Is anything better than nothing? Data quality under compulsory survey participation

Samer Kherfi
Department of Economics
American University of Sharjah

Research Symposium in Finance and Economics Dubai - April 24, 2025

■ Evidence-based decisions

- Evidence-based decisions
- Mandated collection of student feedback on course and instructor effectiveness

- Evidence-based decisions
- Mandated collection of student feedback on course and instructor effectiveness
- Used in contract decisions: Renewal and promotion

- Evidence-based decisions
- Mandated collection of student feedback on course and instructor effectiveness
- Used in contract decisions: Renewal and promotion
- Major role in holding institutions accountable and shaping public opinion (rankings based on academic reputation and other metrics)

- Evidence-based decisions
- Mandated collection of student feedback on course and instructor effectiveness
- Used in contract decisions: Renewal and promotion
- Major role in holding institutions accountable and shaping public opinion (rankings based on academic reputation and other metrics)
- What if data quality is declining when data is playing an increasingly important role

- Evidence-based decisions
- Mandated collection of student feedback on course and instructor effectiveness
- Used in contract decisions: Renewal and promotion
- Major role in holding institutions accountable and shaping public opinion (rankings based on academic reputation and other metrics)
- What if data quality is declining when data is playing an increasingly important role
- Apart from questioning the respondent's ability to produce good data, does the respondent always have the incentive to deliver good data?

 A decision-making strategy that aims for a satisfactory (good enough) solution, rather than the optimal solution

- A decision-making strategy that aims for a satisfactory (good enough) solution, rather than the optimal solution
- Survey responses requested from those who are not motivated to optimally respond (busy - no time, distracted - no attention, desire to conserve cognitive effort: understand the question, think about it, then articulate and convey an answer) - Krosnik (1991)

- A decision-making strategy that aims for a satisfactory (good enough) solution, rather than the optimal solution
- Survey responses requested from those who are not motivated to optimally respond (busy - no time, distracted - no attention, desire to conserve cognitive effort: understand the question, think about it, then articulate and convey an answer) - Krosnik (1991)
- Non-response versus sub-optimal response

- A decision-making strategy that aims for a satisfactory (good enough) solution, rather than the optimal solution
- Survey responses requested from those who are not motivated to optimally respond (busy no time, distracted no attention, desire to conserve cognitive effort: understand the question, think about it, then articulate and convey an answer) Krosnik (1991)
- Non-response versus sub-optimal response
 What if the response is mandatory, and is solicited during the last two weeks of classes, close to exam dates and project deadlines?

■ Sub-optimal response => poor data quality

- Sub-optimal response => poor data quality
- Aspects/metrics of satisficing:

- Sub-optimal response => poor data quality
- Aspects/metrics of satisficing:

Selecting the first satisfactory option or argument presented

- Sub-optimal response => poor data quality
- Aspects/metrics of satisficing:
 Selecting the first satisfactory option or argument presented
 Not differentiating between response options (providing a straight-line response)

- Sub-optimal response => poor data quality
- Aspects/metrics of satisficing:

Selecting the first satisfactory option or argument presented

Not differentiating between response options (providing a straight-line response)

Skipping items

- Sub-optimal response => poor data quality
- Aspects/metrics of satisficing:

Selecting the first satisfactory option or argument presented

Not differentiating between response options (providing a straight-line response)

Skipping items

Rushing the answers (a short gap between survey start and end times)

Strong definitions of satisficing (conservative estimate):

Strong definitions of satisficing (conservative estimate):
Zero character length of the answer to an open-ended question

Strong definitions of satisficing (conservative estimate):
 Zero character length of the answer to an open-ended question
 Zero standard deviation by selecting the same response choice

- Strong definitions of satisficing (conservative estimate):
 Zero character length of the answer to an open-ended question
 Zero standard deviation by selecting the same response choice
- Two questions to answer: What is the extent of satisficing? Are the responses sensitive to satisficing behavior?

Extent

Table 1: (Strong) Satisficing Metrics

	Percent of
Metric	Respondents
Non-differentiation	
(Lickert scale questions, 5 options)	
Q1-Q10 (course)	31.4
Q11-Q21 (instructor)	43.4
Q1-Q21	26.2
Skipping (open-ended questions)	
Q22 (suggestions to instructor)	80.2
Q23 (experience with instructor)	77.2
Q24 (examples used)	84.9
Q25 (suggestions for course improvement)	87.1
Q22-Q25	73.2
Q1-Q25 (non-differentiation and skipping)	21.0

Extent

Table 2: Correlation Matrix

	length0_q22	length0_q23	length0_q24	length0_q25	sd0_110	sd0_1121
length0_q22	1.0000					
length0_q23	0.6963*	1.0000				
length0_q24	0.6709*	0.7334*	1.0000			
length0_q25	0.6325*	0.6141*	0.6974*	1.0000		
sd0_110	0.0908*	0.0950*	0.0679*	0.0671*	1.0000	
sd0_1121	0.0725*	0.0390*	0.0325*	0.0585*	0.5716*	1.0000

^{*} Significant at 0.05

Table 2a: Propensity to Satisfice - Gender

sd0_1121	0	1
Men	53.7%	46.3%
Women	59.2%	40.8%
Total	56.6%	43.4%

Chi2(1) = 95.96 (P< 0.001)

Impact

Table 3: Impact of Satisficing on Satisfaction -1

Satisfaction	Non-Differentiators	Differentiators
Strongly Disagree	1.3	3.3
Disagree	0.4	5.7
Neutral	5.0	18.1
Agree	19.1	37.9
Strongly Agree	73.8	34.4
	0.4	0.7
Total	100.0	100.0

Chi2(5): p<0.01

Impact of Satisficing on Satisfaction -2

	Item	Obs	Mean	Std. dev.
sd0_1121==1	q21	13,668	4.64*	0.72
sd0_1121==0	q21	17,775	3.95*	1.03
sd0_1121==1	q11q21	13,719	4.64**	0.72
sd0_1121==0	q11q21	17,833	4.06**	0.69

^{*/**}Difference is statistically significant (p<0.01)

Impact

Table 3 (cont.)
Impact of Satisficing on Satisfaction -3 (Skipping)

	Q21	Q11-Q21
length0_q22=0	4.110	4.238
length0_q22=1	4.286	4.332
Difference	-0.176*	0948*
length0_q23=0	4.258	4.348
length0_q23=1	4.249	4.304
Difference	0.092	.0440*
length0_q24=0	4.270	4.352
length0_q24=1	4.247	4.307
Difference	0.023	.0449*
length0_q25=0	4.269	4.247
length0_q25=1	4.129	4.324
Difference	-0.140*	0767*

^{*}Significant p<0.01

Satisficing is prevalent

- Satisficing is prevalent
- Satisficers are different from non-satisficers

- Satisficing is prevalent
- Satisficers are different from non-satisficers
- However, bivariate results can be spurious

- Satisficing is prevalent
- Satisficers are different from non-satisficers
- However, bivariate results can be spurious
- Reverse causation: Impact of satisfaction on satisficing; happy/unhappy students satisfice

- Satisficing is prevalent
- Satisficers are different from non-satisficers
- However, bivariate results can be spurious
- Reverse causation: Impact of satisfaction on satisficing;
 happy/unhappy students satisfice
 satisficing => satisfaction and satisficing <= satisfaction
 (identification and causality as opposed to correlations)

- Satisficing is prevalent
- Satisficers are different from non-satisficers
- However, bivariate results can be spurious
- Reverse causation: Impact of satisfaction on satisficing; happy/unhappy students satisfice satisficing => satisfaction and satisficing <= satisfaction (identification and causality as opposed to correlations) Need for randomization (Barge and Gehlbach, 2012)

■ Student evaluation of teaching (SET) data: 31,610 responses

Student evaluation of teaching (SET) data: 31,610 responses
 18 periods (semesters)

Student evaluation of teaching (SET) data: 31,610 responses

18 periods (semesters)

Percent female: 52.5%

Student evaluation of teaching (SET) data: 31,610 responses

18 periods (semesters)

Percent female: 52.5%

8 courses

Data

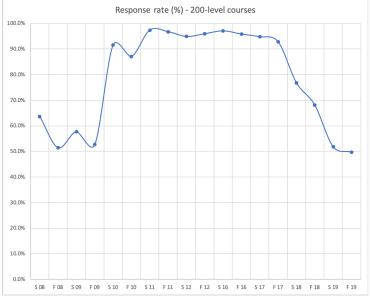
■ Student evaluation of teaching (SET) data: 31,610 responses

18 periods (semesters)

Percent female: 52.5%

8 courses

137 instructors



copy.png

Hypothesis: Satisficing is stronger when survey participation is compulsory.

Hypothesis: Satisficing is stronger when survey participation is compulsory.

Higher participation generates more data. Satisficing generates poorer data.

Hypothesis: Satisficing is stronger when survey participation is compulsory.

Higher participation generates more data. Satisficing generates poorer data.

Policy change twice! S08-F09 and S18-F19 (8 terms) are periods of optional survey participation. Forced participation (next semester registration or graduation are conditional on participation): S10-F17

Hypothesis: Satisficing is stronger when survey participation is compulsory.

Higher participation generates more data. Satisficing generates poorer data.

Policy change twice! S08-F09 and S18-F19 (8 terms) are periods of optional survey participation. Forced participation (next semester registration or graduation are conditional on participation): S10-F17 Policy rule as an instrument: Correlated with satisficing for reasons unrelated to satisfaction (exogenous variation in satisficing that affects satisfaction).

Hypothesis: Satisficing is stronger when survey participation is compulsory.

Higher participation generates more data. Satisficing generates poorer data.

Policy change twice! S08-F09 and S18-F19 (8 terms) are periods of optional survey participation. Forced participation (next semester registration or graduation are conditional on participation): S10-F17 Policy rule as an instrument: Correlated with satisficing for reasons unrelated to satisfaction (exogenous variation in satisficing that affects satisfaction).

The intensity of satisficing jumps when the survey is compulsory. Just before or after the rule change, other characteristics uncontrolled for (ex.: CGPA) are unchanged.

OLS

Linear regress	ion						
regress q1121	. male gender	_inter length0_q22	i.crs i.fid, vc	e(robust)			
Number of obs = 31,552							
R-squared= 0.	1002						
Root MSE= 0.	72179						
Dependant va	riable: q11q21	1					
	Coefficient	Robust std. err	Z	P> z	[95%	6 C I]	
male	0.037	0.018	2.080	0.038	0.002	0.072	
gender_inter	-0.019	0.020	-0.950	0.343	-0.059	0.020	
length0_q22	0.086	0.011	7.830	0.000	0.065	0.108	
_cons	4.445	0.150	29.610	0.000	4.151	4.739	
Instructor and	course fixed	effects are omitted	i				

x.png

Instrumental variable

х.	png
	Inst

Instrumental	variables	2SLS regression	

ivregress 2sls q11q21 male gender_inter (length0_q22 = option) i.crs i.fid, vce(robust)

Number of obs= 31,552

R-squared= Root MSE=0.93295

Dependant variable: q11q21

	Coefficient	Robust std. err	Z	P> z	[95% C I]		Ī
length0_q22	-1.411	0.229	-6.170	0.000	-1.860	-0.963	
male	0.070	0.024	2.950	0.003	0.024	0.116	Ī
gender_inter	-0.017	0.026	-0.670	0.505	-0.068	0.034	Ī
_cons	5.637	0.264	21.390	0.000	5.120	6.154	Ī
							Ī
Instructor and course fixed effects are omitted							

Corr(option, satisficing)=-0.06*

Corr(option, satisficing)=-0.06*

Change in sign and larger impact

Corr(option, satisficing)=-0.06*

Change in sign and larger impact

Either case: Impact is significant

Corr(option, satisficing)=-0.06*

Change in sign and larger impact

Either case: Impact is significant

Either case: Tade-off between quantity and quality when forced to

participate.

■ More observations are not always a blessing

- More observations are not always a blessing
- Next step: Student level to class level analysis: Avoid anonymity (CGPA, percent satisficing, standing, major)

- More observations are not always a blessing
- Next step: Student level to class level analysis: Avoid anonymity (CGPA, percent satisficing, standing, major)

Recall characteristics are similar in the neighborhood (before and after), which can be verified.

Thank you!

skherfi@aus.edu