

Research Methods

Primary Data Collection: Quantitative Data Analysis

Two main types of data: Categorical and Numerical data.

- Categorical data:
 - grouped into a descriptive set (nominal);
 - put in rank order (ordinal).
- Numerical data - measured using numbers.
 - discrete data –discrete units
 - continuous data –any value

Describing data

- Description:
 - • The central tendency, which is the common, middle or average value: mode, median, mean.
 - • The dispersion, which is how the data values are spread, or dispersed, around the central tendency: standard deviation. IQR, range.
- Inference
 - Exploring Relationships: correlation, regression
 - Comparing Groups: t-test, ANOVA, ANCOVA

Presenting your Data

- Tables
- Bar Graph
- Histogram
- Pie Chart
- Trend Line
- Scatter graph

Population/Sample

- Your sample must represent your population.

Exploring Relationships:

- The association between two variables.
- The correlation between two variables.

Correlation

Correlation Coefficient (r)

- Categorical data: Chi-squared coefficient
- Ranked data: Kendall's rank correlation coefficient, Spearman's rank correlation coefficient.
- Numerical data: Pearson's product moment correlation coefficient

Hypotheses

H_0 - **null hypothesis** - no relationship between two measured phenomena.

H_1 - **alternative hypothesis**

Rejecting or disproving the **null hypothesis** is a central task in modern scientific practice.

Statistically significant means that the null hypothesis has been rejected. This means we can accept the **alternative hypothesis**.

Significance

How likely the relationship between your variables has occurred by chance.

- Significance (alpha level) p
- Significant at $p < .05$ or 0.01
- There is less than a 5% (1%) chance that the result is due to chance.
- This means that the finding has a 95% (99%) chance of being true.

NB:

Statistical Significance $\neq$ Meaningful or Important

Correlation $\neq$ Causality: Third variable/Confounding variable

Regression

Coefficient of Determination - R^2 : a measure of how much variance is shared, how much of one variable can be explained by another.

Comparing Groups: Difference

t-test (unrelated) – independent samples t-test

Compares 2 unrelated groups (e.g. different towns) on same test. Need one categorical independent variable with only two groups (e.g. M/F) and one continuous dependent variable (e.g. items purchased).

t-test (related) – paired samples / matched t-tests / matched pairs

Compares the same group on 2 occasions (e.g. same product at two times). Need one categorical independent variable (e.g. time 1 & time 2) and one continuous dependent variable (e.g. product).

Effect Size

Effect sizes either measure the sizes of associations between variables or the sizes of differences between group means.

Several measures have been proposed and used:

- coefficient of determination – R
- the standardized mean difference - Cohen's d – and η^2 - eta squared - the correlation ratio.

Benchmarks (Cohen, 1988)

	Small	Medium	Large
R^2	.01	.09	.25
Cohen's d	.20	.50	.80
Eta squared - η^2	.01	.06	.14

Analysis of Variance (ANOVA): If three or more distinct groups:

One-Way Between-Groups ANOVA: Single independent variable.

Two-Way Between-Groups ANOVA: Factorial/Multifactorial ANOVA

2 independent variables - two factors

Post Hoc Tests

Homogeneous Subsets

Analysis of Covariance (ANCOVA): If three or more distinct groups – need to control for 3rd variable:

Two-Way ANCOVA) – two factors, controlling for 3rd variable

Summary

- We identify the variable of interest (dependent variable).
- We use descriptive statistics to get ideas about potential patterns in the data.
- These patterns then help us to devise experimental and null hypotheses.
- We then use inferential statistics to test the null hypothesis.
- If these inferential statistics return a p-value less than or equal to 0.05, then we have statistical significance and can reject the null hypothesis.
- If the p-value is greater than 0.05, then the null hypothesis cannot be rejected and we are unable to support the claims made by the experimental hypothesis.

References & Reading

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