

Protect Yourself from  $p$ -Hacking:  
7 Things to Do to  
Avoid Committing Scientific Malpractice

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**Abstract**

Replication crisis

LaCour scandal

p-Hacking

malpractice not always so blatant or intentional: confirmation bias, garden of forking paths (see [http://www.stat.columbia.edu/~gelman/research/unpublished/p\\_hacking.pdf](http://www.stat.columbia.edu/~gelman/research/unpublished/p_hacking.pdf))

Introduce Newman, Johnston, and Lown (2015*a*), perhaps noting the press attention it has received (e.g., <http://www.psmag.com/health-and-behavior/five-studies-bernie-sanders-says-the-rich-are-deranged>)

## 1 Ensure Reproducibility

Reproducibility as bare minimum for replication

script all work

packrat and checkpoint packages in R; version command in Stata

quote Newman, Johnston, and Lown (2015*a*) replication materials

Table 1 and 2 cannot be reproduced exactly

Table 3 cannot be reproduced at all: more parameters than observations

## 2 Work in Public

Github as baseline

discuss preregistration as option, at least for some research designs

Newman, Johnston, and Lown (2015*a*) didn't do either of these things, of course

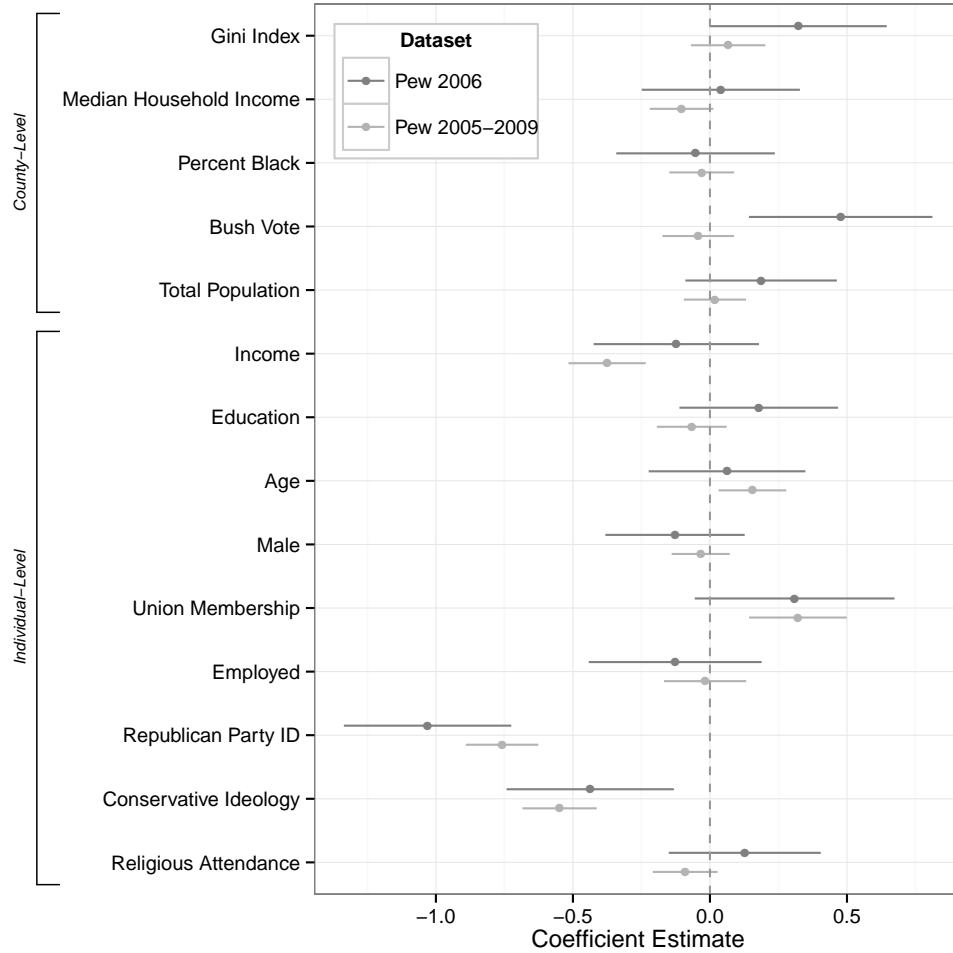
## 3 Examine All Available Data

examine as much relevant evidence as possible

discuss Figure 1

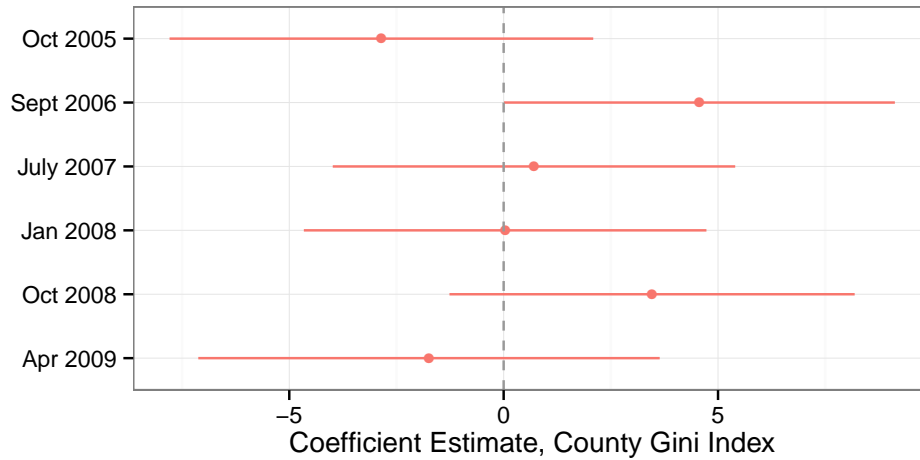
discuss Figure 2

Figure 1: Local Inequality and the Perception of America as Divided into ‘Haves’ and ‘Have-Nots’: Results Using All Available Data



*Notes:* Results from replications of the model presented in Table 2 of Newman, Johnston, and Lown (2015a) on the 2006 Pew survey analyzed in that article and on pooled data from the six Pew surveys that included the same item and were conducted in the time period the article examines. The statistically significant result for county income inequality in the 2006 survey presented in that article is not evident when all of the available data are examined.

Figure 2: Local Inequality and the Perception of America as Divided into ‘Haves’ and ‘Have-Nots’: Results Using Each Available Dataset



*Notes:* Results for county income inequality from replications of the model presented in Table 2 of Newman, Johnston, and Lown (2015*a*) on data from each of six available surveys conducted in the in the time period examined in that article. Of the six surveys, the only one that yields a statistically significant result is the 2006 survey presented in the article.

## 4 Use Consistent Measures

text

## 5 Wrangle Data with Care

need to be really careful: double-check! Also need to be transparent.

merging: data on Bush share of vote don't match

As shown in Table ??, a quick glance at the first handful of counties analyzed in Table 1 and Table 2 (Newman, Johnston, and Lown 2015*a*) reveals that something went wrong when information on the Bush share of the vote in the 2004 election was merged into the datasets: across all counties examined in both tables, fewer than 10% have matching data, even when rounded to two decimal places. (The data analyzed in Table 1 correspond

that available from other sources, so it appears that it is the Table 2 dataset that is problematic.)

Table 1: Mismatched Data on Bush Vote, from Replication Data

County	Table 1 Data	Table 2 Data
Baldwin, AL	0.76	0.79
Calhoun, AL	0.66	0.66
Chambers, AL	0.58	0.56
Cherokee, AL	0.65	0.65
Choctaw, AL	0.54	0.51
Clarke, AL	0.59	0.57

*Notes:* Newman, Johnston, and Lown (2015*b*) replication data on the share of the vote won by Bush in the 2004 presidential election. The first six counties, when listed alphabetically by state and county, are shown; they reveal that the data employed in Table 2 only occasionally matches that employed in Table 1, even when rounded to two digits. Overall, these data match for fewer than 10% of all counties.

coding and recoding: NJL’s five point party id scale collapses leaners and weak partisans (not weak and strong partisans, and not leaners and ‘true’ independents). Should really use the full seven point scale; no reason to throw away that information (or to deviate from common practice)

unemployment is mismeasured in 2005, 2007, and 2009 in Table 1 due to missing employ2 variable—all of those who are not working (students, retired, etc.) are coded as unemployed

## 6 Multiply Impute Missing Data

missing data should be multiply imputed (e.g., King et al. 2001)

church attendance—all missing are simply assigned “once or twice a month”

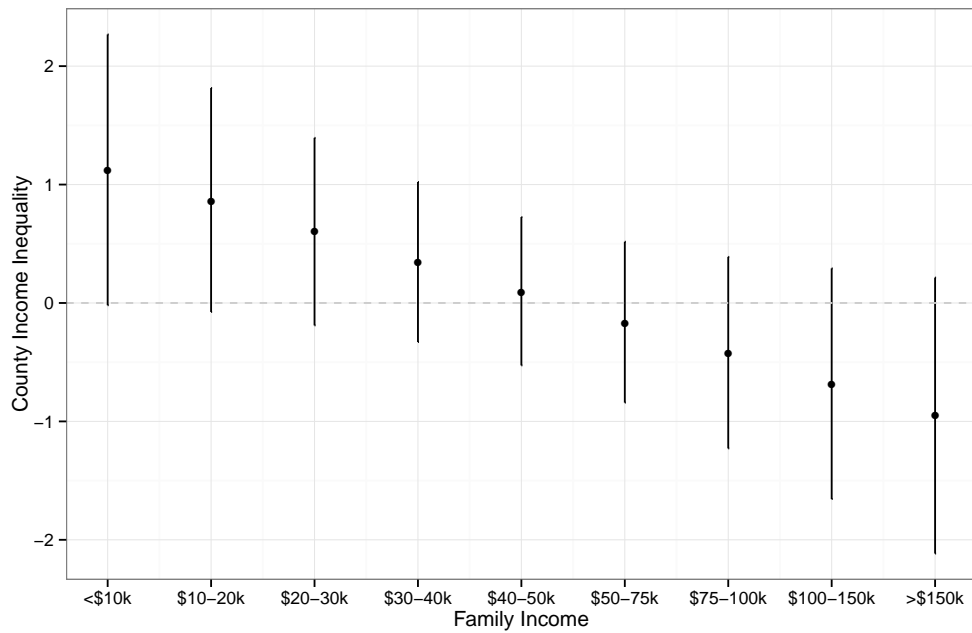
income, a variable of interest, is missing for over 10% of the sample, but values are mysteriously single-imputed (where did these values come from? they aren’t meaningful—they fall between categories)

ideology, partyid also single-imputed, it seems

## 7 Plot Interaction Terms

It has been well known for over a decade that models containing multiplicative interaction terms require particular care in interpretation (see, e.g., Golder 2003; Braumoeller 2004; Brambor, Clark, and Golder 2006; Kam and Franzese 2007).

Figure 3: Logit Coefficients of Local Income Inequality by Respondent Income: Table 1, Model 1, From Replication Data



*Notes:* The coefficient for county income inequality fails to reach statistical significance for any observed level of respondent family income.

## References

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