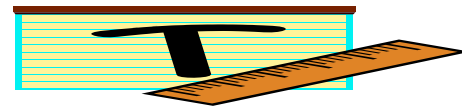


The Scientific Methodologies assignment

- Devise short, multi-item, instruments (knowledge-attitude)
- Item-analysis
- Reliability-validity
- Using statistics “these research tools” to test some assumptions.
- Descriptive statistics
- Normality
- Parametric - Nonparametric Stats
- Correlation-Regression
- T-test – ANOVA-etc.

Scales of Measurement

- **Nominal scale:** A nominal scale is really a list of categories to which objects can be classified.
- **Ordinal scale:** is a measurement scale that assigns values to objects based on their ranking with respect to one another.
- **Interval scale:** one unit on the scale represents the same magnitude on the trait or characteristic being measured across the whole range of the scale. For example, if anxiety were measured on an interval scale, then a difference between a score of 10 and a score of 11 would represent the same difference in anxiety as would a difference between a score of 50 and a score of 51.
- **Ratio scale:** Ratio scales are like interval scales except they have true zero points.



- **Nominal Data**
 - Sex-Nationality-
- **Ordinal Data**
 - ordered but differences between values are not important
 - e.g., political parties on left to right spectrum given labels 0, 1, 2
 - e.g., Likert scales, rank on a scale of 1..5 your degree of satisfaction
 - e.g., restaurant ratings
- **Interval Data**
 - ordered, constant scale, but no natural zero
 - Achievement scale
 - differences make sense, but ratios do not (e.g., $30^{\circ}-20^{\circ}=20^{\circ}-10^{\circ}$, but $20^{\circ}/10^{\circ}$ is not twice as hot!)
 - e.g., temperature (C,F), dates
- **Ratio Data**
 - ordered, constant scale, natural zero
 - e.g., height, weight, age, length

Parametric versus Nonparametric Statistics – When to use them and which is more powerful?

Parametric Assumptions

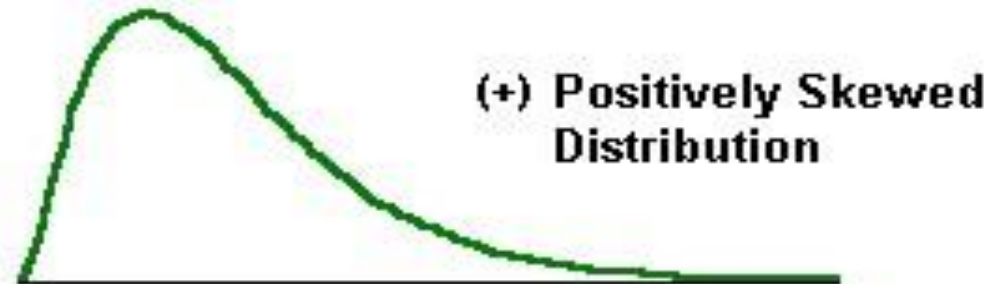
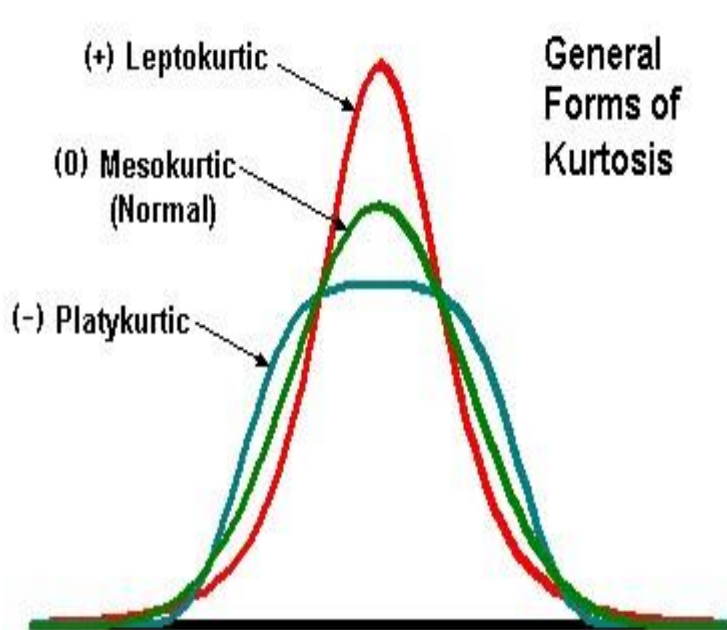
- The observations must be independent (For example participants need to have completed the dependent variable separately, not in groups).
- The observations must be drawn from normally distributed populations
- These populations must have the same variances

- **parametric test**, of course, is a test that **requires a parametric assumption**, such as normality. A **nonparametric test does not rely on parametric assumptions** like normality.
- **a nonparametric test protects against some violations of assumptions and not others.**
The two sample t-test requires three assumptions, normality, equal variances, and independence. The non-parametric alternative, the Mann-Whitney-Wilcoxon test, does not rely on the normality assumption,

Measures of Skewness and Kurtosis

Skewness is a measure of symmetry, or more precisely, the lack of symmetry. A distribution, or data set, is symmetric if it looks the same to the left and right of the center point.

Kurtosis is a measure of whether the data are peaked or flat relative to a normal distribution. Data sets with low kurtosis tend to have a flat top near the mean rather than a sharp peak. A statistical measure used to describe the distribution of observed data around the mean



Differences between independent groups

- Two samples – compare mean value for some variable of interest

Parametric	Nonparametric
t-test for independent samples	Wald-Wolfowitz runs test
	Mann-Whitney U test
	Kolmogorov-Smirnov two sample test

Differences between independent groups

- Multiple groups

Parametric	Nonparametric
Analysis of variance (ANOVA/ MANOVA)	Kruskal-Wallis analysis of ranks
	Median test

Differences between dependent groups

- Compare two variables measured in the same sample
- If more than two variables are measured in same sample

Parametric

t-test for dependent samples

Repeated measures ANOVA

Nonparametric

Sign test

Wilcoxon's matched pairs test

Friedman's two way analysis of variance

Cochran Q

Relationships between variables

- Two variables of interest are categorical

Parametric	Nonparametric
Correlation coefficient	Spearman R
	Kendall Tau
	Coefficient Gamma
	Chi square
	Phi coefficient
	Fisher exact test
	Kendall coefficient of concordance

Summary Table of Statistical Tests

Level of Measurement	Sample Characteristics						Correlation
	1 Sample	2 Sample		K Sample (i.e., >2)			
		Independent	Dependent	Independent	Dependent		
Categorical or Nominal	X2 or bi-nomial	X2	Macnarmar's X2	X2	Cochran's Q		
Rank or Ordinal		Mann Whitney U	Wilcoxin Matched Pairs Signed Ranks	Kruskal Wallis H	Friendman's ANOVA	Spearman's rho	
Parametric (Interval & Ratio)	z test or t test	t test between groups	t test within groups	1 way ANOVA between groups	1 way ANOVA (within or repeated measure)	Pearson's r	
(Plonskey, 2001)							

Parametric correlation

- **Pearson Correlation.** The most widely-used type of correlation coefficient is *Pearson r* (Pearson, 1896), also called *linear* or *product-moment* correlation (the term *correlation* was first used by Galton, 1888). Using non technical language, one can say that the correlation coefficient determines the extent to which values of two variables are "**proportional**" to each other.

Assumptions of the Single Sample T-Test

- **Normality**: Assume that the population is distributed normally, ANOVA is quite robust over moderate violations of this assumption. Check for normality by creating a [histogram](#).
- **Independent Observations**: The observations within each treatment condition must be independent of each other. For example participants need to have completed the dependent variable separately, not in groups.

Paired-Samples t Test

Paired-Samples t Test

- Also known as the t test for dependent means
- Also known as the Dependent t -test

Definitions for Paired-Samples t Test

Definitions of Test

- Sample mean is what researcher will find
 - The value (score) by using statistical analysis (mean)
- Paired sample = paired scores
- Paired = matched (they go together)



Interpreting SPSS Output for t Test

Interpreting Output

Ch. 12 Holcomb Paired-Samples t Test Output – Unformatted

Paired Samples Statistics

		Mean	N	Std. Deviation	Std. Error Mean
Pair 1	Pretest	11.5556	9	2.29734	.76578
	Posttest	9.6667	9	2.23607	.74536

Paired Samples Correlations

		N	Correlation	Sig.
Pair 1	Pretest & Posttest	9	.527	.145

Paired Samples Test

		Paired Differences					t	df	Sig. (2-tailed)
				Std. Error Mean	95% Confidence Interval of the Difference				
					Mean	Std. Deviation			
Pair 1	Pretest - Posttest	1.88889	2.20479	.73493	.19414	3.58364	2.570	8	.033

One-Tailed and Two-Tailed Significance Tests

- When do you use a one-tailed or two-tailed test of significance?
- The answer is that it depends on your hypothesis.
- When your research hypothesis states **the direction** of the difference or relationship, then you use **a one-tailed** probability. For example, a one-tailed test would be used to test these null hypotheses: Females will not score significantly higher than males on an IQ test. In each case, the null hypothesis (indirectly) predicts the direction of the difference.
- **A two-tailed** test would be used to test these null hypotheses (**no direction**): There will be no significant difference in IQ scores between males and females. There will be no significant difference in the amount of product purchased between blue collar and white collar workers.

Test for Significance

- If your statistic is higher than the critical value from the table:
- Your finding is significant.
- You reject the null hypothesis.
- The probability is small that the difference or relationship happened by chance, and p is less than the critical alpha level ($p < \alpha$).
- If your statistic is lower than the critical value from the table:
- Your finding is not significant.
- You fail to reject the null hypothesis.
- The probability is high that the difference or relationship happened by chance, and p is greater than the critical alpha level ($p > \alpha$).

Independent Samples Test

		Levene's Test for Equality of Variances		t-test for Equality of Means						
		F	Sig.	t	df	Sig. (2-tailed)	Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
									Lower	Upper
Number of Older Siblings	Equal variances assumed	1.669	.203	-1.461	44	.151	-.580	.397	-1.381	.220
	Equal variances not assumed			-1.612	31.607	.117	-.580	.360	-1.314	.153

The columns labeled "Levene's Test for Equality of Variances" tell us whether an assumption of the t-test has been met. The t-test assumes that the variability of each group is approximately equal. If that assumption isn't met, then a special form of the t-test should be used. Look at the column labeled "Sig." under the heading "Levene's Test for Equality of Variances". In this example, **the significance (p value) of Levene's test is .203**. If this value is less than or equal to your **α level for the test (usually .05)**, then you can **reject the null hypothesis** that the variability of the two groups is equal, implying that the variances are unequal. then you should use the bottom row of the output (the row labeled "**Equal variances not assumed.**") If the p value is greater than your α level, then you should use the middle row of the output (the row labeled "Equal variances assumed.") In this example, .203 is larger than α , so we will assume that the variances are equal and we will use the middle row of the output.

Independent Samples Test

		Levene's Test for Equality of Variances		t-test for Equality of Means						
		F	Sig.	t	df	Sig. (2-tailed)	Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
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	Equal variances not assumed			-1.612	31.607	.117	-.580	.360	-1.314	.153

The column labeled "Sig. (2-tailed)" gives the two-tailed p value associated with the test. In this example, the p value is .151. the decision rule is given by: **If $p \leq \alpha$** , then reject H0. In this example, .151 is not less than or equal to .05, so we fail to reject H0. That implies that we failed to observe a difference in the number of older siblings between the two sections of this class.

If we were writing this for publication in an APA journal, we would write it as:

A *t* test failed to reveal a statistically reliable difference between the mean number of older siblings that the 10 AM section has ($M = 0.86$, $s = 1.027$) and that the 11 AM section has ($M = 1.44$, $s = 1.318$), $t(44) = 1.461$, $p = .151$, $\alpha = .05$.

Nonparametric Correlations

- The following are three types of commonly used nonparametric correlation coefficients (Spearman R, Kendall Tau, and Gamma coefficients). Note that the chi-square statistic computed for two-way frequency tables, also provides a careful measure of a relation between the two (tabulated) variables, and unlike the correlation measures listed below, it can be used for variables that are measured on a simple nominal scale.

Reading regression output

Model Summary

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate
1	.629 ^a	.395	.345	3.360

a. Predictors: (Constant), words spoken per minute, IQ score, confidence in speaking time 1

ANOVA^b

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	265.441	3	88.480	7.837	.000 ^a
	Residual	406.459	36	11.291		
	Total	671.900	39			

a. Predictors: (Constant), words spoken per minute, IQ score, confidence in speaking time 1

b. Dependent Variable: attitude to school

Coefficients^a

Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.
		B	Std. Error	Beta		
1	(Constant)	3.379	3.073		1.100	.279
	IQ score	.097	.024	.532	4.055	.000
	confidence in speaking time 1	-.958	.448	-.355	-2.138	.039
	words spoken per minute	.112	.086	.213	1.295	.204

a. Dependent Variable: attitude to school

Reporting Statistics in APA Style

A Short Guide to Handling Numbers and Statistics in APA Format

- http://my.ilstu.edu/~mshesso/apa_stats_format.html
- **Description:** The 16 teenagers who volunteered for the pilot study were younger than expected, $M = 14.2$ years, $SD = 1.3$.
- **Correlation:** The correlation of peer reports ($M = 4.2$, $SD = 2.1$, $N = 367$) and self reports ($M = 5.8$, $SD = 2.3$) of victimization was highly significant, $r(365) = .32$, $p = .008$.
- **Regression:** A linear regression analysis revealed that social skills was a highly significant predictor of aggression scores ($b = .40$, $p = .008$), accounting for 16% of the variance in aggressive behavior.
- Achievement test scores were regressed on class size and number of writing assignments. These two predictors accounted for just under half of the variance in test scores ($R^2 = .49$), which was highly significant, $F(2,289) = 12.5$, $p = .005$. Both the writing assignment ($b = .46$, $p = .001$) and the class size ($b = .28$, $p = .014$) demonstrated significant effects on the achievement scores.

- **t Tests:** The 36 study participants had a mean age of 27.4 ($SD = 12.6$) and were significantly older than the university norm of 21.2 years, $t(35) = 2.95, p = 0.01$.
- The 25 participants had an average difference from pre-test to post-test anxiety scores of -4.8 ($SD = 5.5$), indicating the anxiety treatment resulted in a highly significant decrease in anxiety levels, $t(24) = -4.36, p = .005$ (one-tailed).
- The 36 participants in the treatment group ($M = 14.8, SD = 2.0$) and the 25 participants in the control group ($M = 16.6, SD = 2.5$), demonstrated a significant difference in performance ($t[59] = -3.12, p = .01$); as expected, the visual priming treatment inhibited performance on the phoneme recognition task.