

# Experimental Design and Data Analysis

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## Scientific Method

1. Idea - Question - Something to test.
2. Review previous work.
3. Articulate that.
4. Design an experiment
5. Report and Analyze results

## Method types

1. Quantitative research.
  - a. Numbers.
  - b. Large size (increases representation).
  - c. Error range.
  
2. Qualitative research.
  - a. Ideas and words.
  - b. Small size (time constraint)
  - c. Biases.

## Experiment Design

1. Baseline experiment
2. Change exactly one variable.
3. Record new results.
4. Repeat runs
  - a. Reduce error.

Be careful !!!!

## Do

- Try different approaches.
- Repeat experiments
- Compare your results with others
- Analyze results
- Control for different sources of variability.
- Plan for repeatability by outsiders.

## Don't

- Ignore controlled variables effect
- make large approximations
- Claim first to result unless really sure
- Compare different metrics

## Experiments Discussion

- What do you think is most important in designing an experiment?
- What do you try to avoid?

## Data Collection

1. What types of data?
2. Data size?
3. Sampling?
4. Method limitations?
5. Collection design.
6. How to keep the data?
7. Reporting and representation?

## Avoid bias

- Researchers bias
- Participants bias
  - Different genders.
  - Different cultures.
- Methodology bias
- Interpretation bias
  - Income increase: people surviving inflation - luxury



## Ethical Considerations

1. Validity.
2. Accurate reporting.
3. Human involved?
4. Confidentiality?
5. Voluntary Participation and Consent?
6. No harm to volunteers.
7. Institutional Review Board (IRB)

## Ethical Considerations Discussion

- How do you collect data? avoid bias?
- How to avoid unethical issues?

## Data Analysis

- One of the more important parts of presenting your work; it conveys empirical validation of your hypothesis.
- Both a messaging component and a technical component
- Reproducibility just as important as in experiments
- Also an integrity component

## Data Analysis: Technical

- Use enough data samples
  - Try to make sure data isn't biased somehow
  - Make sure analysis supports usefulness of method in the real world.
  - Verify correctness
- Use correct metrics\*
- Version control + config files\*

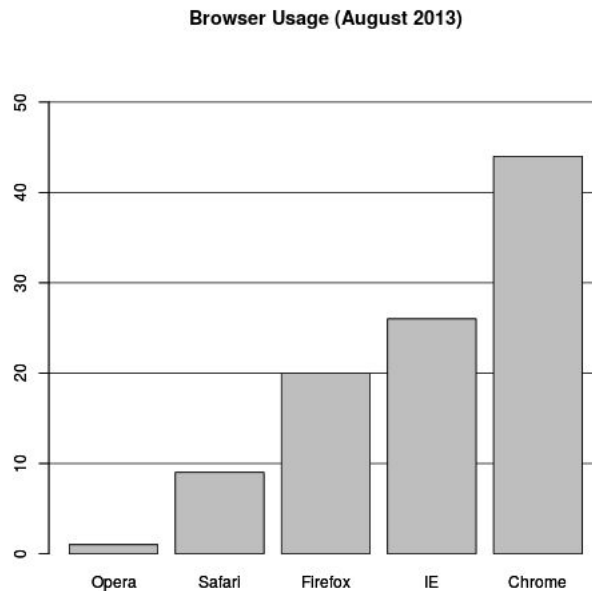
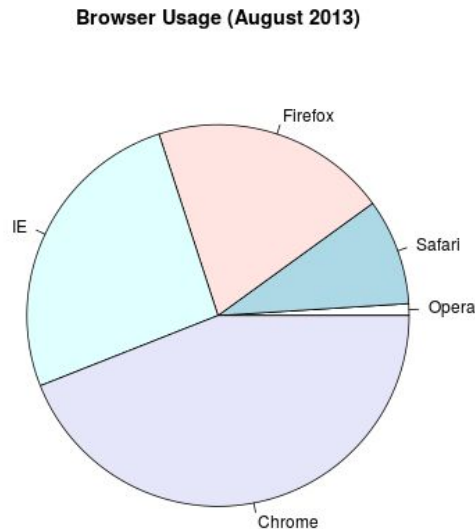
\* Credit to Arsenii, Feyza, & Samarth ([https://cs-people.bu.edu/mathan/classes/CS697/slides/cs697\\_week8.pdf](https://cs-people.bu.edu/mathan/classes/CS697/slides/cs697_week8.pdf))

## Data Analysis: Messaging

- Use clear description of methodology; reproducibility in analysis is important too.
- Make clear figures

## Data Analysis: Messaging

- Which figure is better for describing browser usage?



Credit to [https://genomicsclass.github.io/book/pages/plots\\_to\\_avoid.html](https://genomicsclass.github.io/book/pages/plots_to_avoid.html)

## Data Analysis: Integrity

- Highlighting strong results or withholding poor results?
  - Example: Models which generate high-quality images. Show random selection or all?
- Discuss with your advisor about this
- Discussion question: Is cherry-picking results acceptable if you don't claim they are randomly sampled? Everyone else does it. What if you're asked to cherry pick by collaborators?

## Reproducibility in Data Analysis

- How can you set up analysis to make it easy to reproduce?
  - Version control + configs!!!
- How do community expectations differ around reproducibility? Are they changing?
  - For ML; papers with code webpage!
- Discussion question: Problems in reproducing analysis?