Evaluating experiments

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Spurious correlations

http://www.tylervigen.com/spurious-correlations

Depression and memory

- Depression is associated with over-general memory.
- Depression causes memory problems?
- Memory problems cause depression?
- Both causal directions?
- Neither causal direction (e.g. both caused by childhood trauma).
- ▶ It is not possible to distinguish between these accounts on the basis of correlational data.

Longitudinal data does not solve this problem

- Use of night lights in infancy is correlated with myopia in later life (true).
- Seems causal? Causes must precede effects. The later myopia cannot cause the earlier use of night lights. So, night lights must be causing myopia?
- Ban night lights? (genuinely recommended on basis on these data).

Third factor explanations are still possible in longitudinal research

- ► A third factor causes both the presence of night lights and myopia.
- Developing myopia in later life has a genetic component. If your parents are myopic, this increases the chance you will become myopic.
- Myopic adults, on average, favour higher levels of illumination. This drives their decision to use night lights in their baby's room.
- ► The parents' myopia causes both the presence of infant night lights and later myopia.
- ▶ Ban night lights? Clearly, this would be ineffective.

Correlation does not imply causation

- ► Correlational research is fundamentally limited.
- ▶ It is extremely unlikely that any two variables are completely unrelated.
- Many correlations in psychology are very small e.g.
 - Extroversion explaining 2% of the variation in some other variable.
 - ▶ 2% is detectably different from no correlation
 - but not meaningful (everything likely to be related to some degree).

Determining causation through the Experimental Method

- Simplest form
 - Take two groups of people
 - Do different things to those two groups.
 - Measure something
- Independent variable Intended difference in what we do to the two groups
- Dependent variable The thing we measure

Example: Testing a treatment for depression

- ► Group 1 6 weeks of the new therapy
- ► Group 2 Nothing.
- ► Take measure of depression at end (e.g. Beck Depression Inventory).
- ▶ Group 1 are less depressed than Group 2
- ► This has the potential to show that the therapy *causes* a reduction in depression.
- ...but there are other explanations.

Pre-existing differences

- Group 1 6 weeks of the new therapy
- ► Group 2 Nothing.
- What if Group 1 were happier to start with?
- ► Approaches to this problem
 - Detection
 - Prevention

Detection

- ► Take pre-treatment measures
- e.g. Measure BDI of both groups before (and after) treatment period.

	Pre	Post
Therapy	25	5
Control	25	25

Prevention

- Construct groups such that we eliminate pre-existing differences.
- Matching Take BDI measures for everyone. Allocate people to groups in such a way that the average BDI for the two groups is identical (or at least, minimised).
- Randomisation Allocate people to groups randomly.
- Matching versus Randomisation pros and cons.

Our therapy experiment

- ▶ Use large, randomised groups.
- ► Take pre-treatment measures
- ▶ Treatment caused the reduction in depression?

	Pre	Post
Therapy	25	5
Control	25	25

Attrition

- Attrition participants dropping out before the end of the study
- ► If attrition rates vary between conditions, you may have a major problem.

Example

Pre-treatment BDI scores

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Mean
Therapy 6 8 12 15 30 14.2
Control 6 8 12 15 30 14.2
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- ► The most-depressed 20% drop out of therapy (perhaps because the therapy is quite demanding).
- There are no drop-outs in the control condition (there's not much to drop out from).
- Both therapy and control are inert (no effect) post-treatment BDI equals pre-treatment BDI.

Example

Pre-test BDI scores

						Mean		
Therapy	6	8	12	15	30	14.2		
Control	6	8	12	15	30	14.2		
► Post-test BDI scores								
						Mean		
Therapy	6	8	12	15		10.25		
Control	6	8	12	15	30	14.2		

► A therapy we know to be ineffective appears to have worked, due to non-random attrition.

Placebo effect

- Classic example
 - Someone has a headache
 - Give them a pill with no active ingredient
 - ► Tell them it's a headache tablet
 - ► Their headache symptoms reduce
- ► Lesson In order to assess drug effectiveness you need to test drug vs. placebo, NOT drug vs. nothing.

Placebo effect in psychological therapy

- ▶ Perhaps the therapy is inert?
- ► The treatment group are happier because they have the expectation that what they are receiving will work.
- Problem a placebo pill is known to be inert; what is the equivalent in therapy?
- ► There is no agreement there's someone willing to endorse the effectiveness of almost any therapy.

Placebo effect in psychological therapy

- ➤ Solution set out to show that your new therapy works better than an existing treatment (or, as well as existing treatment, if yours is better in some practical way e.g. cheaper).
- Problem this is seldom done.

Experimenter Effects - Data analysis - Example

- Diary entries as a measure of happiness.
- ▶ Participants write about their feelings
- Experimenter rates for level of happiness.
- ► If experimenter knows which condition the participant is in, this may bias their assessment of happiness.

Experimenter Effects - Data analysis

- Machine-recorded measures (e.g. reaction time) immune?
- No! Data analysis typically involves many decisions, all open to bias.
- ► If the experimenter knows which condition the participants are in, this could bias their decisions.

Blind testing

- Single-blind testing participant does not know which condition they are in.
 - e.g. Drug vs. placebo. Participants do not know which condition they are in.
- ▶ Double-blind testing single-blind testing plus the experimenters do not know which condition is which until after they have completed their analysis.

Pre-registration

"The first principle is that you must not fool yourself, and you are the easiest person to fool" - Richard Feynman.

► Record your hypothesis, method, and analysis plan, before you analyse the data.

Difference versus no difference designs

- ► The preferred hypothesis is that people differ in the speed with which they react to auditory and visual alarm signals.
- ► The alternative theory against which this is compared is that there is no difference (nil hypothesis).
- Problem Experimental control is never perfect.
- ► Thus the nil hypothesis is almost certainly wrong, and detectably so if you test enough people.
- Thus the result of the study is known before you run it.
- ▶ Thus There was no point in running it.

Better alternatives 1

- Directional hypotheses
 - ▶ The preferred theory is that auditory is faster.
 - ► The alternative theory against which this is compared is that there is no difference (nil hypothesis).
 - If you find visual faster, you have disproved your theory.
 - So, whatever the result, there was a point to running this experiment (because the theory was falsifiable).

Better alternatives 2

- Strong inference
 - ▶ One well-established theory predicts that auditory is faster.
 - ▶ Another well-established theory predicts that visual is faster.
 - Whatever you find in this study, you've gained information (except in the unlikely case where the nil hypothesis was true).

Evaluating an experiment

Mueller, P. A., & Oppenheimer, D. M. (2014). The pen is mightier than the keyboard: Advantages of longhand over laptop note taking. *Psychological science*, *25*, 1159-1168.

- 1. Find the full text of this paper on Google Scholar
- 2. Read from the title up to, but not including, the "Study 2" subheading.
- 3. Evaluate how good Study 1 is, using the *checklist* to help remind you of what we've covered today.
- 4. Agree on a score, be ready to report your score, and to answer some questions.

Further reading/ watching

The notes for this lecture cover a number of additional relevant topics.

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