**Welcome to #BITSS2014!**
Agenda:
Thursday 9A-12P Training
Thursday 1P-5P Conference
Friday 8:30A-2:30P Research Seminar

# Registrations and Pre-Analysis Plans
## Making research more transparent and reproducible

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University of California Berkeley
and
Center for Open Science

BITSS Annual Meeting, 2014

# Outline

# Publication Bias

- There is a higher fraction of rejected hypothesis tests in the social sciences than in physical sciences (Fanelli 2010).

- Published null results are disappearing over time, in all disciplines (Fanelli 2011).

- This is very unlikely to represent the true state of the universe.

- Data on the full set of experiments run with a large survey shows strong results are 40pp more likely to be published, and 60pp more likely to be written up. The file drawer problem is massive. (Franco, Malhotra, Simonovits 2014—see tomorrow)

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## Publication Bias

If we only write up/publish significant results, and we have no record of all the insignificant results, we have no way to tell if our 'significant' results are real, or if they're the 5% we should expect due to randomness.

# P-Hacking

- Also called fishing, researcher degrees of freedom, data mining, data massaging, or specification searching.

- Definition: flexibility in data analysis allows portrayal of *anything* as below an arbitrary p-value threshhold; significance loses its meaning.

- Not something only evil people do. It can be subconcious—humans are really good at motivated reasoning.

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Does this actually happen? (JLP 2011)

1. In a paper, failing to report all of a study's dependent measures
2. Deciding whether to collect more data after looking to see whether the results were significant
3. In a paper, failing to report all of a study's conditions
4. Stopping collecting data earlier than planned because one found the result that one had been looking for
5. In a paper, "rounding off" a $p$ value (e.g., reporting that a $p$ value of .054 is less than .05)
6. In a paper, selectively reporting studies that "worked"
7. Deciding whether to exclude data after looking at the impact of doing so on the results
8. In a paper, reporting an unexpected finding as having been predicted from the start
9. In a paper, claiming that results are unaffected by demographic variables (e.g., gender) when one is actually unsure (or knows that they do)
10. Falsifying data

1. In a paper, failing to report all of a study's dependent measures — 63.4

2. Deciding whether to collect more data after looking to see whether the results were significant — 55.9

3. In a paper, failing to report all of a study's conditions — 27.7

4. Stopping collecting data earlier than planned because one found the result that one had been looking for — 15.6

5. In a paper, "rounding off" a $p$ value (e.g., reporting that a $p$ value of .054 is less than .05) — 22.0

6. In a paper, selectively reporting studies that "worked" — 45.8

7. Deciding whether to exclude data after looking at the impact of doing so on the results — 38.2

8. In a paper, reporting an unexpected finding as having been predicted from the start — 27.0

9. In a paper, claiming that results are unaffected by demographic variables (e.g., gender) when one is actually unsure (or knows that they do) — 3.0

10. Falsifying data — 0.6

| | Admission rate | Defensibility rate |
|---|---|---|
| 1. In a paper, failing to report all of a study's dependent measures | 63.4 | 1.84 (0.39) |
| 2. Deciding whether to collect more data after looking to see whether the results were significant | 55.9 | 1.79 (0.44) |
| 3. In a paper, failing to report all of a study's conditions | 27.7 | 1.77 (0.49) |
| 4. Stopping collecting data earlier than planned because one found the result that one had been looking for | 15.6 | 1.76 (0.48) |
| 5. In a paper, "rounding off" a p value (e.g., reporting that a p value of .054 is less than .05) | 22.0 | 1.68 (0.57) |
| 6. In a paper, selectively reporting studies that "worked" | 45.8 | 1.66 (0.53) |
| 7. Deciding whether to exclude data after looking at the impact of doing so on the results | 38.2 | 1.61 (0.59) |
| 8. In a paper, reporting an unexpected finding as having been predicted from the start | 27.0 | 1.50 (0.60) |
| 9. In a paper, claiming that results are unaffected by demographic variables (e.g., gender) when one is actually unsure (or knows that they do) | 3.0 | 1.32 (0.60) |
| 10. Falsifying data | 0.6 | 0.16 (0.38) |

# P-Hacking

- Do people actually do this? (Previous—John, Loewenstein, Prelec 2011)
- Listening to the Beatles' "When I'm Sixty-Four" makes you younger. (Simmons, Nelson, Simonsohn 2011)
- Inordinately many .049 p-values, and indordinately few .051's. (Brodeur et al 2013)
- Political ideologues literally see in black and white (Nosek, Spies, Motyl 2012)

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## Solutions

- Publication Bias—Registration
- P-Hacking—Pre-Analysis Plans

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- P-Hacking—Pre-Analysis Plans

# Pre-Analysis Plan

- Often part of a registration

- From 3ie: "A pre-analysis plan is a detailed description of
  the analysis to be conducted that is written in advance of
  seeing the data on impacts of the program being
  evaluated. It may specify hypotheses to be tested, variable
  construction, equations to be estimated, controls to be
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# Glennerster, Takavarasha Suggestions

1. the main outcome measures,

2. which outcome measures are primary and which are secondary,

3. the precise composition of any families that will be used for mean effects analysis,

4. the subgroups that will be analyzed,

5. the direction of expected impact if we want to use a one-sided test, and

6. the primary specification to be used for the analysis.

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http://blogs.worldbank.org/impactevaluations/
a-pre-analysis-plan-checklist

1. Description of the sample to be used in the study
2. Key data sources
3. Hypotheses to be tested throughout the causal chain
4. Specify how variables will be constructed
5. Specify the treatment effect equation to be estimated
6. What is the plan for how to deal with multiple outcomes and multiple hypothesis testing?
7. Procedures to be used for addressing survey attrition
8. How will the study deal with outcomes with limited variation?
9. If you are going to be testing a model, include the model
10. Remember to archive it

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2. Authors must collect at least 20 observations per cell or else provide a compelling cost-of-data-collection justification.

3. Authors must list all variables collected in a study.

4. Authors must report all experimental conditions, including failed manipulations.

5. If observations are eliminated, authors must also report what the statistical results are if those observations are included.

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## Examples

Wide range of when to write and how detailed to make the plan. At the extreme level of detail you would have your entire code already written before you got any data.

- J-PAL Hypothesis Registry (11), see http://www.povertyactionlab.org/Hypothesis-Registry

- AEA Registry has relatively few, plentiful in EGAP.

- Casey, Glennerster, Miguel, "Reshaping Institutions: Evidence on Aid Impacts Using a Pre-Analysis Plan" *QJE* 2012. (Paper,Plan)

    - Government-sponsored program.
    - Broad program (Community Driven Development)
    - Broad outcomes (trust, public goods, public services, community groups, information, participation, crime, welfare, attitudes)

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| Outcome variable | (1) Mean for controls | (2) Treatment effect |
|---|---|---|
| Panel A: GoBifo "weakened" institutions | | |
| Attended meeting to decide what to do with the tarp | 0.81 | –0.04+ |
| Everybody had equal say in deciding how to use the tarp | 0.51 | –0.11+ |
| Community used the tarp (verified by physical assessment) | 0.90 | –0.08+ |
| Community can show research team the tarp | 0.84 | –0.12* |
| Respondent would like to be a member of the VDC | 0.36 | –0.04* |
| Respondent voted in the local government election (2008) | 0.85 | –0.04* |
| Panel B: GoBifo "strengthened" institutions | | |
| Community teachers have been trained | 0.47 | 0.12+ |
| Respondent is a member of a women's group | 0.24 | 0.06** |
| Someone took minutes at the most recent community meeting | 0.30 | 0.14* |
| Building materials stored in a public place when not in use | 0.13 | 0.25* |
| Chiefdom official did not have the most influence over tarp use | 0.54 | 0.06* |
| Respondent agrees with "Responsible young people can be good leaders" and not "Only older people are mature enough to be leaders" | 0.76 | 0.04* |
| Correctly able to name the year of the next general elections | 0.19 | 0.04* |

# Conclusion

- Neither registration nor pre-analysis plans are limited to randomized trials.

- Spectrum from confirmatory to exploratory research, all has value.

- I'd just like know which research is which.

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