\sigma}^2$ (or standard deviation $\hat{\sigma}$)
 3. The sample proportion \hat{p} of items in a population, x/n , where x is the number of items in a random sample of size n
 4. The difference in sample means of two populations, $\hat{\mu}_1 - \hat{\mu}_2$
 5. The difference in two population sample proportions, $\hat{p}_1 - \hat{p}_2$

Central Limit Theorem:

If X_1, X_2, \dots, X_n is a random sample of size n taken from a population with mean μ and finite variance σ^2 , and if \bar{X} is the sample mean, the form of the distribution of

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$$

as $n \rightarrow \infty$, is the standard normal distribution.

Example 1:

An electronics company manufactures resistors that have a mean resistance μ of 100 ohms and a standard deviation σ of 10 ohms. The distribution of resistance is normal. Find the probability that a random sample of $n = 25$ resistors will have an average resistance \bar{X} less than 95 ohms.

Solution:

The distribution is normal, and since because *linear functions of independent, normally distributed random variables are also normally distributed* so the sample mean will be:

$\mu_{\bar{X}} = \mu$, so $\mu_{\bar{X}} = 100$ and

$\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$, so $\sigma_{\bar{X}} = \frac{10}{\sqrt{25}} = 2$

After finding sample mean and sample standard deviation, we will solve normally by standardizing $\bar{X} = 95$ using these statistics and finding the value using Z-table.

- Regarding the difference in sample means of two populations, $\hat{\mu}_1 - \hat{\mu}_2$, the mean will be equal to the difference of the means of the population: be: $\mu_{\bar{X}_1 - \bar{X}_2} = \mu_{\bar{X}_1} - \mu_{\bar{X}_2} = \mu_1 - \mu_2$ AND standard deviation will be same but SUMMED instead of subtracted.