

One-Way ANOVA

Between-Group Mean Square	Within-Group Mean Square	F-Ratio
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1) (Subtract overall mean of pop from each group's mean)²

1) (subtract overall mean of pop from each group (sample) mean),

1) [(between group mean square) / (w/in-group mean square)]

2) (squared difference) (sample size)

2) then multiple each difference by (n-1)

2) if ~ 1, then btwn-g groups & w/in-groups variances similar, accept H0

3) compute degree of freedom (number of groups minus 1)

3) calculate the grand sum

3) if >1, then reject H0

4) calculate between-g-groups mean square = [(btwn-group variance) / (df)]

4) calculate the degrees of freedom total (N-n of groups)

One-Way ANOVA (cont)

5) calculate the w/in groups mean square = [(sum of squares) / (degrees of freedom total)]

- Analysis of Variance (compares means between 3+ samples)

Does not indicate which group(s) are different from which other groups (s)

- Parametric test

- Bonferroni post hoc test, reveals which specific means differed. Use if ANOVA was sig. using for pairwise comparison

- It multiplies each of the significance levels from the LSD test by the number of tests performed. If this value is greater than 1, then a significance level of 1 is used.

Chi-Square Test

1) calculate the expected frequency (E) = [(row total) (column total) / total sample N]

2) for each cell, find (difference between observed & expected counts)²

reveal what cell adds the most statistical value to the test.

to measure the strength of association of chi-square test

Chi-Square Test (cont)

3) divide square difference by expected count for each cell, then sum results

2x2 table
greater than 2x2 table

4) df = [(n of rows -1) (n of columns -1)]

5) check X2 table for significance at @ 0.05 alpha level

- **Dependent & Independent nominal/nominal or nominal/ordinal data**

- H0= no relationship between variables; expected counts for each cells = observed counts

- n is greater/equal to 20; no expected frequencies less/equal to 5 in 20% or more of the cells

Fisher's Exact Test for Chi-Square

-Use when Chi-Square assumptions are violated (>20%)
- Very small samples

Spearman's Rank Correlation

1) Turn raw scores into ranks

Rho varies from -1 to +1

2) find d2 = (difference between rankings)²

-1 (a perfect negative correlation; as X increases, y decreases)

3) add up all the data in d2 column to obtain sumd2

0 = no association

Spearman's Rank Correlation (cont)

4) calculation +1 (a perfect
spearman's rank correl-
ation coefficient (rho) rs
= $[1 - (6 \cdot \text{sumd}^2) / (N^3 - N)]$ increases, Y
df= n-2 increases

- Measures of associate for two ordinal variables; whether a relationship exists, how strong it is, what is the direction/pattern of relationship) (what happens to one variable, happens to the other variable)
- Nonparametric version of Pearson correlation coefficient
- H_0 = no sig

independent = x ; dependent = y

Pearson's R Correlation Coefficient

r = Rho = measure of association (-1 to +1)
assumes x and y is normally distr. & linearly related

(Pearson's r^2) = PRE stat (strength of predicting amount of variance in Y based on X)

r^2 = % of variance in dependent (Y) explained by independent (X)

usually **interval/ratio level data**

Parametric vs. Non-parametric Tests

Parametric	Non-Parametric
interval or ratio data	nominal and/or ordinal data
one-way ANOVA	Distribution free
Pearson's R Correlation Coefficient	Wilcoxon Signed-Rank Test for Two Related Conditions
	Mann-Whitney U Test for Two Independent Conditions
	Wilcoxon Rank Sum Test for Two Independent Conditions
	Chi-Square Test

Parametric vs. Non-parametric Tests (cont)

Kruskal-Wallis
Spearman's Rank Correlation

Wilcoxon Rank-Sum & Mann-Whitney U tests

nonparametric equivalent of independent-sample t-test

nominal and/or ordinal data

Tests two independent conditions

Wilcoxon Signed-Rank

- Use this test for **two related conditions (paired, matched)**
- ordinal data
- nonparametric equivalent to the **dependent-sample t-test**

H_0 = The two groups are identically distributed.

Kruskal-Wallis

nonparametric equivalent of one-way ANOVA

nominal or ordinal data, but more than two independent samples

uses chi-square distribution

Regression

Predicts **dependent (y)** based on value of **independent (x)**

Regression Formula: line that makes the sum of squares of the vertical distances of the data points from the line as small as possible

Principle of least-squares - finds estimates of parameters in a stat model based on observed data

$y = a + bx$; a= y-axis; b= slope

interval/ratio level data

assumes linear relationship
observes independent (x)

Correlation

Tests for	Difference between (r) and (r) ²	Assumptions
How well X predicts Y	$r =$ Pearson's correlation coefficient = measure of association	For each independent (x), dependent (y) must be normal
how "tightly the predicted values fit regression line	r^2 = PRE stat (strength of predicting amount of variance in Y based on X)	Dependent variable variances same for all independent values (homoscedasticity)
to what degree X covaries with Y	r^2 = % of variance in dependent (Y) explained by independent (X)	Avoid predictions outside the observed values; beware extremes; relationships must be linear over all values.
		linear relationship, observes independent (X)
		usually, interval/ratio level data

