# **CSE 599K**

# **Empirical Research Methods**

Winter 2025

**SE** meets Science

**Analysis validity** 

## **Today**

- Recap: analysis validity
- SE principles for rigorous Science
- Two example studies

#### Analysis validity: open questions

#### **External validity**

- Does the experiment generalize (to larger population, other subjects, etc.)?
- How representative is the sample?

#### **Internal validity**

- Does the experiment isolate the variable(s) of interest?
- Does the experiment control for confounders and unwanted effects?

#### **Construct validity**

- Does the experiment measure what it claims to measure?
- Do the proxy measures and tools adequately measure the concept of interest?

#### (Statistical) conclusion validity

- Are the conclusions valid based on the chosen statistical test and sample size?
- Are the conclusions valid based on the observed significance (p value)?

# SE principles for rigorous science

#### Science to practice is not a one-way street!







# Let's improve scientific rigor with SE principles and best practices!

## **Design reviews**



Design reviews are common in practice.



Embrace and value pre-registrations.



RFCs and public discussions (e.g., GH) provide valuable context.



Public (open) reviews should be a no-brainer!

#### **Quality assurance**



Modern code review is incremental (not holistic).



Move to pre-acceptance artifact evaluations.



Software testing is the most common QA approach in practice.



Require evidence for artifact testing.

Hark no more: On the preregistration of chi experiments, Cockburn et al., CHI 2018 https://openreview.net/

Expectations, outcomes, and challenges of modern code review, Bacchelli and Bird, ICSE 2013 Modern code review: a case study at Google, Sadowski et al., ICSE 2018

#### **Process**



Merge conflicts (branches) are resolved by branch authors.



Expect resolution (knowledge) of conflicting results.



Don't expect others to resolve your merge conflict!

#### **Process**



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Expect resolution (knowledge) of conflicting results.



No premature optimizations.



**Focus** on **design validity** before scrutinizing artifacts.

#### Science is a collaborative effort!







Software Engineering is a **collaborative effort**. We should view science the same way!

# Science as Amateur Software Development



- 1. How can software engineering principles improve the rigor of data analyses?
- 2. Are these principles equally applicable to computational notebooks?
- 3. Describe three specific quality control mechanisms.
- 4. McElreath attributes a significant number of incorrect (scientific) studies to "sloth". What are the specific issues he is calling out, and what solutions does he propose?
- 5. Provide an argument for why or why not general-purpose programming languages such as Python are an adequate choice for data analysis.

# Two example studies

### An example study: design

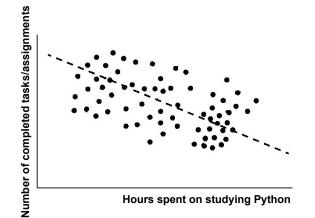
#### Goal:

Studying the **relationship** between **time spent** on **studying** Python and **success rate** in completing coding assignments.

#### Methodology:

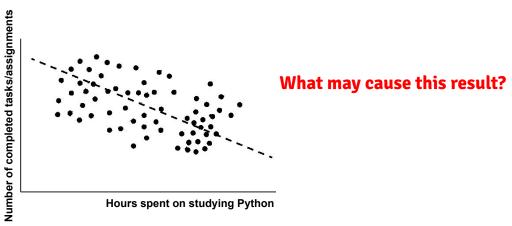
- ~100 participants are randomly selected in front of CSE.
- Each participant is given a high-level overview of the study.
- Each participant decides on how long to study before attempting to solve any coding assignment.
- Each participant solves as many coding assignments as possible in one hour (after studying).

#### An example study: conclusions



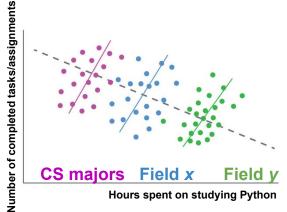
Conclusion: Spending more time on learning Python makes you a worse Python programmer.

#### An example study: conclusions



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# An example study: Simpson's paradox



Where did this study fail?

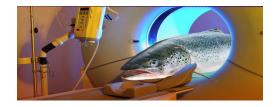
This phenomenon is called: Simpson's paradox.

#### Another example study



http://www.prefrontal.org/files/posters/Bennett-Salmon-2009.pdf

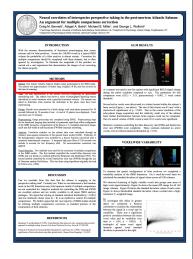
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**Subject:** One mature **Atlantic Salmon** (Salmo salar) participated in the **fMRI study**. The salmon was approximately 18 inches long, weighed 3.8 lbs, and was **not alive at the time of scanning**.



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Task: [...] open-ended mentalizing task. The salmon was shown a series of photographs depicting human individuals in social situations with a specified emotional valence. The salmon was asked to determine what emotion the individual in the photo must have been experiencing.



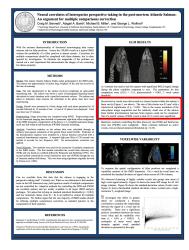


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Task: [...] open-ended mentalizing task. The salmon was shown a series of photographs depicting human individuals in social situations with a specified emotional valence. The salmon was asked to determine what emotion the individual in the photo must have been experiencing.

**Results:** Several active voxels were discovered [...] Out of a search volume of 8064 voxels a total of **16 voxels** were significant.



#### Another example study: conclusions

#### Interpretation of pure noise

- Noisy data source
- Multiple hypotheses tested on the same data
- An argument for multiple comparisons correction
- Analysis grounded in a **conceptual model?**
- Clear operationalization (implementation)?
- Implementation consistent with the model?
- Proper use of statistical methods?
- Data interpreted in context of prior knowledge?
- Explored and validated alternative hypotheses?

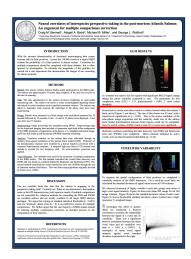
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Valid data analysis goes well beyond implementation correctness.



Where did this study fail (on purpose)?