17-803 Empirical Methods Bogdan Vasilescu, Institute for Software Research

UPMC

Designing Experiments (Part II)

Tuesday, March 16, 2021





Outline for Today

Example experiment papers Experimental design



O'REILLY®



& Peter Gedeck



STATISTICS

Fourth Edition

David Freeman, Robert Pisani, and Roger Purves



Example papers

WSDM (Conference on Web Search and Data Mining) Experiment

Setup

- Four committee members reviewed each paper
- Two single blind, two double blind

Results

- "Reviewers in the single-blind condition [...] preferentially bid for papers from top universities and companies."
- universities [1.58], and top companies [2.10]."

Tomkins, A., Zhang, M., & Heavlin, W. D. (2017). Reviewer bias in single-versus double-blind peer review. Proceedings of the National Academy of Sciences, 114(48), 12708-12713.

Single-blind reviewers are significantly more likely than their double-blind counterparts to recommend for acceptance papers from famous authors [odds multiplier 1.64], top





NeurIPS (Conference on Neural Information Processing Systems) Experiment

Setup

- Organizers split the program committee down the middle
- Most submitted papers were assigned to a single side
- 10% of submissions (166) were reviewed by both halves of the committee

Results

(with a 95% confidence interval of 40-75%)"

http://blog.mrtz.org/2014/12/15/the-nips-experiment.html

"most papers [57%] at NeurIPS would be rejected if one reran the conference review process





- Two classes of students at Miami University of Ohio that studied objectoriented (OO) design in a one semester course:
 - Control group (random sample): OO design class
 - Treatment group (volunteers): OO design class + formal methods No statistical difference between the abilities of the two groups on standardized ACT pre-tests
- As project, both classes were assigned the development of an elevator system
 - Students had to hand in:
 - functioning executable + source code
 - (+ formal specification written using first-order logic)

Sobel, A. E. K., & Clarkson, M. R. (2002). Formal methods application: An empirical tale of software development. IEEE Transactions on Software Engineering, 28(3), 308-320.





- Standard set of test cases:
 - > 45.5% of control teams passed all tests
 - 100% of treatment teams
- Conclusions:
 - "formal methods students had increased complex-problem solving skills"

Sobel, A. E. K., & Clarkson, M. R. (2002). Formal methods application: An empirical tale of software development. IEEE Transactions on Software Engineering, 28(3), 308-320.

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"the use of formal methods during software development produces 'better' programs"



"Unfortunately, the paper contains several subtle problems. The reader unfamiliar with the basic principles of experimental psychology may easily miss them and interpret the results incorrectly. Not only do we wish to point out these problems, but we also aim to illustrate what to look for when drawing conclusions from controlled experiments."

Berry, D. M., & Tichy, W. F. (2003). Comments on "Formal methods application: an empirical tale of software development". IEEE Transactions on Software Engineering, 29(6), 567-571.



- Confounding variables:

 - b differences in motivation (treatment group volunteers more motivated) differences in exposure (treatment group more instruction)
 - differences in learning style (treatment group better learners)
 - differences in skills (outside of ACT)
- Novelty effects

Berry, D. M., & Tichy, W. F. (2003). Comments on "Formal methods application: an empirical tale of software development". IEEE Transactions on Software Engineering, 29(6), 567-571.





Causal relationships

Cause

- inus condition "insufficient but nonredundant part of an unnecessary but sufficient condition"
- Example: match to start a forest fire Fires can start even without matches → Match is not a necessary condition Matches don't always start forest fires (e.g., not on long enough, rainy weather) → Match is not a sufficient condition





Cause

- inus condition "insufficient but nonredundant part of an unnecessary but sufficient condition"
- Match is part of a bigger constellation of conditions without which a fire would not result
 - Insufficient: needs oxygen, dry leaves, etc
 - Nonredundant: needs to add something unique besides oxygen, dry leaves, etc

redundant condition" on of d not result

c unique







Effect

- Counterfactual: what would have happened to these subjects had the cause not been present?
 - What did happen when people received a treatment, vs
 - What would have happened to those same people if they simultaneously had not received the treatment ("counterfactual", i.e., contrary to fact)
 - Effect is distance between the two

Can't observe, must infer / approximate.







Experimental design: Creating a high-quality but necessarily imperfect source of counterfactual inference Understanding how this source differs from the treatment condition

Ingredients for Establishing a Causal Relationship?



Ingredients for Establishing a Causal Relationship

The cause preceded the effect

The cause was related to the effect

We can find no plausible alternative explanation for the effect other than the cause



Ingredients for Establishing a Causal Relationship

Cause

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Note how this mirror what happens in experiments.

No other scientific method regularly matches the characteristics of causal relationships so well.

Aside: Mediators & Moderators



Independent variable X





Links in the explanatory chain: Mediator

Independent variable X

Mediating variable M

Dependent variable Y



Socioeconomic status

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Independent variable X

Causal relationship varies in strength (or direction) as Z varies: Moderator

Dependent variable Y

Moderating variable Z



Work experience



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Salary

Gender



Aside: Correlation is not enough!

Correlation Does Not Prove Causation

- > Which variable came first?
- Are there alternative explanations for the presumed effect?
- Example: income ~ education or education ~ income? Confounding variable: intelligence, family socioeconomic status (causes both high)
 - education and high income)







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

Number of people who drowned by falling into a pool correlates with Films Nicolas Cage appeared in Correlation: 66.6% (r=0.666004) 1999 2005 2008 2000 2001 2002 2003 2004 2006 2007 140 drownings Swimming pool drownings 120 drownings 100 drownings 80 drownings 2001 2002 2003 2004 2005 2006 2007 2008 1999 2000 Nicholas Cage Swimming pool drownings

Data sources: Centers for Disease Control & Prevention and Internet Movie Database

Per capita cheese consumption correlates with Number of neonle who died by becoming tangled in their bedsheets

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



Data sources: U.S. Census Bureau and National Science Foundation

Worldwide non-commercial space launches correlates with Sociology doctorates awarded (US)

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Experiments: Summary Pros and Cons

Advantages and Disadvantages of Experiments

- Disadvantages of experiments:
 - Conditions may be unrealistic
 - Fell nothing about how and why effects occurred
 - Cannot deal with cases when we first observe effect and need to look for causes



occurred serve





Advantages and Disadvantages of Experiments

- Disadvantages of experiments:
 - Conditions may be unrealistic
 - Fell nothing about how and why effects occurred
 - Cannot deal with cases when we first observe effect and need to look for causes
- Unique advantage:
 - Causal description: describe consequences attributable to deliberately varying a treatment
 - (But not causal explanation / mechanisms)





... to be continued

Credits

- Graphics:
 - Dave DiCello photography (cover)

Content:

- - Ch1: Experiments and generalized causal inference
 - Ch2: Statistical conclusion validity and internal validity
 - Ch3: Construct validity and external validity
 - Ch8: Randomized experiments
- Bruce, P., Bruce, A., & Gedeck, P. (2020). Practical Statistics for Data Scientists: 50+ Essential Concepts Using R and Python. O'Reilly Media.
- Freedman, D., Pisani, R., Purves, R., & Adhikari, A. (2007). Statistics.

Chapters from Shadish, W. R., Cook, T. D., & Campbell, D. T. (2002). Experimental and quasi-experimental designs for generalized causal inference. Wadsworth Publishing

