Summary & Conclusion



Lecture 10

Survey Research & Design in Psychology James Neill, 2017 Creative Commons Attribution 4.0

Overview

- A
- 1. Module 1: Survey research and design
 - 1. Survey research
- Survey design
- 2. Module 2: Univariate and bivariate
 - 1. Descriptives & graphing
 - 2. Correlation
- 3. Module 3: Psychometrics
 - Exploratory factor analysis
 - 2. Psychometric instrument development
- 4. Module 4: Multiple linear regression
 - 1. MLR I
 - 2. MLR II
- 5. Module 5: Power & summary
 - 1. Power & effect sizes
 - 2. Summary and conclusion

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Survey research (Lecture 1)

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Types of research

- Surveys are used in all types of research:
 - Experimental
 - Quasi-experimental
 - Non-experimental

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What is a survey?

- · What is a survey?
 - A standardised stimulus for converting fuzzy psychological phenomenon into hard data.
- History
 - Survey research has developed into a popular social science research method since the 1920s.

Purposes of research

- Purposes/goals of research:
 - Information gathering
 - Exploratory
 - Descriptive
 - Theory testing/building
 - Explanatory
 - Predictive

Survey research

Survey research

Pros include:

- · Ecological validity
- Cost efficiency
- · Can obtain lots of data

Cons include:

- · Low compliance
- · Reliance on self-report

Survey design (Lecture 2)

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interviewadministered

Survey types

Self-administered

Pros:

- cost

- demand characteristics
- access to representative sample
- anonymity

Cons:

- non-response
- adjustment to cultural differences, special needs

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Survey questions

- Objective versus subjective questions:
 - Objective there is a verifiably true answer
 - Subjective based on perspective of respondent
- Open versus closed questions:
 - Open empty space for answer
 - Closed pre-set response format options

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Response formats

- 1. Dichotomous and Multichotomous
- 2. Multiple response
- 3. Verbal frequency scale (Never ... Often)
- 4. Ranking (in order → Ordinal)
- 5. Likert scale (equal distances)
- 6. Graphical rating scale (e.g., line)
- 7. Semantic differential (opposing words)
- 8. Non-verbal (idiographic)

Level of measurement

1. Categorical/Nominal

- 1. Arbitrary numerical labels
- 2. Could be in any order

2. Ordinal

- 1. Ordered numerical labels
- 2. Intervals may not be equal

3. Interval

- 1. Ordered numerical labels
- 2. Equal intervals

4. Ratio

- 1. Data are continuous
- 2. Meaningful 0

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Sampling

- 1. Key terms
 - 1. (Target) population
 - 2. Sampling frame
 - 3. Sample
- 2. Sampling
 - 1. Probability
- 2. Non-probability
- 1. Simple (random) 1. Convenience
- 2. Systematic
- 2. Purposive
- 3. Stratified
- 3. Snowball

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Non-sampling biases

- 1. Acquiescence
- 2. Order effects
- 3. Fatigue effects
- 4. Demand characteristics
- 5. Hawthorne effect
- 6. Self-serving bias
- 7. Social desirability

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Descriptives & graphing (Lecture 3)

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Getting to know data

- · Play with the data
- Don't be afraid you can't break data
- Check and screen the data
- Explore the data
- · Get intimate with data
- Describe the main features
- Test hypotheses

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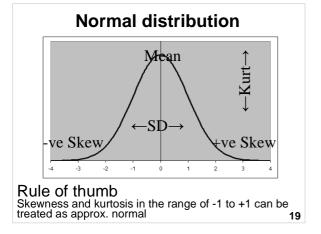
Summary: LOM & statistics

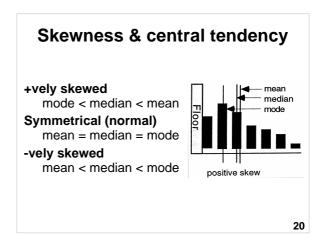
- If a normal distribution can be assumed, use parametric statistics (more powerful)
- If not, use non-parametric statistics (less power, but less sensitive to violations of assumptions)

Descriptive statistics

- What is the **central tendency**?
 - Frequencies, Percentages (Non-para)
 - -Mode, Median, Mean (Para)
- What is the variability?
 - -Min, Max, Range, Quartiles (Non-para)
 - -Standard Deviation, Variance (Para)

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Principles of graphing

- Clear purpose
- Maximise clarity
- Minimise clutter
- Allow visual comparison

Univariate graphs
Bar graph
Pie chart
Histogram
Stem & leaf plot
Data plot / Error bar
Box plot

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Correlation (Lecture 4)

Covariation

- 1. The world is made of covariations.
- 2. Covariations are the building blocks of more complex multivariate relationships.
- Correlation is a standardised measure of the covariance (extent to which two phenomenon co-relate).
- Correlation does not prove causation - may be opposite causality, bi-directional, or due to other variables.

Purpose of correlation

The underlying purpose of correlation is to help address the question:

What is the

- relationship or
- association or
- shared variance or
- co-relation

between two variables?

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What is correlation?

- 1. Standardised covariance
- 2. Ranges between -1 and +1, with more extreme values indicating stronger relationships
- 3. Correlation does not prove causation may be opposite causality, bi-directional, or due to other variables.

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Types of correlation

- Nominal by nominal:
 Phi (Φ) / Cramer's V, Chi-squared
- Ordinal by ordinal:
 Spearman's rank / Kendall's Tau b
- Dichotomous by interval/ratio: Point bi-serial r_{pb}
- Interval/ratio by interval/ratio: Product-moment or Pearson's r

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Correlation steps

- 1. Choose correlation and graph type based on levels of measurement.
- 2. Check graphs (e.g., scatterplot):
 - -Linear or non-linear?
 - -Outliers?
 - -Homoscedasticity?
 - -Range restriction?
 - -Sub-samples to consider?

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Correlation steps

- 3. Consider
 - -Effect size (e.g., Φ , Cramer's V, r, r^2)
 - -Direction
 - -Inferential test (p)
- 4. Interpret/Discuss
 - -Relate back to hypothesis
 - -Size, direction, significance
 - -Limitations e.g.,
 - Heterogeneity (sub-samples)
 - Range restriction
 - Causality?

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Interpreting correlation

- Coefficient of determination
 - -Correlation squared
 - -Indicates % of shared variance

 Strength
 r
 r^2

 Weak:
 .1 - .3
 1 - 10%

 Moderate:
 .3 - .5
 10 - 25%

 Strong:
 > .5
 > 25%

Assumptions & limitations

- 1. Levels of measurement
- 2. Normality
- 3. Linearity
 - 1. Effects of outliers
 - 2. Non-linearity
- 4. Homoscedasticity
- 5. No range restriction
- 6. Homogenous samples
- 7. Correlation is not causation

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Exploratory factor analysis (Lecture 5)

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What is factor analysis?

- Factor analysis is a family of multivariate correlational data analysis methods for summarising clusters of covariance.
- FA summarises correlations amongst items into a smaller number of underlying fuzzy constructs (called factors).

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Steps / process

- 1. Test assumptions
- 2. Select extraction method and rotation
- 3. Determine no. of factors (Eigen Values, Scree plot, % variance explained)
- 4. Select items (check factor loadings to identify which items belong in which factor; drop items one by one; repeat)
- 5. Name and define factors
- 6. Examine correlations amongst factors
- 7. Analyse internal reliability

Lecture

8. Compute composite scores

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Assumptions

- Sample size
 - 5+ cases per variables (ideally 20+ cases per variable)
 - Another guideline: N > 200
- Bivariate & multivariate outliers
- Factorability of correlation matrix (Measures of Sampling Adequacy)
- Normality enhances the solution

Types of FA

- PAF (Principal Axis Factoring): Best for theoretical data exploration
 - -uses shared variance
- **PC** (Principal Components): Best for data reduction
 - -uses all variance

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Rotation

- Orthogonal (Varimax)
 - perpendicular (uncorrelated) factors
- Oblique (Oblimin)
 - angled (correlated) factors
- Consider trying both ways
 - Are solutions different? Why?

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Factor extraction

How many factors to extract?

- Inspect EVs
 - look for > 1 or sudden drop (inspect scree plot)
- % of variance explained
 - aim for 50 to 75%
- Interpretability
 - does each factor 'make sense'?
- Theory
 - does the model fit with theory?

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Item selection

An EFA of a good measurement instrument ideally has:

- a simple factor structure (each variable loads strongly (> +.50) on only one factor)
- each factor has multiple loading variables (more loadings → greater reliability)
- target factor loadings are high (> .6) and cross-loadings are low (< .3), with few intermediate values (.3 to .6).

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Psychometrics instrument development

(Lecture 6)

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Psychometrics

- Science of psychological measurement
- 2. Goal: Validly measure individual psychosocial differences
- 3. Develop and test psychological measures e.g., using
 - 1. Factor analysis
 - 2. Reliability and validity

Concepts & their measurement

- 1. Concepts name common elements
- 2. Hypotheses identify relations between concepts
- 3. Brainstorm indicators of a concept
- 4. Define the concept
- 5. Draft measurement items
- 6. Pre-test and pilot test
- 7. Examine psychometric properties
- 8. Redraft/refine and re-test

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Measurement error

- 1. Deviation of measure from true score
- 2. Sources:
 - 1. Non-sampling (e.g., paradigm, respondent bias, researcher bias)
 - 2. Sampling (e.g., non-representativeness)
- 3. How to minimise:
 - 1. Well-designed measures
 - 2. Reduce demand effects
 - 3. Representative sampling
 - 4. Maximise response rate
 - 5. Ensure administrative accuracy

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Reliability

- 1. Consistency or reproducibility
- 2. Types
 - 1. Internal consistency
 - 2. Test-retest reliability
- 3. Rule of thumb
 - 1. > .6 OK
 - 2. > .8 Very good
- 4. Internal consistency
 - 1. Split-half
 - 2. Odd-even
 - 3. Cronbach's alpha

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Validity

- Extent to which a measure measures what it is intended to measure
- 2. Multifaceted
 - 1. Compare with theory and expert opinion
 - 2. Correlations with similar and dissimilar measures
 - 3.Predicts future

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Composite scores

Ways of creating composite (factor) scores:

1. Unit weighting

- 1.Total of items or
- 2. Average of items (recommended for lab report)

2. Regression weighting

1. Each item is weighted by its importance to measuring the underlying factor (based on regression weights)

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Writing up instrument development

- 1. Introduction
 - 1. Review constructs & previous structures
 - 2. Generate research question
- 2. Method
 - 1. Explain measures and their development
- 3. Results
 - 1. Factor analysis
 - 2. Reliability of factors
 - 3. Descriptive statistics for composite scores
 - 4. Correlations between factors
- 4. Discussion
 - 1. Theory? / Measure? / Recommendations?

Multiple linear regression (Lectures 7 & 8)

General steps

- 1. Develop model and hypotheses
- 2. Check assumptions
- 3. Choose type
- 4. Interpret output
- 5. Develop a regression equation (if needed)

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Linear regression

- 1. Best-fitting straight line for a scatterplot of two variables
- 2. Y = bX + a + e
 - 1. Predictor (X; IV)
 - 2. Outcome (Y; DV)
- 3. Least squares criterion
- Residuals are the vertical distance between actual and predicted values

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MLR assumptions

- 1. Level of measurement
- 2. Sample size
- 3. Normality
- 4. Linearity
- 5. Homoscedasticity
- 6. Collinearity
- 7. Multivariate outliers
- 8. Residuals should be normally distributed

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Level of measurement and dummy coding

- 1. Levels of measurement
 - 1. DV = Continuous (Likert or ratio + normal)
 - 2. IV = Continuous or dichotomous
- 2. Dummy coding
 - Convert complex variable into series of dichotomous IVs

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General steps

- 1. Develop model and hypotheses
- 2. Check assumptions
- 3. Choose type
- 4. Interpret output
- 5. Develop a regression equation (if needed)

Multiple linear regression

1. Multiple IVs to predict a single DV:

$$Y = b_1 x_1 + b_2 x_2 + \dots + b_i x_i + a + e$$

- 2. Overall fit: R, R^2 , and Adjusted R^2
- 3. Coefficients
 - 1. Relation between each IV and the DV, adjusted for the other IVs
 - 2. B, β, t, p, and sr2
- 4. Types
 - 1. Standard
 - 2. Hierarchical
 - 3. Stepwise / Forward / Backward

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Summary: Semi-partial correlation (*sr*)

- In MLR, sr is labelled "part" in the regression coefficients table SPSS output
- 2. Square these values to obtain sr^2 , the unique % of DV variance explained by each IV
- Discuss the extent to which the explained variance in the DV is due to unique or shared contributions of the IVs 55

Residual analysis

- Residuals are the difference between predicted and observed Y values
- 2. MLR assumption is that residuals are normally distributed.
- 3. Examining residuals also helps assess:
 - 1. Linearity
 - 2. Homoscedasticity

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Interactions

- 1. In MLR, IVs may interact to:
 - 1. Increase the IVs' effect on the DV
 - 2. Decrease the IVs' effect on the DV
- 2. Model interactions using hierarchical MLR:
 - 1. Step 1: Enter IVs
 - 2. Step 2: Enter cross-product of IVs
 - 3. Examine change in R2

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Analysis of change

Analysis of changes over time can be assessed by:

- 1. Standard regression
 - Calculate difference scores (Post-score minus Pre-score) and use as a DV
- 2. Hierarchical MLR
 - 1. Step 1: "Partial out" baseline scores
 - 2. Step 2: Enter other IVs to help predict variance in changes over time.

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Writing up an MLR

- 1. Introduction
 - 1. Establish purpose
 - 2. Describe model and hypotheses
- 2. Results
 - 1. Univariate descriptive statistics
 - 2. Correlations
 - 3. Type of MLR and assumptions
 - 4. Regression coefficients
- 3. Discussion
 - 1. Summarise and interpret, with limitations
 - 2. Implications and recommendations

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Power & effect size (Lecture 9)

Significance testing

- 1. Logic At what point do you reject H₀?
- 2. History Started in 1920s & became very popular through 2nd half of 20th century
- 3. Criticisms Binary, dependent on $\it N$, ES, and critical $\it \alpha$
- 4. Practical significance
 - 1. Is an effect noticeable?
 - 2. Is it valued?
 - 3. How does it compare with benchmarks?

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Inferential decision making

		Reality	
		H_0 False	H_0 True
Test	Reject H ₀	Correct rejection H_0 = Power = 1 - β	Type I error = α
	Accept H_0	Type II error	Correct acceptance of H_0

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Statistical power

- 1. Power = probability of detecting a real effect as statistically significant
- 2. Increase by:
 - -↑ N
 - -↑ critical α
 - -↑ES
- Power
 - > .8 "desirable"
 - ~ .6 is more typical
- Can be calculated prospectively and retrospectively

Effect size

- 1. Standardised size of difference or strength of relationship
- Inferential tests should be accompanied by ESs and CIs
- 3. Common bivariate ESs include:
 - 1. Cohen's d
 - 2. Correlation r
- Cohen's d not in SPSS use an online effect size calculator

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Confidence interval

- 1. Gives 'range of certainty'
- 2. Can be used for B, M, ES etc.
- 3. Can be examined
 - 1. Statistically (upper and lower limits)
 - 2. Graphically (e.g., error-bar graphs)

Publication bias

- Tendency for statistically significant studies to be published over nonsignificant studies
- Indicated by gap in funnel plot → filedrawer effect
- 3. Counteracting biases in scientific publishing:
 - low-power studies tend to underestimate real effects
 - bias towards publish sig. effects over nonsig. effects66

Academic integrity

- Violations of academic integrity are evident and prevalent amongst those with incentives to do so:
 - 1. Students
 - 2. Researchers
 - 3. Commercial sponsors
- 2. Adopt a balanced, critical approach, striving for objectivity and academic integrity

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Unit outcomes

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Learning outcomes

- Design and conduct survey-based research in psychology;
- Use SPSS to conduct and interpret data analysis using correlation-based statistics, including reliability, factor analysis and multiple regression analysis;
- 3. Communicate in writing the results of survey-based psychological research

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Graduate attributes

- Display initiative and drive, and use organisation skills to plan and manage workload
- 2. Employ up-to-date and relevant knowledge and skills
- 3. Take pride in professional and personal integrity
- 4. Use creativity, critical thinking, analysis and research skills to solve theoretical and real-world problems

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Feedback

- •What worked well?
- •What could be improved?
- Direct feedback (e.g., email, discussion forum)
- •Interface Student Experience Questionnaire (ISEQ).

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