

### Ownership Diversity and Innovation: Using Split Sample Design to Generate More Credible Findings

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Essentially, all models are wrong, but some are useful.

--George Box



#### Outline

- Pr(false discovery) increases considerably if the same data are used for specification and hypothesis testing
- Study of diversity particularly vulnerable to false discovery
- Splitting data into exploratory and confirmatory samples can restore validity of hypothesis tests
- Protocol for testing how ownership diversity is associated with innovation in the Annual Business Survey (ABS)
  - Diversity measure selection: axiomatic or inductive?
  - Sample split, sample isolation, pass through, Registered Report
  - Hypothesis testing and submitting final manuscript



# Replicability Crisis: Publication Bias, Rush to Regression, and Canned Test Statistics

- Null hypothesis statistical testing makes statistical inference dependent on positive results
- Searches for significance facilitated by low cost of estimation
- p-value of single coefficient estimate invalid if preceded by specification tests
- No transparency in specification, and hypothesis testing has resulted in the proliferation of robustness checks
- Faith in robustness checks: what you see is all there is



# How specification testing degrades hypothesis test statistics

A significance level of  $\alpha = 0.05$  from statistical software assumes an unselected model.

If you have a set of *k* null hypotheses and all are true, then the probability of one erroneous rejection is

 $Pr(\text{at least 1 rejection}) = 1 - Pr(\text{all accepted}) = 1 - (1 - \alpha)^k$ 

Number of Hypotheses	Nominal p-value	True p-value
1	0.05	0.05
3	0.05	0.14
6	0.05	0.26
12	0.05	0.46



# Statistical analysis of diversity particularly vulnerable to false discovery

- "Essentially, all models are wrong, some models may be useful for reinforcing your priors"
- There are numerous ways to measure diversity as something other than homophily
- In ABS, at least 7 principal owner attributes can capture ownership diversity
- Combinatorial measures using multiple attributes further increase hypotheses that could be tested



# Split sample design: Restoring transparency to specification and hypothesis testing

- 1. 35% of dataset to discover potentially useful models
- 2. Document the potentially useful models in a public Registered Report
- 65% of dataset used for hypothesis testing, generating valid test statistics
- Apply false discovery (FDR) and family-wise error rate (FWER) correction for assessing significance across multiple comparisons
- 5. Publish full set of hypothesis tests



## Selecting a diversity index axiomatically

HOMOPHILY AXIOM: Given other things, all owners belonging to the same group must result in the lowest diversity measure value.

**FRACTIONALIZATION AXIOM:** Given other things, an increase in the number of groups must increase the diversity measure value.

**TEAM SIZE AXIOM:** Given other things, larger ownership teams not demonstrating homophily must increase the diversity measure value relative to smaller ownership teams.

CONCENTRATION OF OWNERSHIP AXIOM: Given other things, ownership concentrated in one member of the team must reduce the diversity measure value relative to ownership that is more equally distributed among team members.



### Ownership fractionalization index

Derived from the ethnolinguistic fractionalization index (ELF):

$$ELF = 1 - \sum_{i=1}^{n} p_i^2$$

where *p* is the population share of *n* groups
Invariant to population size so violates TEAM SIZE AXIOM

A minor modification of the ELF satisfies all four axioms

Ownership Fractionalization Index

$$=1-\sum_{i=1}^{n}p_{i}^{n}$$

where *p* represents the ownership share of the *i*<sup>th</sup> owner, and *o* is the number of owners

### Split sample design protocol

- Anderson and Magruder (2017) NBER Working Paper
- Monte Carlo simulations suggest exploratory share of 0.35
- FSRDC research makes isolation of confirmatory share credible
- Deciding on threshold for passing estimates to confirmation
  - t-statistic of 2 given large size of exploratory sample, and
  - minimum effect size (Cohen's d ≥ 0.2 or odds ratio ≥ 1.44)
- Estimate New-to-market Innov = f(OF<sub>i</sub>, Firm Size, NAICS)
- Publish results in Center for Open Science Registered Report



### Split sample design protocol (continued)

- Access to confirmatory sample granted only after Registered Report published
- Hypotheses from Registered Report tested, adjusting test statistics with FDR and FWER correction
  - FDR:  $p_i$  < 0.05\*(i/m) where p-values ranked from smallest to largest ( $p_m$ )
  - FWER:  $p_i$  < 0.05/m > w/12 tests only p < 0.00416 significant
- Publish all statistics from confirmatory tests and include link to Registered Report

### Limitations of split sample design

- FSRDC proposals for single use of confidential data ideally suited to split sample design
- Multiple analyses using the same data vulnerable to overfitting and false discovery:
  - Dwork, et al. 2015. The reusable holdout: Preserving validity in adaptive data analysis, Science.
  - Access the holdout set only via a differentially private mechanism
  - Ideas are tested against aggregate information, whereas individual dataset components remain confidential



# Contentious questions raise cost of false discovery

- What kinds of ownership diversity enable or hinder innovation?
- The focus on Type I errors may exacerbate Type II errors:
  - Large ABS sample size lessens concerns wrt statistical power
  - No inferences wrt negative results in confirmatory analysis
- Annual collection of ABS refreshes data required for sequential investigation of phenomena
- FSRDC can provide strong evidence that claims of restricted access to confirmatory data are valid





Thank you!
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