

# **30 Hours Online Certificate Course On “*Research & Data Analysis*”**

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# Hypothesis Testing

- Hypothesis: Statements or Assumptions or Claims which can be Verified.
- Null Hypothesis ( $H_0$ ): No Effect/Impact/Relationship between Variables.
- Alternate or Research Hypothesis ( $H_a$ ): It contains the prediction/hypothesis from our Theory. Presence of some Effect/Impact/Relationship between Variables.
- We always test Null but want Alternate to be true & null to be not accepted.

# One Tail and Two Tail Tests

- A statistical model that tests a directional hypothesis is called a one-tailed test, whereas one testing a non-directional hypothesis is known as a two-tailed test.

## Summary of One- and Two-Tail Tests...

One-Tail Test  
(left tail)

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu < \mu_0$$



Two-Tail Test

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu \neq \mu_0$$



One-Tail Test  
(right tail)

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu > \mu_0$$



# Test Statistics

- We look at a simple signal-to-noise ratio: we compare how good the model/hypothesis is against how bad it is (error).

- $$\text{Test statistic} = \frac{\text{signal}}{\text{noise}} = \frac{\text{variance explained by the model}}{\text{variance not explained by the model}} = \frac{\text{effect}}{\text{error}}$$

- t, F, and Chi Square ( $\chi^2$ ) are three very commonly used and popular test Statistics.
- The exact form of this equation changes depending on which test statistic you're calculating, but they all, broadly speaking, represent the same thing: signal-to-noise.

# Hypothesis Testing Approaches

- To test null hypothesis, we use sample information to obtain the test statistic.
- Mostly, this test statistic turns out to be the point estimator of the unknown population parameter.
- Then, we try to find out the sampling, or probability, distribution of test statistic.
- Finally, we use one of the following two approaches to test the null hypothesis:
  - ✓ Confidence Interval Approach; or
  - ✓ Test of Significance (P Value) Approach.



# Test of Significance Approach

- Instead of preselecting  $\alpha$  at some arbitrary levels (like 1, 5, or 10%), we can obtain the p (probability) value, or the exact level of significance of a test statistic.
- P value is lowest significance level at which a  $H_0$  can be rejected. It is probability of finding observed, or more extreme value of a test statistic when ( $H_0$ ) is true. If P-Value is less than given  $\alpha$ , we have significant effect/relationship,  $H_0$  is rejected.
- Larger Test Statistic > Smaller P value.

# Test of Significance Approach

- If the p-value is very small (usually  $< .05$ ), we conclude that the model (Alternate Hypothesis) fits the data well, & we assume our initial prediction is true, i.e., we gain confidence in alternative hypothesis ( $H_a$ ).

Situations	✓ Always Say	X Never Say	Real Meaning
If P-Value $>$ 0.05 (5%), or 0.01 (1%), or 0.10 (10%)	Null Hypothesis ( $H_0$ ) can not be rejected or We don't have enough evidence to reject $H_0$ .	<ul style="list-style-type: none"><li>➤ Accept <math>H_0</math></li><li>➤ Reject <math>H_a</math></li><li>➤ <math>H_0</math> is True</li><li>➤ <math>H_a</math> is False</li></ul>	Little statistical evidence/support for our Alternate/ Research Hypothesis
If P-Value $<$ 0.05 (5%), or 0.01 (1%), or 0.10 (10%)	Null Hypothesis ( $H_0$ ) can be rejected or We have enough evidence to reject $H_0$ .	<ul style="list-style-type: none"><li>➤ Accept <math>H_a</math></li><li>➤ <math>H_0</math> is False</li><li>➤ <math>H_a</math> is True</li></ul>	Strong statistical evidence/support for our Alternate/ Research Hypothesis

# Sample Size & Statistical Significance

- The significance of a test statistic is directly linked to the sample size.
- The same effect will have different p-values in different sized samples.
- Small differences can be deemed 'Significant' in Large Samples; and
- Large Effects might be deemed 'Non Significant' in Small Samples.
- This is also called Large Sample Bias.
- So for Significant results, get Large Samples.



# Scales of Measurement - Permissible Statistics & Tests

Scale/Level of Measurement	Permissible Statistics	Permissible Tests
Binary	Frequencies, Percentages, Mode	Binomial, Chi-Square
Nominal	Frequencies, Percentages, Mode	Binomial, Chi-Square
Ordinal	Nominal + Median, Percentiles, Quartiles, Deciles, Rank Correlation	Binomial, Chi-Square, Non Parametric Tests
Interval	Ordinal + Mean, Range, Standard Deviation, Regression	Binomial, Chi-Square, t, Z, F
Ratio	All Under Sun	All Known to Mankind

# Tests for Normality: K-S & Shapiro-Wilk Tests

- Both Kolmogorov-Smirnov (K-S) test and Shapiro-Wilk test compare scores in sample to a normally distributed set of scores with the same mean and standard deviation.
- If the test is non-significant ( $p > .05$ ) it tells us that the distribution of the sample is not significantly different from a normal distribution (i.e., it is probably normal).
- If, the test is significant ( $p < .05$ ) then the distribution is significantly different from a normal distribution (i.e., it is non-normal).

# Tests for Normality: K-S & Shapiro-Wilk Tests

- Shapiro-Wilk test has more power to detect differences from normality (so it might be significant when the K-S test is not).
- In large samples, these tests can be significant even when the scores are only slightly different from a normal distribution.
- So, they are not particularly recommended and they should always be interpreted in conjunction with Histograms, P-P or Q-Q Plots, and values of Skewness & Kurtosis.

# Tests for Normality: K-S & Shapiro-Wilk Tests

- Performing Kolmogorov-Smirnov (K-S) & Shapiro-Wilk Tests in IBM SPSS Statistics to detect Non Normality:
- ✓ Analyze > Descriptive Stats. > Explore;
- ✓ Enter one or more Scale Variables (Percent);
- ✓ Select “Normality Plots with Tests” in Plots.
- ✓ Ho: Distribution of the Variable is not Significantly Different from a Normal Distribution;
- ✓ Interpret as per Significance/P-Values.

# **Analysis of Categorical Data - Cross Tabulation & Chi-Square Analysis**



# Crosstabs

- Type of table in matrix format that displays the (multivariate) frequency distribution of the variables.
- Heavily used in survey, social science and business research.

Event forecast	Event observed		
	Yes	No	Marginal total
Yes	a	b	$a + b$
No	c	d	$c + d$
Marginal total	$a + c$	$b + d$	$a + b + c + d = n$

# Chi Square Test of Independence

- Used for 2 Categorical variables (Nominal or Ordinal Scale) with 2 or more sub categories.
- Basically, compares Observed and Expected Frequencies & computes Chi Square Statistic.
- In Crosstabs, it tests the Hypothesis that Row & Column Variables are Independent.
- Interpretation: If Asymptotic Significance (2-sided) of Pearson Chi Square & Likelihood Ratio is less than 0.05 (5%), we reject the null hypothesis that the two variables are independent. We gain evidence that the two variables are significantly related to each other.

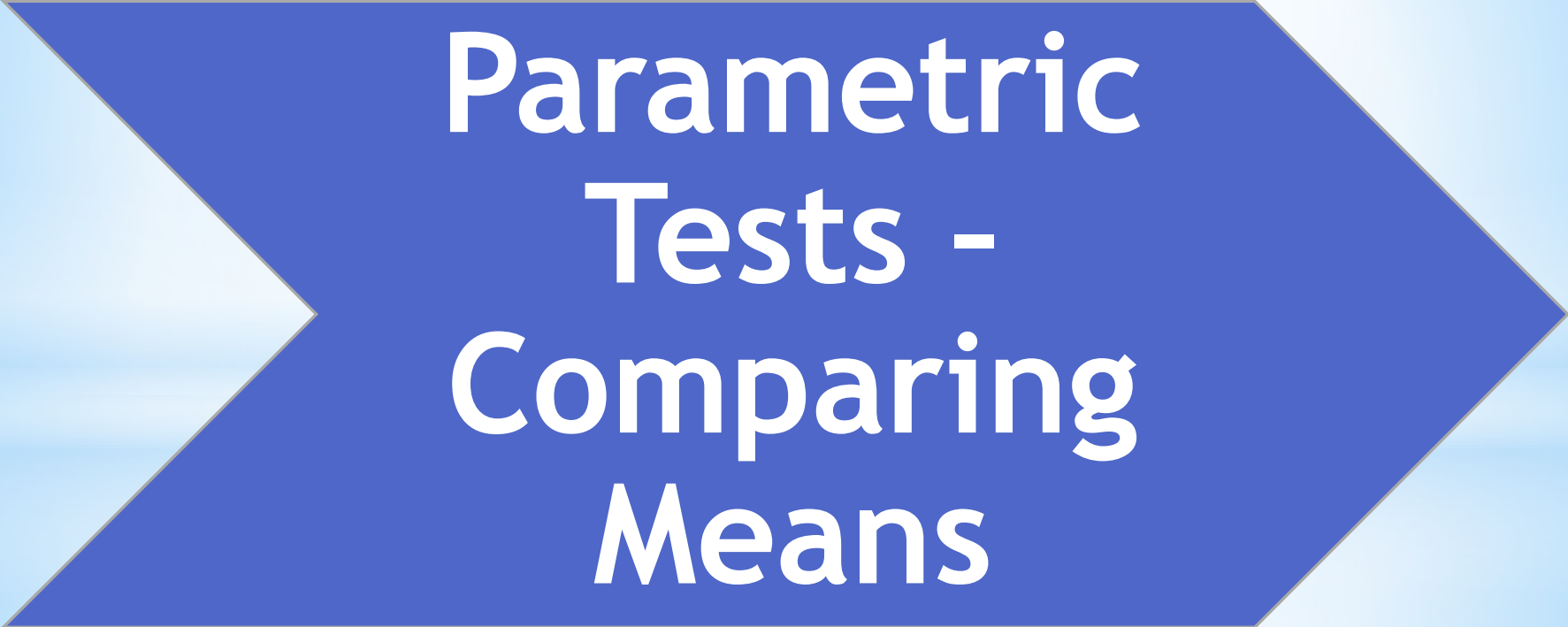
# SPSS: Crosstabs & Chi Square Test of Independence

## ➤ Crosstabs (*Gender - Ethnicity*)

Analyze > Descriptive Statistics > Crosstabs

### ❖ Steps:

- ✓ Select & Enter Row & Column Variables.
- ✓ In Statistics, Select Chi Square.
- ✓ In Cells, Select Percentages (Total).



# Parametric Tests - Comparing Means

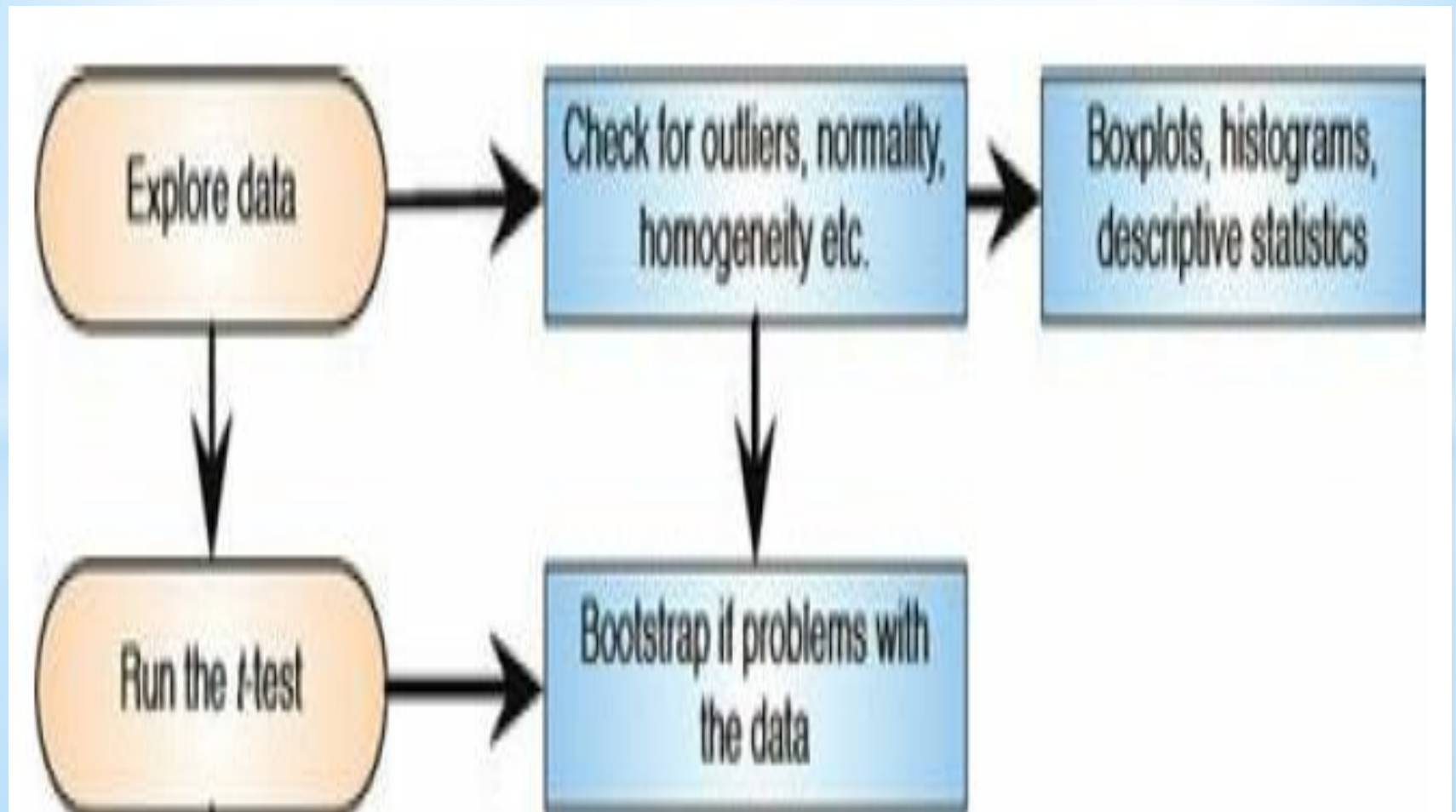
# Comparing Means - “t” Tests

- ❖ When to use “t” tests?
  - When Population Std. Dev. not known;
  - Sample size less than 30. For sample size more than 30 (it resembles Normal distribution).
- ❖ t tests:
  - One Sample t Test
  - Independent Samples t Test
  - Paired Samples t Test
- ❖ Assumptions of All t Tests & ANOVA - Lack of Outliers, Normality of Data, Homogeneity of Variance. If there is any doubt in meeting any of these assumptions - use Bootstrap.



# Comparing Means - “t” Tests

## ➤ The General Process of “t Tests”



# Comparing Means - One-Sample t Test

- Tests whether the mean of a single variable differs from a specified constant (e.g. 0 or 2.5 on a 5 point Likert Scale).
- SPSS:
  - ✓ Analyze > Compare Means > One-Sample T Test
  - ✓ Select one or more variables to be tested against same hypothesized value (Quiz 1-5).
  - ✓ Enter a numeric test value against which each sample mean is compared (5 & 7.5).
- Reject  $H_0$  (Null Hypothesis) of Mean Value = Constant Value if Significance of t Statistic < 0.05 (5%).

# Comparing Means - Paired Sample t Test

- Compares means of two Quantitative (Interval or Ratio Scale) variables for a single sample.
- It computes the differences between values of the two variables for each case and tests whether the average differs from 0.
- **SPSS:**
  - ✓ Analyze > Compare Means > Paired-Samples T Test.
  - ✓ Then Select 1 or more pairs of Variables (*Quiz 1-2; 1-3; 1-4 & 1-5*).
- Reject  $H_0$  (Null Hypothesis) of No Difference in Mean Values if Significance of t Statistic < 0.05.

# Comparing Means - Independent Sample t Test

- Compares means of a Quantitative Variable (Interval or Ratio Scale) for 2 categories/groups.
- Same as One Way ANOVA, but only for 2 Categories or Groups.
- The procedure uses a grouping variable with two values to separate the cases into two groups.
- The grouping variable can be numeric (values such as 1 and 2) or short string (yes and no).

# Comparing Means - Independent Sample t Test

## ➤ SPSS:

- ✓ Analyze > Compare Means > Independent-Samples T Test.
- ✓ Select one or more quantitative test variables (*Percent and GPA*). A separate t test is computed for each variable.
- ✓ Select a single grouping variable, and then click Define Groups to specify two codes for the groups that you want to compare.  
[Variable: *extcr* - *Did extra credit project?*;  
Groups - 1 (No) & 2 (Yes)].



# Comparing Means - Independent Sample t Test

## ➤ Interpretation:

- ✓ First Check the result of Levene's test. If Sig. of F > 0.05; Null Hypothesis of Equal Variance is Accepted. So, Interpret t test results corresponding to that (the upper one).
- ✓ If Ho of equal variance is rejected, interpret the other t test (the lower one).
- ✓ If Sig. of t < 0.05, the Null Hypothesis of no difference in mean of Quantitative variable across the two groups is rejected.

# Comparing Means - One Way ANOVA

- Calculates & compares subgroup means for dependent variables across categories of an independent variable.
- Dependent variable(s) must be quantitative (interval or ratio scale).
- Independent variable must be categorical having categorical values coded as integers (e.g. 1- Sec. A, 2- Sec. B & 3 - Sec. C).
- Categories can be two or more.
- Null Hypothesis: There is no Significant Difference between Mean Values of the Dependent Variable across the Categories of Independent variable.

# One Way ANOVA - Post Hoc Tests

- After an ANOVA we need further analysis to find out which groups differ called Post Hoc Analysis.
- When you have no specific hypotheses before the experiment, use post hoc tests.
- When you have equal sample sizes and group variances are similar use REGWQ or Tukey.
- If sample sizes are slightly different then use Gabriel's procedure, but if sample sizes are very different use Hochberg's GT2.
- If there is any doubt that group variances are equal then use the Games-Howell procedure.

# SPSS - One Way ANOVA

- From the menus choose:
  - ❖ Analyze > Compare Means > One-Way ANOVA.
  - ❖ Select one or more dependent variables (*Percent and GPA*).
  - ❖ Select a single independent factor variable (*Section and Ethnicity*).
  - ❖ In Options, Select Descriptives within Statistics.
- Reject  $H_0$  (Null Hypothesis) of Equal Means if Significance of F Statistic  $< 0.05$  (5%).

The end!  
Thank you for viewing and  
listening!

