### **Confidence Intervals**

List of things to Do:-

- Compute the point estimate of  $\mu$
- Compute the confidence intervals about  $\mu$  with  $\sigma$  known
- Error term in the confidence interval
- Determine the sample size n for estimating the population means
- Explain the meaning of Confidence interval
- SAS command to calculate C.I (Confidence Interval)

# Basics

• A Population is any entire collection of <u>people</u>, animals, plants or things from which we may collect data. It is the entire group we are interested in, which we wish to describe or draw conclusions about.

**Example:-** The population for a study of infant health might be all children born in the U.K. in the 2000's.

• A **Sample** is a group of units selected from a larger group (<u>the population</u>). By studying the sample it is hoped to draw valid conclusions about the larger group.

**Example:-** The population for a study of infant health might be all children born in the U.K. in the 1980's. The sample might be all babies born on 7th May in any of the years.

# Basics

• A **Parameter** is a value, usually **unknown** (and which therefore has to be estimated), used to represent a certain population characteristic.

For example, the <u>population mean</u> is a parameter that is often used to indicate the <u>average</u> value of a quantity.

 $\mu$  represents the population mean.  $\sigma$  represents the population standard deviation. p represents the population proportion.

• A **Statistic** is a quantity that is calculated from a <u>sample</u> of data. It is used to give information about unknown values in the <u>corresponding population</u>.

For example, the <u>average of the data</u> in a sample is used to give information about the <u>overall average</u> in the <u>population</u> from which that sample was drawn.

- $\bar{x}$  will estimate  $\mu$ .
- s will estimate  $\sigma$ .
- $\tilde{p}$  will estimate p.

### **About Normal Distribution**

A normal distribution with mean  $\mu$  and variance  $\sigma^2$  is a statistic distribution with probability function

$$N(\mu,\sigma) = P(X) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

where  $x \in (-\infty, \infty), \mu \in R$  and  $\sigma \ge 0$ .

While statisticians and mathematicians uniformly use the term 'Normal Distribution' for this distribution, physicists sometimes call it a Gaussian distribution and, because of its curved flaring shape, social scientists refer to it as the Bell Curve.'

Normal to Standard Normal Let ,

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

$$N(0,1) = P(Z) = \frac{1}{\sqrt{2\pi}}e^{-\frac{Z^2}{2}}$$

At the back of the book the normal table used is based on the above formula.

# Finding the Area under the Standard Normal Curve

# Finding the value of $Z_{\alpha}$

# **Binomial to Standard Normal** $B(n,p) = \binom{n}{x} p^x (1-p)^{n-x}$

where n is the number of trials, r is the number of

success and p is the probability of success. If  $np(1-p) \ge 10$ , the binomial experiment is approximately Normal, with mean  $\mu = np$  and standard deviation  $\sigma = \sqrt{np(1-p)}$ .

Standardize the Binomial random variable X we obtain,

$$Z = \frac{x - \mu}{\sigma} = \frac{x - np}{\sqrt{np(1 - p)}}$$

### **Point Estimates**

**Def:-** A point estimate of a parameter is the value of a statistic that estimates the value of the parameter.

**Example:-** Suppose  $X_1, X_2, X_{10}$  are random samples which follow normal distribution with mean  $\mu$  and standard deviation s = 2.5. Hence in this problem the parameter  $\mu$  is unknown and we want to estimate  $\mu$ from the given 10 samples. In order to understand this fact more clearly we need following *definition:-*

**Unbiased Estimator:** An estimate s is said to be an unbiased estimator of the parameter  $\mu$  if  $E(S) = \mu$ . **Theorem:-** Let  $X_1, X_2, ..., X_n$  be the random sample from the population with mean  $\mu$ . The sample mean  $\overline{X}$ is an unbiased estimator for the population mean  $\mu$ .

**Solution:**- We have to show that  $E(\bar{X}) = \mu$ .

$$E(\bar{X}) = E(\frac{\sum X_i}{n}) = \frac{1}{n}E(\sum X_i) = \frac{1}{n}\sum E(X_i)$$

But we know that  $E(X_i) = \mu$ . Therefore the above equation boils down to

$$E(\bar{X}) = \frac{n\mu}{n} = \mu.$$

Note:- From the above proof we observed that  $X'_is$ and  $\overline{X}$  are the unbiased estimators of  $\mu$ . Which one is the best one, the  $X'_is$  or the  $\overline{X}$ . The answer is  $\overline{X}$  is the better unbiased estimator, also note that we can find many unbiased estimators. A good question is which is the best. The answer to this question is beyond the reach of this Class. Hence for your example the

$$\frac{X_1 + X_2 + \dots + X_{10}}{10} = \bar{X}$$

estimates  $\mu$ . Hence the unknown population mean has been estimated by some known fact. Isn't it great!!!!!

#### Computing the confidence intervals about $\mu$ with $\sigma$ known.

Assumption:- The data should be from a normal population or the sample size is greater than 30. The way to check the normality of the data if sample size is less than 30, is to plot the data and see whether it is approximately a straight line with no outliers. I will talk about outliers in SAS.

**Confidence Interval:** – A confidence interval estimate of a parameter consists of an interval of numbers, along with the measure of the likelihood that the interval contains the unknown parameter.

$$(\bar{x}-z_{rac{lpha}{2}}*rac{\sigma}{\sqrt{n}}\ ,\ \bar{x}+z_{rac{lpha}{2}}*rac{\sigma}{\sqrt{n}})$$

- $\bar{x}$  is the point estimate for  $\mu$ .
- $\alpha$  our level of confidence.
- the standard deviation of the sample mean. (What represents standard deviation of the sample mean  $\bar{x}$ ?)

#### C.I can be rewritten as

Point Estimate  $\pm$  Margin of Error

# Interpretation of Confidence Interval

Let  $X_1,..., X_k$  denote K samples, each of size n, from a population with mean  $\mu$  and standard deviation  $\sigma$ (Known). Let us construct  $(1 - \alpha)100\%$  C.I for each of these samples of size n. Let the C.I's are denoted by  $C_1,...,C_k$ . We will observe that, approximately  $(1 - \alpha)100\%$  of the C.I's will contain  $\mu$ .

Suppose k = 20 and  $\alpha = .10$ . Suppose 90% confidence interval for  $\mu$  is calculated as (25, 36). That means "we are 90% confident" that  $\mu$  lies between 25 and 36. In other words out of 20 C.I's approximately 18 of  $C'_k s$  will contain  $\mu$ .

#### Error term in the confidence interval

The margin of error term in confidence interval is denoted as  $\mathbf{E}$ .

$$\mathbf{E} = z_{\frac{\alpha}{2}} * \frac{\sigma}{\sqrt{n}}$$

The error  $\mathbf{E}$  suggests that if we increase the sample size then we will expect less error, that means the confidence interval will become more shallower, hence will be a good estimator of the unknown population parameter.

#### Determining the sample size n

The sample size required to estimate the population mean  $\mu$ , with a given level of confidence and with a specified margin of error(or minimum error tolerance) **E** is given by

$$n = (\frac{\frac{z_{\frac{\alpha}{2}} * \sigma}{\mathbf{E}}}{\mathbf{E}})^2$$

#### Miles on Cavalier

A researcher is interested in approximating the mean number of miles on three-year old Chevy Cavaliers. She finds a random sample in Orlando, Florida area and obtained the following result

37,815	20,000	57,103	46,585	24,822
49,678	30,983	52,969	8,000	39,862
6,000	65,192	34,285	30,906	41,841
39,851	43,000	74,362	52,664	33,587
52,896	45,280	30,000	41,713	76,315
22,442	43,301	52,899	41,526	28,381
55,163	51,812	36,500	31,947	16,529

• Obtain a point estimate of the population mean number of miles on a three-year old Cavalier.

**Ans:-**  $\bar{x} = 40,463.11$  miles is the point estimate for population mean. That means we can roughly claim that a three-year old Cavalier will have 40,463.11 miles on it.

• Construct a 99% confidence interval for the population mean number of miles on a three-year old Cavalier assuming that  $\sigma = 16,100$ . Ans:- The 99% confidence interval is given by

 $(40463.11 - z_{\underline{0.01}} * \frac{16100}{\sqrt{35}} , \ 40463.11 + z_{\underline{0.01}} * \frac{16100}{\sqrt{35}})$  or

(33456.14, 47469.97).

Hence we are 99% confident that the population mean number of miles will lie between 33456.14 and 47469.97.

• From the above result can we say something regarding the population mean number of miles in U.S.A. If not then how nicely the data describes the mean population miles for a three-year old car in an around Florida.

**Ans:-** No. As this is a local Florida data. The confidence interval is pretty big. Hence we can say that this interval is a moderate estimate of the population mean at 99% level of significance. In order to estimate the population mean more accurately we need to look at other factors, like sex, driving conditions, age group and so on...

#### Computing the confidence intervals about $\mu$ with $\sigma$ uknown.

**Assumption:-** The data should be from a normal population or the sample size is greater than 30.

Confidence Interval in this case is given by

$$(\bar{x}-t_{rac{lpha}{2}}*rac{s}{\sqrt{n}}$$
 ,  $\bar{x}+t_{rac{lpha}{2}}*rac{s}{\sqrt{n}})$ 

where t denotes the Student's-t distribution with n-1 degrees of freedom. Everything remains the same as above.

Using the same data set from the previous problem Construct a 99% confidence interval for the population mean number of miles on a three-year old Cavalier assuming that  $\sigma = unknown$ .

**Ans:-** The 99% confidence interval is given by

$$(40463.11 - t_{\frac{0.01}{2}} * \frac{s}{\sqrt{35}} , 40463.11 + t_{\frac{0.01}{2}} * \frac{s}{\sqrt{35}})$$

$$= (40463.11 - t_{\frac{0.01}{2}} * \frac{16149.65}{\sqrt{35}} , 40463.11 + t_{\frac{0.01}{2}} * \frac{16149.65}{\sqrt{35}})$$

(33015.17 , 47911.06).

Hence we are 99% confident that the population mean number of miles will lie between 33015 and 47911, when  $\sigma$  is unknown.

#### SAS Program

options linesize=80 nodate; run; title 'Miles in Three-Year old Chevy Cavalier'; data Chevy; input miles; cards; 37815 20000 57103 46585 24822 49678 30983 52969 8000 39862 6000 65192 34285 34285 30906  $41841 \\ 39851$ 39851 43000 74362 52664 33587 52896 45280 30000 41713  $76315 \\ 22442$  $\begin{array}{c} 22442\\ 43301\\ 52899\\ 41526\\ 28381\\ 55163\\ 51812\\ 36500\\ 31947\\ 16529 ; \end{array}$ proc print data=Chevy; run; proc means data=chevy; proc means data=Chevy lclm uclm alpha=0.01; VAR miles; run;

#### SAS Output

Miles in Three-Year old Chevy Cavalier in Orlando, Florida

Obs	miles	
1	37815	
2	20000	
3	57103	
4	46585	
5	24822	
6	49678	
7	30983	
8	52969	
9	8000	
10	39862	
11	6000	
12	65192	
13	34285	
14	30906	
15	41841	
16	39851	
17	43000	
18	74362	
19	52664	
20	33587	
21	52896	
22	45280	
23	30000	
24	41713	
25	76315	
26	22442	
27	43301	
28	52899	
29	41526	
30	28381	
31	55163	
32	51812	
33	36500	
34	31947	
35	16529	

SAS Output Continued...

The MEANS H	Procedure
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N	Mean	Std Dev	Maximum	Minimum
35	40463.11	16149.65	6000	76315.00

Lower 99% CL for Mean	Upper 99% CL for Mean
33015.17	47911.06

The dean of a college wants to use the mean of a random sample to estimate the average amount of time students take to get from one class to the next, and she wants to be able to assert with probability 0.95 that her error will be at most 0.30 minute. If she knows from studies of a similar kind that it is reasonable to let  $\sigma = 1.50$  minute, how large a sample will she need?

Ans:- The things that are given to us

- E = 0.30
- $\sigma = 1.50$
- $z\frac{.05}{2} = 1.96$

We need to find the sample size n.

$$n = \left(\frac{z_{\frac{\alpha}{2}} * \sigma}{\mathbf{E}}\right)^2 = \left(\frac{z_{\frac{0.05}{2}} * 1.50}{0.30}\right)^2 = 96.04$$

#### Computing the confidence intervals about a Population proportion.

Let p denote the population proportion. The best point estimate for p is denoted by  $\tilde{p}$ .

$$\frac{x}{n} = \tilde{p} \longrightarrow p$$

The **Confidence Interval** in this case is denoted by

$$(\tilde{p} - z_{\frac{\alpha}{2}} * \sqrt{\frac{\tilde{p}(1-\tilde{p})}{n}} \ , \ \tilde{p} + z_{\frac{\alpha}{2}} * \sqrt{\frac{\tilde{p}(1-\tilde{p})}{n}})$$

**Determining the Sample size n** The sample size required to obtain a  $(1 - \alpha)$ %100 confidence interval for p with a given margin of error E, is given by

$$n = \tilde{p}(1 - \tilde{p})(\frac{z_{\frac{\alpha}{2}}}{E})^2$$

#### The Death Penalty (8.3 Problem 21)

In a Harris Poll conducted in July, 2000, 64% of the people polled answered yes to the following question: "Do you believe in capital punishment, that is the death penalty, or are you opposed to it?" The margin of Error was  $\pm 3\%$ , and the estimate was made with 95% confidence. How many people were surveyed?

**Ans:-** The things that are given to us

- E = 0.03
- $\tilde{p} = 0.64$
- $z_{\frac{.05}{2}} = 1.96$

We need to find the sample size n.

$$n = \tilde{p}(1 - \tilde{p})(\frac{z_{\frac{\alpha}{2}}}{E})^2 = 0.64(1 - 0.64)(\frac{1.96}{0.03})^2 = 983.4$$

Hence 984 people were surveyed.

# Computing the confidence intervals about a Population Variance.

If a random sample of size n is obtained from a normally distributed population with mean  $\mu$  and standard deviation  $\sigma$ , then

$$\chi^2 = \frac{(n-1)s^2}{\sigma^2}$$

has a chi-square distribution with n-1 degrees of freedom.

The **Confidence Interval** about  $\sigma^2$  is given by

$$(\frac{(n-1)s^2}{\chi_{\frac{\alpha}{2}}^2}$$
 ,  $\frac{(n-1)s^2}{\chi_{\frac{1-\alpha}{2}}^2})$ 

#### Tests of Hypothesis

List of things to do:-

- Determine the null and alternative hypothesis from a claim
- Type I and Type II error
- Calculation of Power

#### Basic Definitions:- The Null Hypothesis is,

denoted by  $H_0$ , is a statement to be tested. The null hypothesis is assumed to be true until we have evidence against it.

The Alternative Hypothesis denoted by  $H_1$ , is a claim to be tested.

Ways to set up  $H_0$  and  $H_1$ :-

# Figuring out Null and Alternative Hypothesis from a given data.

The following data is a survey of 15 Blood pressure patients before and after taking B.P medication for 6 months.

No. of Observations	Before Medication	After Medication
1	140	134
2	123	126
3	150	130
4	174	140
5	134	120
6	147	142
7	132	125
8	150	120
9	190	150
10	162	133
11	167	135
12	168	142
13	145	129
14	145	120
15	159	130

Let  $\mu_0$  denote the mean Blood Pressure **before** taking Medication.

Let  $\mu_1$  denote the mean Blood Pressure **after** taking Medication.

$$H_0: \mu_0 = \mu_1$$
  
 $H_1: \mu_0 > \mu_1$ 

Hence we claim that after taking medication the Blood Pressure will go down.

#### Different Types of Hypothesis Testing

• Simple vs Simple Example:-

$$H_0: \mu = 0$$
$$H_1: \mu = 3$$

• Simple vs Composite Example:-

$$H_0: \mu = 0$$
$$H_1: \mu \ge 0$$

• Composite vs Composite Example:-

 $H_0: \mu < 0$  $H_1: \mu \ge 0$ 

# Type I and Type II Error

	$H_0$ Is true	$H_1$ Is True
Do not Reject $H_0$	$\checkmark$	Type II Error
Reject $H_0$	Type I Error	$\checkmark$

- Hence a Type I Error would occur if the Null Hypothesis is Rejected when, in Reality the Null Hypothesis is True.
- Power of a Hypothesis Test is the Probability of Rejecting Null Hypothesis when, in reality  $H_0$  is False. Hence its a correct decision.

We are now ready to define the <u>Level of Significance</u> and <u>Power</u> of a Hypothesis Test. **Level of Significance**:- The level of Significance,  $\alpha$ , is the probability of making a Type I Error.

$$\alpha = P(RejectingH_0|H_0isTrue)$$

**Probability of type II Error**:- Probability of Type II Error is denoted by  $\beta$  and is defined as

 $\beta = P(NotRejectingH_0 \mid H_1istrue).$ 

#### Probability of type II Error:-

**Remarks:-** In problems we fix  $\alpha$ , as

.05(95%), .01(99%), .1(90%). That means we fix the margin of error and the small it is , the good for the experiment. In general it is known that Type I error is more severe than Type II error, hence always want a small  $\alpha$ . On the other hand we will like to get a bigger Power to decide how powerful the test is.

**Problem 27.** Potato Consumption: – According to the Statistical Abstract of the United States, the mean per capita consumption of potatoes in 1999 was 48.3 pounds. A Researcher believes that the potatoes consumption has risen since then.

1. Determine the Null and the Alternative Hypothesis. Ans:-

$$H_0: \mu = 48.3$$
  
 $H_1: \mu > 48.3$ 

Suppose, in reality the mean capita consumption of potatoes is 48.3 pounds. Was a Type I or Type II error committed. If we test this hypothesis at the α = 0.05 level of significance, what is the probability of committing a Type I error.
 Ans:- Hence H<sub>0</sub> is true and its rejected, we are committing a Type I Error. And the Probability of Type I Error is α = 0.05.

Important Numbers

Confidence Level	$\alpha$	$z_{lpha}$
90%(One Sided Test)	0.100	1.285
95% (One Sided Test)	0.050	1.645
95% (Two Sided Test)	0.025	1.960
98%(One Sided Test)	0.02	2.055
99%(One sided Test)	0.010	2.325
99%(Two Sided Test)	0.005	2.575

#### Note:-

- $z_{\alpha}$  is a Number ranging from  $-\infty$  to  $+\infty$ .
- If the significance level is not stated then assume that we are talking about 95% confidence level and hence we will use  $\alpha = 0.05$ .

The Critical Region or Rejection Region for Two Tailed, Left Tailed and Right Tailed Hypothesis

Two Tailed:-

$$H_0: \mu = \mu_0$$
$$H_1: \mu \neq \mu_0$$

**Rejection Region**:-

$$Z_{Cal} > Z_{\frac{\alpha}{2}}.$$

or

$$Z_{Cal} < -Z_{\frac{\alpha}{2}}.$$

Where  $Z_{Cal} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$
Left Tailed:-

$$H_0: \mu = \mu_0$$
$$H_1: \mu < \mu_0$$

Rejection Region:-

$$Z_{Cal} < -Z\alpha.$$

Where  $Z_{Cal} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$ 

Right Tailed:-

$$H_0: \mu = \mu_0$$
$$H_1: \mu > \mu_0$$

Rejection Region:-

$$Z_{Cal} > Z\alpha.$$

Where  $Z_{Cal} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$ 

## Problems of Power

For a given population with  $\sigma = \$12$ , we want to test the null hypothesis  $\mu = \$75$  on the basis of random sample of size n = 100. If the Null hypothesis is rejected when  $\bar{x}$  is greater than or equal to \$76.50. and otherwise it is accepted, find

• the probability of Type I error.

#### Ans:-

 $\alpha$  = Probability of Type I Error = P(Reject  $H_0|H_0$ is True) The rejection region is given as  $\bar{x} \ge 76.50$ . Mathematically,

$$\alpha = P(x \ge 76.50 \mid H_0 isTrue)$$

$$\alpha = P(\bar{x} \ge 76.50 \mid \mu_0 = \$75)$$

$$\alpha = P(Z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} \ge \frac{76.50 - \mu_0}{\frac{\sigma}{\sqrt{n}}}) = P(Z \ge \frac{76.50 - 75}{\frac{12}{\sqrt{100}}})$$

$$\alpha = P(Z \ge \frac{76.50 - 75}{\frac{12}{\sqrt{100}}}) = P(Z \ge 1.25) = 0.1056$$
Hence level of confidence is approx. 90%.

• The power of the hypothesis when  $\mu_0 = 75.30$ . Is it a good power?

### Ans:-

Power = P(Reject  $H_0|H_1$  is True) Mathematically, Power =  $P(\bar{x} \ge 76.50| \mu_0 = 75.30)$ 

$$\begin{aligned} Power &= P(Z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} \geq \frac{76.50 - \mu_0}{\frac{\sigma}{\sqrt{n}}}) \\ &= P(Z \geq \frac{76.50 - 75.30}{\frac{12}{\sqrt{100}}}) \\ &= P(Z \geq \frac{76.50 - 75.30}{\frac{12}{\sqrt{100}}}) \\ &= P(Z \geq 1) = \mathbf{0.1587} \end{aligned}$$

**Conclusion:-** Its a Awful Power. A good Power should be close to 1. And .1587 is not.

The probability of Type II Error when μ = 77.22. Is it good? Ans:-

Power = P(Reject  $H_0|H_1$  is True) Mathematically, Power =  $P(\bar{x} \ge 76.50| \mu_0 = 77.22)$ 

$$\begin{aligned} Power &= P(Z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} \geq \frac{76.50 - \mu_0}{\frac{\sigma}{\sqrt{n}}}) \\ &= P(Z \geq \frac{76.50 - 77.22}{\frac{12}{\sqrt{100}}}) \\ &= P(Z \geq \frac{76.50 - 77.22}{\frac{12}{\sqrt{100}}}) \\ &= P(Z \geq -0.6) = \mathbf{0.7257} \end{aligned}$$

**Conclusion:-** Its a Good Power. A good Power should be close to 1. And .7257 is close to 1.

**Problem 5, Sec 9.6**:- In order to test  $H_0: \mu = 20$  v.s  $H_1: \mu < 20$ , a simple random sample of size n = 18 is obtained from a population that is known to be normally distributed with  $\sigma = 3$ .

What would it mean to make a Type II Error?
 Ans:- A Type II Error would occur if the sample data led to a conclusion of not rejecting μ = 20 when in fact μ < 20.</li>

 If the researcher decides to test this hypothesis at α = 0.05 level of significance, compute the Power of the test if the true population mean is 17.4. What is P(Type II Error)? Ans:

Power = P(Reject 
$$H_0|H_1$$
 is True)

Computing The Rejection Region:- Observe that this is a Left Tailed Test , hence the rejection region is given in terms of  $\bar{x}$  as follows

$$\bar{x} < \mu - z_{\alpha} \frac{\sigma}{\sqrt{n}}$$

or

$$\bar{x} < 20 - 1.645 \frac{3}{\sqrt{18}} = 18.84$$

Power = P(Reject  $H_0|H_1$  is True)

$$= P(\bar{x} < 18.84 \mid \mu = 17.40)$$
  
Standardizing we get  
$$= P(Z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} \le \frac{18.84 - \mu_0}{\frac{\sigma}{\sqrt{n}}}) = P(Z \le \frac{18.84 - 17.40}{\frac{3}{\sqrt{18}}})$$
  
$$= P(Z \le 2.04)$$
  
$$= 0.9793$$

Geometrically:-

• Redo part b if the true population mean is 19.2.

**Problem 12, Sec 9.6**:- A school administrator states that students whose first language learned is not english do not score differently on the Math portion of the SAT Exam from students whose first language is English. The mean SAT math score of students whose first language is english is 516., according to the data obtained from the College Board. Suppose the researcher obtains a simple random sample of 20 students whose first language learned was not english, SAT math scores are normally distributed with a population standard standard deviation of 109.

What would it mean to make a Type II Error?
 Ans:- A Type II Error would occur if the sample data led to a conclusion of not rejecting μ = 516 when in fact μ ≠ 516.

 If the researcher decides to test this hypothesis at α = 0.05 level of significance, compute the Probability of Type II Error if the true population mean is 505. What is Power of the test?

#### Ans:-

 $P(Type II Error) = P(Do Not Reject H_0|H_1 is True)$ 

Computing The Rejection Region: – Observe that this is a **Two Tailed Test**, hence the rejection region is given in terms of  $\bar{x}$  as follows

$$\bar{x} > \mu + z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}$$
 or  $\bar{x} < \mu - z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}$ 

Hence do not reject  $H_0$  if ,  $\mu - z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}} < \bar{x} < \mu + z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}$ or  $516 - z_{\frac{0.05}{2}} \frac{109}{\sqrt{20}} < \bar{x} < 516 + z_{\frac{0.05}{2}} \frac{109}{\sqrt{20}}$  $516 - 1.96 \frac{109}{\sqrt{20}} < \bar{x} < 516 + 1.96 \frac{109}{\sqrt{20}}$ 

Hence D not reject Reject  $H_0$  if  $\bar{x}$  lies (468.23, 563.77).

 $P(Type II Error) = P(Do Not Reject H_0|H_1 is True)$ 

 $= P(468.23 < \bar{x} < 563.77 | \mu = 505)$ 

Standardizing we get..

$$= P(\frac{468.23 - 505}{\frac{109}{\sqrt{20}}} < Z < \frac{563.77 - 505}{\frac{109}{\sqrt{20}}})$$
$$= P(-1.51 < Z < 2.41)$$
$$= .9920 - 0.0655 = 0.9265 = \beta$$

Hence the **Power** of this Two Tailed Test is 1 - .9265 = 0.0735. Which is a **Bad Power**.

# The Classical method of Testing Hypothesis about $\mu\,\text{,}\,\,\sigma\,\,\text{known}$

Steps used in Classical method of Testing Hypothesis about  $\mu$ ,  $\sigma$  known...

- **Step 1** Figuring out the **Null** and **Alternative** Hypothesis.
- **Step 2** Write down  $\alpha$  value given in the problem and find the Critical Value  $Z_{tab}$ ,  $z_{\alpha}$  or  $z_{\frac{\alpha}{2}}$  depending on the nature of the test. Also Draw the rejection region or the critical region.
- Step 3 Compute the test Statistic

$$Z = Z_{Cal} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$$

**Step 4** Compare the critical value from the test Statistic. In other words compare  $Z_{Cal}$  with  $Z_{tab}$ .

**Step 5** State the conclusion.

**Note:-** The procedure presented requires that the data comes from a population that is Normally distributed or the sample size is more that 30.

**Problem 12, Sec 9.2**:- A school administrator states that students whose first language learned is not english do not score differently on the Math portion of the SAT Exam from students whose first language is English. The mean SAT math score of students whose first language is english is 516., according to the data obtained from the College Board. Suppose a simple random sample of 20 students whose first language learned was not english, results in a sample mean SAT math score of 518. SAT math scores are normally distributed with a population standard standard deviation of 109.

• Why it is necessary for SAT math scores to be normally distributed in order to test the claim using the Steps?

**Ans:-** Since the sample size is less than 20, we need the SAT math scores to be Normally distributed.

• Test the Researcher's claim using the Classical Approach at the  $\alpha = 0.10$  level of significance. Ans:-

The Classical Approach

Step 1

$$H_0: \mu = 516$$
$$H_1: \mu \neq 516$$

**Step 2**  $\alpha = 0.10$ . As this is a two tailed test the rejection region is given as  $Z \ge Z_{\frac{\alpha}{2}} = 1.645$  or  $Z \le -Z_{\frac{\alpha}{2}} = -1.645$ . Geometrically

Step 3 The Test Statistic

$$Z_{Cal} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} = \frac{518 - 516}{\frac{109}{\sqrt{20}}} = 0.08$$

- **Step 4** As the test statistic  $Z_{Cal}$  doesn't lie in the critical region, we do not reject the Null Hypothesis.
- Step 5 Hence there is not enough evidence to support the claim that the mean SAT math score of Students whose first language is not english is different from 516.

### Testing a Hypothesis about $\mu$ , $\sigma$ known, p- Value Approach

Steps used in Testing Hypothesis about  $\mu$ ,  $\sigma$  known, p-value approach...

- **Step 1** Figuring out the **Null** and **Alternative** Hypothesis.
- **Step 2** Compute the test Statistic

$$Z = Z_{Cal} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$$

Step 3 Determine the *p*- value. For Two tailed test the P-Value is  $2P(Z > |Z_{Cal}|)$ . For Left tailed test the P-Value is  $P(Z < Z_{Cal})$ . For Right tailed test the P-Value is  $P(Z > Z_{Cal})$ .

- **Step 4** Reject the Null Hypothesis if the P-Value is <u>less</u> than the level of Significance  $\alpha$ .
- Step 5 State the conclusion.

#### Same Problem 12, Sec 9.2:- A school

administrator states that students whose first language learned is not English do not score differently on the Math portion of the SAT Exam from students whose first language is English. The mean SAT math score of students whose first language is English is 516., according to the data obtained from the College Board. Suppose a simple random sample of 20 students whose first language learned was not English, results in a sample mean SAT math score of 518. SAT math scores are normally distributed with a population standard standard deviation of 109.

Test the Researcher's claim using the Classical Approach at the  $\alpha = 0.10$  level of significance. **Ans:-**

#### The p-value Approach

Step 1

$$H_0: \mu = 516$$
$$H_1: \mu \neq 516$$

Step 2 The Test Statistic

$$Z_{Cal} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} = \frac{518 - 516}{\frac{109}{\sqrt{20}}} = 0.08$$

- **Step 3** p- Value is 2P(Z > |0.08|) = 2P(Z > 0.08) = 2(1 0.5319) = 2(0.4681) = 0.9362
- **Step 4** As the test statistic p-value is not less than  $\alpha = 0.05$ , we do not reject the Null Hypothesis.
- Step 5 Hence there is not enough evidence to support the claim that the mean SAT math score of Students whose first language is not English is different from 516.

### T Test

# The Classical method of Testing Hypothesis about $\mu\,\text{,}\,\,\sigma$ unknown

Steps used in Classical method of Testing Hypothesis about  $\mu$ ,  $\sigma$  unknown...

- **Step 1** Figuring out the **Null** and **Alternative** Hypothesis.
- **Step 2** Write down  $\alpha$  value given in the problem and find the Critical Value  $t_{tab}$ ,  $t_{\alpha}$  or  $t_{\frac{\alpha}{2}}$  and also mention the degrees of freedom. Also Draw the rejection region or the critical region.

### Step 3 Compute the test Statistic

$$t = t_{Cal} = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}}$$

**Step 4** Compare the critical value from the test Statistic. In other words compare  $t_{Cal}$  with  $t_{tab}$ .

Step 5 State the conclusion.

**Note:-** The procedure presented requires that the data comes from a population that is Normally distributed or the sample size is more that 30.

### Testing a Hypothesis about $\mu$ , $\sigma$ unknown, p- Value Approach

Steps used in Testing Hypothesis about  $\mu$ ,  $\sigma$  unknown, p-value approach...

- **Step 1** Figuring out the **Null** and **Alternative** Hypothesis.
- Step 2 Compute the test Statistic

$$t = t_{Cal} = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}}$$

- **Step 3** Determine the *p* value.
- **Step 4** Reject the Null Hypothesis if the P-Value is <u>less</u> than the level of Significance  $\alpha$ .
- **Step 5** State the conclusion.

**Problem 15, Sec 9.3**:- In 1989, the average age of an inmate on a death row was 36.2 years of age, according to the data obtained from U.S Department of justice. A sociologist wants to test the **Claim that the average age of a death row inmate has changed since then**. She randomly selects 32 death row inmates and finds that there mean age is **38.9**, with a standard deviation of 9.6.

• Using the Classical approach test the sociologist's claim at the  $\alpha = 0.05$  level of significance. Ans:-

Step 1

$$H_0: \mu = 36.2$$
  
 $H_1: \mu \neq 36.2$ 

**Step 2** This is a two sided test, with 31 df and so the critical values are  $\pm t_{0.025} = \pm 2.040$ .

Step 3

•

$$t = t_{Cal} = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} = \frac{38.9 - 36.2}{\frac{9.6}{\sqrt{32}}} = 1.591$$

- Step 4 1.591 doesn't lie in the critical region, hence we will not reject the NULL HYPOTHESIS.
- Step 5 Hence there is not enough evidence to support the claim that the mean age of death -row inmates is different from 36.2 years.

• Determine and Interpret the P-value. Ans:-

How to find the P value from given  $t_{Cal}$  and df?

Solution to Problem 20, Sec 9.3:-Step 1

$$H_0: \mu = 1.62$$
  
 $H_1: \mu < 1.62$ 

**Step 2** This is a one sided test, with 11 df and so the critical values are  $-t_{0.10} = -1.363$ .

Step 3

$$t = t_{Cal} = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} = \frac{1.6141 - 1.62}{\frac{0.0053}{\sqrt{12}}} = -3.856$$

- **Step 4** -1.363 lie in the critical region, hence we reject the NULL HYPOTHESIS.
- Step 5 Hence there is enough evidence to support the claim that the <u>mean wt</u> of Maxfli XS Golf balls is less than 1.62 ounces.

Determine and Interpret the P-value. Ans:-

How to find the P value from given  $t_{Cal}$  and df?

# The Classical method of Testing Hypothesis about population proportion

Steps used in Classical method of Testing Hypothesis about population proportion

- **Step 1** Figuring out the **Null** and **Alternative** Hypothesis.
- **Step 2** Write down  $\alpha$  value given in the problem and find the Critical Value  $Z_{tab}$ ,  $z_{\alpha}$  or  $z_{\frac{\alpha}{2}}$  depending on the nature of the test. Also Draw the rejection region or the critical region.
- Step 3 Compute the test Statistic

$$Z = Z_{Cal} = \frac{\ddot{p} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}}$$

Where  $\tilde{p}$  is the sample Proportion.

- **Step 4** Compare the critical value from the test Statistic. In other words compare  $Z_{Cal}$  with  $Z_{tab}$ .
- **Step 5** State the conclusion.

Note:- For Normality you should check  $np_0(1-p_0) \ge 10$ 

Solution to Problem 12, Sec 9.4:-

Step 1

$$H_0: p_0 = 0.249$$
  
 $H_1: p_0 < 0.249$ 

Note:-

 $np_0(1-p_0) = 150 * 0.249(1-0.249) = 28 \ge 10$ , hence the normality requirement is fulfilled.

- **Step 2**  $\alpha = 0.05$ . As this is a Left tailed test the rejection region is given as  $Z \leq -Z_{\alpha} = -1.645$ . From the Survey,  $\tilde{p} = \frac{28}{150} = 0.187$
- Step 3 The Test Statistic

$$Z_{Cal} = \frac{\tilde{p} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}} = \frac{0.187 - 0.249}{\sqrt{\frac{0.249(1 - 0.249)}{150}}} = -1.76$$

- **Step 4** As the test statistic  $Z_{Cal}$  lies in the critical region, we reject the Null Hypothesis.
- Step 5 Hence there is enough evidence to support the claim that the percentage of 30-40 year old male who do not exercise has decreased from its 1998 level.

# Inference on Two Samples

- Paired Tests(Dependent Samples).
- Welch's t test(Independent Samples).
- Pooled Two sample t test.
- F Test

### Paired Tests(Dependent Samples).

**Definition:-** The sampling method is **dependent** when the individuals selected in one sample are used to determine the individuals to be in the second sample. As an example we use Paired T test, when, Before and After kinds of comparison, the ages of Husbands and wives, first half and second half of the exam, cars stocked and sold by used car dealers, and numerous other kinds of situations in which data are naturally Paired. Steps..

Assumptions:-

1. The sample is obtained using simple random sampling. 2. the sample data are paired.

3. the differences are normally distributed or , the sample size is greater than 30.

**Step 1** Figuring out the **Null** and **Alternative** Hypothesis.

**Step 2** Write down  $\alpha$  value given in the problem and find the Critical Value  $t_{tab}$ ,  $t_{\alpha}$  or  $t_{\frac{\alpha}{2}}$  and also mention the degrees of freedom. Also Draw the rejection region or the critical region.

Step 3 Compute the test Statistic

$$t = t_{Cal} = \frac{\bar{d} - 0}{\frac{s_d}{\sqrt{n}}} = \frac{\bar{d}}{\frac{s_d}{\sqrt{n}}}$$

**Step 4** Compare the critical value with the test Statistic. In other words compare  $t_{Cal}$  with  $t_{tab}$ .

**Step 5** State the conclusion.

### **Confidence Intervals**

A  $(1 - \alpha).100\%$  confidence interval for  $\mu_d$  is given by

$$(\bar{d}-t_{\frac{\alpha}{2}},\frac{s_d}{\sqrt{n}},\bar{d}+t_{\frac{\alpha}{2}},\frac{s_d}{\sqrt{n}})$$

### Problem 8, Sec 10.1:-

Observations	1	2	3	4	5	6	7	8
$X_1$	19.4	18.3	22.1	20.7	19.2	11.8	20.1	18.6
$X_2$	19.8	16.8	21.1	22.0	21.5	18.7	15.0	23.9

1. Determine  $d_i = X_1 - X_2$ .

### Ans:-

$X_1$	$X_2$	$X_1 - X_2$
19.4	19.9	-0.4
18.3	16.8	1.5
22.1	21.1	1.0
20.7	22.0	-1.3
19.2	21.5	-2.3
11.8	18.7	-6.9
20.1	15.0	5.1
18.6	23.9	-5.3

2. Compute  $\bar{d}$  and  $s_d$ .

### Ans:-

options linesize=80 nodate; run; data pair; *infile 'path'* input chaos; proc means data=pair ; var chaos; run;

### Output

- 3. Test the claim that  $\mu_d \neq 0$  at the  $\alpha = 0.01$  level of significance.
- 4. Compute the 99% confidence Interval about the population mean difference  $\mu_d = \mu_1 \mu_2$ .

### Welch's t test(Independent Samples).

**Definition:-** The sampling method is **independent** when the individuals selected in one sample do not dictate which individuals are to be in the second sample. Assumptions:-

1. The sample is obtained using simple random sampling. 2. the samples are independent.

3. the population from which the samples are drawn are normally distributed or the sample sizes are greater than 30.

- **Step 1** Figuring out the **Null** and **Alternative** Hypothesis.
- **Step 2** Write down  $\alpha$  value given in the problem and find the Critical Value  $t_{tab}$ ,  $t_{\alpha}$  or  $t_{\frac{\alpha}{2}}$  and the degrees of freedom will be smaller of  $n_1 - 1$  or  $n_2 - 1$ . Also Draw the rejection region or the critical region.

Step 3 Compute the test Statistic

$$t = t_{Cal} = \frac{(\bar{x_1} - \bar{x_2}) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

**Step 4** Compare the critical value with the test Statistic. In other words compare  $t_{Cal}$  with  $t_{tab}$ .

Step 5 State the conclusion.

# Improving the Conservative df of Welch's t test

The Welch's t test is conservative, in order to get accurate answer we should use

$$df = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\frac{\left(\frac{s_1^2}{n_1}\right)^2}{n_1 - 1} + \frac{\left(\frac{s_2^2}{n_2}\right)^2}{n_2 - 2}}$$

Software will use this particular formula for more accurate answer.

# Confidence Intervals about the difference of two means

A  $(1 - \alpha).100\%$  confidence interval for  $\mu_d$  is given by

$$(\bar{x_1} - \bar{x_2}) \pm t_{\frac{\alpha}{2}} * \sqrt{\frac{{s_1}^2}{n_1} + \frac{{s_2}^2}{n_2}}.$$

where  $t_{\frac{\alpha}{2}}$  is obtained using smaller of  $n_1 - 1$  or  $n_2 - 1$  degrees of freedom or the improved formula.

**Problem 9, Sec 10.2:-** An engineer wanted to know whether the strength of two different concentrate mix designs differed significantly. He randomly selected 9 cylinders, measuring 6 inches in diameter and 12 inches in height, into which mixture 67-0-301 was poured. After 28 days, he measured the strength(in pounds per square inch) of the cylinder. He also randomly selected 10 cylinders of mixture 67-0-400 and performed the same test. the results are as follows:-

Mixture 67-0-301	Mixture 67-0-400
3960	4070
4090	4890
3100	5020
3830	4330
3200	4640
3780	5220
4080	4190
4040	3730
2940	4120
	4620

- 1. The data are obtained from simple random sample that are independent.
- 2. test the claim that mixture 67-0-400 is stronger than 67-0-301 at  $\alpha = 0.05$  level of significance.
Ans:-

Step 1

 $H_0: \mu_{400} = \mu_{301}$  $H_1: \mu_{400} > \mu_{301}$ 

 $\mu_{400}$  = Mean strength of mixture 67-0-400 .  $\mu_{400}$  = Mean strength of mixture 67-0-301 .

**Step 2** This is a right tailed test. df = min{9,8}.  $t_{cri} = 1.860$ . Hence Reject  $H_0$  if  $t_{tab} > 1.860$ .

Step 3 
$$\bar{x}_{400} = 4483, n_{400} = 10$$
  
 $\bar{x}_{301} = 3669, n_{301} = 9$   
 $t = t_{Cal} = \frac{(\bar{x}_{400} - \bar{x}_{301}) - (\mu_{400} - \mu_{301})}{\sqrt{\frac{s_{400}^2}{n_{400}} + \frac{s_{301}^2}{n_{301}}}} = 3.8$ 

**Step 4** As 3.8 > 1.860, we reject  $H_0$ .

Step 5 Hence there is enough evidence to conclude that the mixture 67-0-400 is stronger than 67-0-301. 3. Construct a 90% confidence interval about  $\mu_{400} - \mu_{301}$ , and interpret the result.

**Ans:-** The confidence interval is given by

$$(\bar{x}_{400} - \bar{x}_{301}) \pm t_{\frac{\alpha}{2}} * \sqrt{\frac{s_{400}^2}{n_{400}} + \frac{s_{301}^2}{n_{301}}}.$$
  
= 814 ± 398  
= (416, 1212)

•

We are 90% confident that the true mean

difference is between 416 and 1212.

#### Pooled Two sample t test.

Pooled t-test is used for independent sample case and it assumes equal population variance. While Welch's t-test doesn't assume any thing about the population variance. Comparison between Welch's t-test and Pooled t-test,

Pooled t-test
Assumes that population variance is same
$df = n_1 + n_2 - 2$
$t = t_{Cal} = \frac{(\bar{x_1} - \bar{x_2}) - (\mu_1 - \mu_2)}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$
where $s_p = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}}$
Not conservative as compared to Welch's t-test
P(Type II Error) is normally not so high as
Welch's t-test
Disadvantage is how to find out the fact that the
population variance are the same? The answer is F-Test which
we will learn later in this section.

#### Problems of Pooled t-test

The following random samples are measurements of the heat-producing capacity (in millions per calories of tons) of coal from two mines:-

Use the 0.05 level of significance to test wether the difference between the means of these two samples is significant, assuming that  $\sigma_1 = \sigma_2$ .

Ans:-

Step 1

$$H_0: \mu_1 = \mu_2$$
$$H_1: \mu_1 \neq \mu_2$$

Step 2  $\alpha = 0.05, df = n_1 + n_2 - 2 = 8$ , Reject  $H_0$  if  $|t_{cal}| > t_{cri} = 2.306$ .

**Step 3**  $\bar{x}_1 = 8178, s_1 = 271.1, \bar{x}_2 = 7788, s_2 = 216.8.$ 

$$s_p = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}} = 245.5$$

$$t_{Cal} = \frac{(\bar{x_1} - \bar{x_2})}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} = 2.51$$

- **Step 4** As 2.51 > 2.306, we Reject the Null Hypothesis.
- Step 5 Hence there is enough evidence to conclude that the difference between the two sample mean is significant.

#### Inference about Two Sample Population Standard Deviation

# When to decide that the population Variances are Equal?

In order to answer this question we need the following assumption:-

- The samples are independent simple random samples.
- The population from which the samples are drawn must be normally distributed.

#### Introduction of Fisher's F-distribution

If  $\sigma_1^2 = \sigma_2^2$  and  $s_1^2$  and  $s_2^2$  are sample variances from independent simple random samples of size  $n_1$  and  $n_2$ respectively, drawn from normal populations, then

$$F = \frac{s_1^2}{s_2^2}$$

follows F-Distribution with  $n_1 - 1$  numerator degrees of freedom and  $n_2 - 1$  denominator degrees of freedom. About critical values of F-Distribution:-

#### Steps for Hypothesis Tests on Two Population Standard Deviation

- **Step 1** Figuring out the **Null** and **Alternative** Hypothesis.
- **Step 2** Write down  $\alpha$  value given in the problem and find the Critical Value  $F_{cri}$  as follows:-
  - Two Tailed  $F_{cri} = F_{1-\frac{\alpha}{2},n_1-1,n_2-1}$  and  $F_{cri} = F_{\frac{\alpha}{2},n_1-1,n_2-1}$ .

• Left Tailed  $F_{cri} = F_{1-\alpha,n_1-1,n_2-1}$ .

• Right Tailed  $F_{cri} = F_{\alpha,n_1-1,n_2-1}$ . Step 3

$$F_{cal} = \frac{s_1^2}{s_2^2}$$

**Step 4**  $F_{cal}$  lies in the critical region or not? **Step 5** State the conclusion. Calculate  $F_{0.975,12,14}$ 

#### Problem 11, Sec 10.4:-

	Sample for population 1	Sample for Population 2
n	26	19
S	9.9	6.4

Test the claim that  $\sigma_1 > \sigma_2$  at  $\alpha = 0.01$  level of significance for the given sample data.

#### Ans:-

Step 1

$$H_0: \sigma_1 = \sigma_2$$
$$H_1: \sigma_1 > \sigma_2$$

**Step 2**  $\alpha = 0.01$ . Right tailed test. Therefore

 $F_{cri} = F_{0.01,26-1,19-1} = F_{0.01,25,18} = 2.84$ . Reject  $H_0$  if  $F_{cal} > 2.84$ .

Step 3

$$F_{cal} = \frac{s_1^2}{s_2^2} = \frac{9.9^2}{6.4^2} = 2.39$$

- **Step 4** As  $F_{cal}$  is not in the critical region , we do not reject  $H_0$ .
- **Step 5** There is not enough evidence to support the claim that  $\sigma_1 > \sigma_2$ .

## $\chi^2$ Goodness of fit Test

**Definition:** – A **goodness-of-fit test** is an inferential procedure used to determine whether a frequency distribution follows a claimed distribution.

In other words A statistical test in which the validity of one hypothesis is tested without specification of an alternative hypothesis is called a **goodness-of-fit test.**The idea behind the chi-square goodness-of-fit test is to see if the sample comes from the population with the claimed distribution. Another way of looking at that is to ask if the frequency distribution fits a **specific pattern**.

Two values are involved, an <u>observed value</u>, which is the frequency of a <u>category from a sample</u>, and the other is <u>expected frequency</u>, which is calculated based upon the <u>claimed distribution</u>. (Sometimes know as expected counts  $E_i = np_i$ )

The idea is that if the observed frequency is really close to the claimed (expected) frequency, then the square of the deviations will be small. The square of the deviation is divided by the expected frequency to weight frequencies. The test statistics

$$\chi^{2} = \sum \frac{(O_{i} - E_{i})^{2}}{E_{i}} \qquad i = 1, 2, ..., k$$

approximately follows a Chi-Square Distribution with n-1 degrees of freedom if the following assumptions are met.

- The data are obtained from a random sample.
- All expected frequency must be at least 1.
- At most 20% of the expected frequencies are less than 5.

## The following are properties of the goodness-of-fit test

- 1. The data are the observed frequencies. This means that there is only one data value for each category.
- 2. The degrees of freedom is one less than the number of categories, not one less than the sample size.
- 3. It is always a **right tail test**.
- 4. It has a chi-square distribution. The value of the test statistic doesn't change if the order of the categories is switched.

#### Steps for Testing A Claim Using a Goodness-of-Fit Test:-

- **Step 1** A claim is made regarding a distribution. The claim is used to determine Null and Alternative Hypothesis.  $H_0$ : The random variable follows the claimed distribution.
- Step 2 Calculate the Expected frequency for each of the K categories.
- **Step 3** Verify the assumptions for goodness -of- fit Test.
- **Step 4** Compute the test Statistic

$$\chi^2_{Cal} = \sum \frac{(O_i - E_i)^2}{E_i}.$$

- **Step 5** Find  $\chi^2_{cri} = \chi^2_{\alpha}$  with k 1 degrees of freedom.
- **Step 6** Reject  $H_0$  if  $\chi^2_{cal} > \chi^2_{cri}$ .
- Step 7 State the conclusion.

**Bicycle Deaths:-** A researcher wanted to determine whether bicycle deaths were uniformly distributed over the days of the week. She **randomly** selected 200 deaths that involved a bicycle, recorded the day of the week on which the day occurred, and obtained the following results:

Day of week	Frequency
Sunday	16
Monday	35
Tuesday	16
Wednesday	28
Thursday	34
Friday	41
Saturday	30

Is there reason to believe that the day of the week on which a fatality occurs on a bicycle occurs with equal frequency at the  $\alpha = 0.05$  level of significance?

#### Ans:-Step 1

$$H_0: p_1 = p_2 = p_3 = p_4 = p_5 = p_6 = p_7 = \frac{1}{7}$$
  
 $H_0:$  at least one of the  $p_i$  will is different from  $\frac{1}{7}$ 

Step 2,3,4 will follow from the following diagram.

Observed $\operatorname{Count}(O_i)$	Expected $\operatorname{Count}(E_i)$	$(O_i - E_i)^2$	$\frac{(O_i - E_i)^2}{E_i}$
16	$200(\frac{1}{7}) = 28.57$		
35	28.57		
16	28.57		
28	28.57		
34	28.57		
41	28.57		
30	28.57		
$\overline{\text{Step 5 } df = 7 - 1} =$	$6, \chi^2_{\ 0.05} = 12.592$		

## Problems of $\chi^2$ Goodness of fit Test

1.  $H_0$ : The random variable X is **binomial with** n = 4, p = 0.3.

 $H_1$ : The random variable X is not binomial with n = 4, p = 0.3.

Х	0	1	2	4	5
Observed	260	400	280	50	10

- Find the Expected Frequency.
- Determine the  $\chi^2$  test Statistic.
- Determine the degrees of freedom.
- Test the hypothesis at the  $\alpha = 0.05$  level of significance.

### Problems of $\chi^2$ Goodness of fit Test

**2.**According to the manufacturer of M&Ms, 30% of the plain M&Ms in a bag should be brown, 20% yellow, 20% red, 10% orange, 10% blue, 10% green. A student wanted to determine whether a randomly selected bag of plain M&Ms had contents that followed this distribution. He counted the number of M&Ms that were colored and obtained the following results.

Color	Frequency
Brown	125
Yellow	77
Red	90
Blue	31
Orange	42
Green	35

Test the claim that plain M&Ms follow the distribution stated by manufacturer of M&Ms at 0.05 level of significance.

## Problems of $\chi^2$ Goodness of fit Test

3.	
Number of Radio Messages	Observed Frequency
0	70
1	57
2	46
3	20
4	5
5	2

Number of Radio Messages	Expected Frequency
0	44.6
1	66.9
2	50.2
3	25.1
4 or more	13.1

Note:- The expected frequency table was constructed using Poisson Distribution with  $\lambda=1.5\,.$ 

 $H_0$ : The population sampled has the Poisson distribution with  $\lambda = 1.5$ .

Test the claim at 0.01 level of significance.

## **Curve Fitting**

#### **Model** : $Y_i = \beta_0 + \beta_1 X_i + \epsilon_i$

Where

 $\mathbf{1}$ . $X_i$  is the independent variable (predictor variable)

 $\mathbf{2}.Y_i$  is the dependent variable or (response variable)

**3**.  $\beta_1$  is the slope and  $\beta_0$  is the Y-intercept.

**4**.  $\epsilon_i$  is the Error term which follows  $N(0, \sigma^2)$ .

This model is a population (notice the Greek  $\beta's$  and  $\epsilon$  known as estimators) simple (meaning one X variable) linear (meaning straight line) regression model. If we take the **expected value** (average) of the Regression model it becomes  $E(Y_i) = \beta_0 + \beta_1 X_i$ .

(1) The disappears because  $E(\epsilon_i) = 0$ , that is we expect our error to be zero.

(2) This model indicates that given a level of  $X_i$ , you can generate the mean of the probability distribution of the  $Y'_i$ s at that  $X_i$  level.

We must estimate the population Regression line using sample data.

The equation of the Least-squares Regression Line is given by

$$\hat{Y}_i = b_1 x + b_0$$

 $b_1$  and  $b_0$  are the estimates for  $\beta_1$  and  $\beta_0$  respectively. We can calculate  $b_1$  and  $b_0$  from Normal equations. ("y-hat" is the Predicted Value for any given x.)

#### **Normal Equations**

$$\sum y = nb_0 + b_1 \sum x$$
$$\sum xy = b_0 \sum x + b_1 \sum x^2$$

where  $\Sigma$  means summing over all the data points. Solving the above Normal equations we obtained  $b_1$  and  $b_0$  as follows.

$$b_1 = \frac{\sum xy - \frac{1}{n}(\sum x)(\sum y)}{\sum x^2 - \frac{1}{n}(\sum x)^2} = \frac{S_{xy}}{S_{xx}}$$
$$b_0 = \frac{\sum y - b_1(\sum x)}{n}$$

Where

$$S_{xx} = \sum x^2 - \frac{1}{n} (\sum x)^2$$

is the variance of observed  $x'_i s$  multiplied by the factor n-1.

We will use similar notations later on like

$$S_{yy} = \sum y^2 - \frac{1}{n} (\sum y)^2$$

is the variance of observed  $y'_i s$  multiplied by the factor n-1.

A few more Notations...

Standard Deviation of  $x'_i s$  is denoted by  $s_x = \sqrt{\frac{S_{xx}}{n-1}}$ .

Standard Deviation of  $y'_i s$  is denoted by  $s_y = \sqrt{\frac{S_{yy}}{n-1}}$ .

#### What are the steps we require to fit a Least Square Line from the given Data

- 1. Plot the data. Equivalently draw the **Scatter Diagram.**
- 2. Write down the Population Simple Linear Regression Model.  $(y = \beta_0 + \beta_1 x + \epsilon).$
- 3. Two Normal equations.
- 4. Solve Normal Equations to find  $b_1$  and then  $b_0$ . Plug in back  $b_1$  and  $b_0$  in the Model.
- 5. Fit the Least Square Line in the Scatter Diagram.

Number of weeks	Hearing Range
x	У
47	15.1
56	14.1
116	13.2
178	12.7
19	14.6
75	13.8
160	11.9
31	14.8
12	15.3
164	12.6
43	14.7
74	14.0

#### **Problems on Curve Fitting**

Here x is the length of time that a person has been living near a major airport directly in the flight path of departing jets, and y is his or her hearing range(in thousands of cycles per second)

**Solution:-** Solved in the class.

#### Multiple regression

**Model:-**  $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \epsilon$ 

The least square multiple regression line is

$$\hat{Y} = b_0 + b_1 X_1 + b_2 X_2$$

The normal equations from which  $b_0$ ,  $b_1$ ,  $b_2$  are obtained are,

$$\sum y = nb_0 + b_1 \sum x_1 + b_2 \sum x_2$$
  
 $\sum yx_1 = b_0 \sum x_1 + b_1 \sum x_1^2 + b_2 \sum x_1 x_2$   
 $\sum yx_2 = b_0 \sum x_2 + b_1 \sum x_1 x_2 + b_2 \sum x_2^2$ 

## Important Formulas

$$S_{xx} = \sum x^2 - \frac{1}{n} (\sum x)^2$$

$$S_{yy} = \sum y^2 - \frac{1}{n} (\sum y)^2$$

$$S_{xy} = \sum xy - \frac{1}{n} (\sum x) (\sum y)$$

$$s_e = \sqrt{\frac{SSE}{n-2}} = \sqrt{\frac{S_{yy} - b_1 S_{xy}}{n-2}}$$

$$s_{b_1} = \frac{s_e}{\sqrt{S_{xx}}}$$

$$b_1 = \frac{S_{xy}}{S_{xx}}$$

$$r = \frac{S_{xy}}{\sqrt{S_{xx}}S_{yy}}$$

$$SST = S_{yy}$$

$$SSE = S_{yy} - b_1 S_{xy}$$

$$SSR = b_1 S_{xy}$$

#### ANALYSIS OF VARIANCE

Motivation:- Suppose we are asked to compare the cleansing action on the basis of following whiteness reading of three Detergents on 15 dinner plates. The data is listed below.

Detergent X	77	81	71	76	80
Detergent Y	72	58	74	66	70
Detergent Z	76	85	82	80	77

Other factors at the time of washing the dinner plates :-

- how dirty the plates were?
- washing time?
- water temperature and hardness while cleaning.
- what kind of instrument used for washing?

If we calculate the means of whiteness reading we will have 77 for detergent X, 68 for detergent Y, 80 for detergent Z.

Remember that a significant test may show that the differences between sample means are too large to be attributed to chance,

# but cannot say why the differences occurred?

Hence we should introduce a Controlled Experiment procedure , where every above factors remain the same during the washing process. That means same washing time, same dirty ness, use water of exactly the same temperature and hardness , inspecting the instrument after each use and so many other factors... . Isn't the above Controlled Experiment tedious?

#### Ans:-

So what we do to get rid of this Controlled Factors? We can conduct an experiment in which none of these factors is controlled, but in which we protect ourselves against their effect by **RANDOMIZATION**. That is we design, or plan, the experiment in such a way that the variation caused by these factors can be combined under the general heading of "Chance".

#### Mathematically what we do?

**Def:-** An **analysis of variance** expresses a measure of the total variation in a set of data as a sum of terms, each attributed to a specific source, or cause, of variation.

#### The Assumptions of Analysis of Variance

- 1. Treatment effects are additive.
- 2. Experimental errors are random.
- 3. Experimental errors are independently distributed.
- 4. Experimental errors follow a normal distribution.
- 5. Experimental errors have mean zero and constant variance.

Source					
Treatment 1	$x_{11}$	$x_{12}$	$x_{13}$	•••	$x_{1n}$
Treatment 2	$x_{21}$	$x_{22}$	$x_{23}$	•••	$x_{2n}$
•••	•••			•••	•••
Treatment $K$	$x_{K1}$	$x_{K2}$	$x_{K3}$	•••	$x_{Kn}$

#### notations:-

Sample in i<sup>th</sup> row and j<sup>th</sup> coloumn = 
$$x_{ij}$$
  
Treatment K variance =  $s_i^2$   
Grand Mean =  $\bar{x} = \frac{x_{11} + x_{12} + \dots + x_{Kn}}{Kn}$   
Mean Treatment 1 =  $\bar{x}_1 = \frac{x_{11} + x_{12} + \dots + x_{1n}}{n}$   
Mean Treatment 2 =  $\bar{x}_2 = \frac{x_{21} + x_{22} + \dots + x_{2n}}{n}$ 

•

•

Mean Treatment 
$$K = \bar{x}_k = \frac{x_{k1} + x_{k2} + \dots + x_{kn}}{n}$$

$$SST = SS(Tr) + SSE$$

Equivalently we can re-write the above equations;

SST = SSB + SSW

Mathematically

$$SST = \sum \sum (x_{ij} - \bar{x})^2$$
$$SSB = n \sum (\bar{x}_i - \bar{x})^2$$
$$SSW = \sum \sum (x_{ij} - \bar{x}_i)^2 = (n - 1)(s_1^2 + s_2^2 + \dots + s_K^2)$$

**Test**  $H_0: \mu_1 = \mu_2 = \mu_3 = ... = \mu_K$ 

#### AnOVA TABLE

Source	DF	Sum of Squares	Mean Square	F
Treatments	K-1	SSB	$MSB = \frac{SSB}{K-1}$	$F = \frac{MSB}{MSW}$
Error	K(n-1)	SSW	$MSW = \frac{SSW}{K(n-1)}$	
Total	nK-1	SST		

note:- This AnOVA works when each Treatment have same number of observations. The Formulas for SSB, and SSW will slightly change when we have different observations in each treatments.

## COMPUTING FORMULAS FOR SUM OF SQUARES (Equal Sample Size)

$$T = x_{11} + x_{12} + \dots + x_{Kn}$$
$$T_i = x_{i1} + x_{i2} + \dots + x_{in}$$

$$SST = \sum \sum x_{ij}^2 - \frac{1}{Kn}T^2$$
$$SSB = \frac{1}{n}\sum T_i^2 - \frac{1}{Kn}T^2$$
$$SSW = SST - SSB$$

## COMPUTING FORMULAS FOR SUM OF SQUARES (UnEqual Sample Size)

Source					
Treatment 1	$x_{11}$	$x_{12}$	$x_{13}$	•••	$x_{1n_1}$
Treatment 2	$x_{21}$	$x_{22}$	$x_{23}$	•••	$x_{2n_2}$
				•••	•••
Treatment $K$	$x_{K1}$	$x_{K2}$	$x_{K3}$	•••	$x_{Kn_K}$

$$T = x_{11} + x_{12} + \dots + x_{Kn_K}$$
$$T_i = x_{i1} + x_{i2} + \dots + x_{in_1}$$
$$N = n_1 + n_2 + \dots + n_K$$

$$SST = \sum \sum x_{ij}^2 - \frac{1}{N}T^2$$
$$SSB = \sum \frac{T_i^2}{n_i} - \frac{1}{N}T^2$$
$$SSW = SST - SSB$$

## WILCOXON SIGNED RANK TEST FOR DEPENDENT SAMPLES (Small $(n \le 30)$ )

Null Hypothesis	$H_0: M_D = 0$	$H_0: M_D = 0$	$H_0: M_D = 0$
Alternate Hypothesis	$H_0: M_D \neq 0$	$H_0: M_D > 0$	$H_0: M_D < 0$
Test Statistic	$\operatorname{Smaller}(T, T_+)$	$T_{-}$	$T_+$
Rejection Region	$T \leq T_0$	$T_{-} \leq T_{0}$	$T_+ \le T_0$

- 1.  $T_{-} =$  Sum of Ranks of all negative differences.
- 2.  $T_+ =$  Sum of Ranks of all positive differences.
- 3.  $T_0 =$ Critical Value.

#### WILCOXON SIGNED RANK TEST FOR DEPENDENT SAMPLES (Large(n > 30))

Null Hypothesis	$H_0: M_D = 0$	$H_0: M_D = 0$	$H_0: M_D = 0$
Alternate Hypothesis	$H_0: M_D \neq 0$	$H_0: M_D > 0$	$H_0: M_D < 0$
Test Statistic	$z_{cal}$	$z_{cal}$	$z_{cal}$
Rejection Region	$ z  > z_{\frac{\alpha}{2}}$	$z < -z_{\alpha}$	$z > z_{\alpha}$

Where  $T_+$  is sum of Ranks of all positive differences.

$$z = \frac{T_{+} - \frac{n(n+1)}{4}}{\sqrt{\frac{n(n+1)(2n+1)}{24}}}$$

## Problem WILCOXON SIGNED RANK TEST FOR DEPENDENT SAMPLES (Small $(n \le 30)$ )

Α	В	A-B	A - B	Dummy	Ranking	Ranking
7	9					
4	5					
8	8					
9	8					
3	6					
6	10					
8	9					
10	8					
9	4					
5	9					
COMPARING TWO POPULATION INDEPENDENT SAMPLES or WILCOXON RANK SUM TEST (SMALL $(n_1 \le 20, n_2 \le 20)$ )

Location 1:- 0.37 0.70 0.75 0.30 0.45 0.16 0.62 0.73 0.33 Location 2:- 0.86 0.55 0.80 0.42 0.97 0.84 0.24 0.51 x

Sorting	$\mathbf{the}$	numbers	in	ascending	order	and
		Ra	nki	ing		

Data	Location	Rank
0.16	1	1
0.24	2	2
0.30	1	3
0.33	1	4
0.37	1	<b>5</b>
0.42	2	6
0.45	1	7
0.51	2	8
0.55	2	9
0.62	1	10
0.69	2	11
0.70	1	12
0.73	1	13
0.75	1	14
0.80	2	15
0.84	2	16
0.86	2	17
0.92	2	18
0.97	2	19

 $T_1 = 1 + 3 + 4 + 5 + 7 + 10 + 12 + 13 + 14 = 69$ 

$$T_2 = 2 + 6 + 8 + 9 + 11 + 15 + 16 + 17 + 18 + 19 = 121 = \frac{19 * 20}{2} - 69$$

Null Hypothesis	$H_0: M_1 = M_2$	$H_0: M_1 = M_2$	$H_0: M_1 = M_2$
Alternate Hypothesis	$H_0: M_1 \neq M_2$	$H_0: M_1 > M_2$	$H_0: M_1 < M_2$
N0.	Test		Statistic
$n_1 < n_2$	$T_1$	$T_1$	$T_1$
$n_1 > n_2$	$T_2$	$T_2$	$T_2$
$n_1 = n_2$	Either one	Either one	Either one

**Rejection Region** 

Problem:

The following are the burning times of random samples of two kinds of emergency flares:

Brand 1	Brand 2
17.2	13.6
18.1	19.1
19.3	11.8
21.1	14.6
14.4	14.3
13.7	22.5
18.8	12.3
15.2	13.5
20.3	10.9
17.5	14.8

Use the Wilcoxon Rank Sum Test at 0.05 l.o.s to test whether it is reasonable to say that on the average brand 1 flares are better than Brand 2 flares.

## Summary of Non Parametric Tests for Final Exam

• ONE SAMPLE SIGN TEST FOR A POPULATION MEDIAN (Small  $(n \le 30)$ )

We are testing , Null Hypothesis  $H_0: M = 0, M$  is the population Median.

• WILCOXON SIGNED RANK TEST FOR DEPENDENT SAMPLES (Small  $(n \le 30)$ )

We are testing, Null Hypothesis  $H_0: M_D = 0, M_D$  is the population Median difference.

- WILCOXON SIGNED RANK TEST FOR DEPEN-DENT SAMPLES (Large (n > 30)) We are testing, Null Hypothesis  $H_0: M_D = 0$
- COMPARING TWO POPULATION INDEPEN-DENT SAMPLES or WILCOXON RANK SUM TEST (SMALL( $n_1 \le 20$ ,  $n_2 \le 20$ )) We are testing, Null Hypothesis  $H_0: M_1 = M_2$  $M_1$  and  $M_2$  are Population median for sample 1 and sample 2 respectively.
- KRUSKAL-WALLIS H-TEST FOR COMPLETELY RANDOMIZED DESIGN (ONE WAY ANOVA) We are testing, Null Hypothesis  $H_0: M_1 = M_2 = M_3 = \dots = M_k$

• FRIEDMAN F<sub>r</sub>-TEST FOR COMPLETELY RAN-DOMIZED BLOCK DESIGN (TWO WAY ANOVA)

We are testing , Null Hypothesis  $H_0$  : Effects of Each Treatment = 0, Effect of Each Block = 0

## • SPEARMAN'S RANK CORRELATION COEFFI-CIENT

We are testing, Null Hypothesis  $H_0: \rho = 0$ , where  $\rho$  is the population correlation between x and y.