



Comparing Errors from Non-coverage to Other Errors in a Mobile Web Survey

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Context

- Focus is on mobile Web surveys that require a smartphone to participate
 - necessary if want to use an app or other mobile-only feature
 - following Fuchs and Busse (2009) analysis of coverage errors in mobile-only Web surveys

Smartphone surveys

- Offer several advantages over PC Web data collection:
 - Auxiliary data collection opportunities
 - GPS; Bluetooth medical devices (De Nazell et al. 2013)
 - App with prompts
 - TV viewing; EMA; time-use (Sonck & Fernee 2013)
 - Sampling
 - RDD sample of mobile-Web users; send invite using text messaging



Smartphone surveys

- But not everyone has their own smartphone which is necessary to participate
 - At least 20% of U.S. adults don't own a smartphone
 - Owners are younger and better educated than non-owners (Smith, 2012)
 - Survey variables related to smartphone ownership may be biased due to non-coverage
- Owning a smartphone is necessary but not sufficient -- one also has to use (and know how to use) mobile browsers, etc.

Current Study

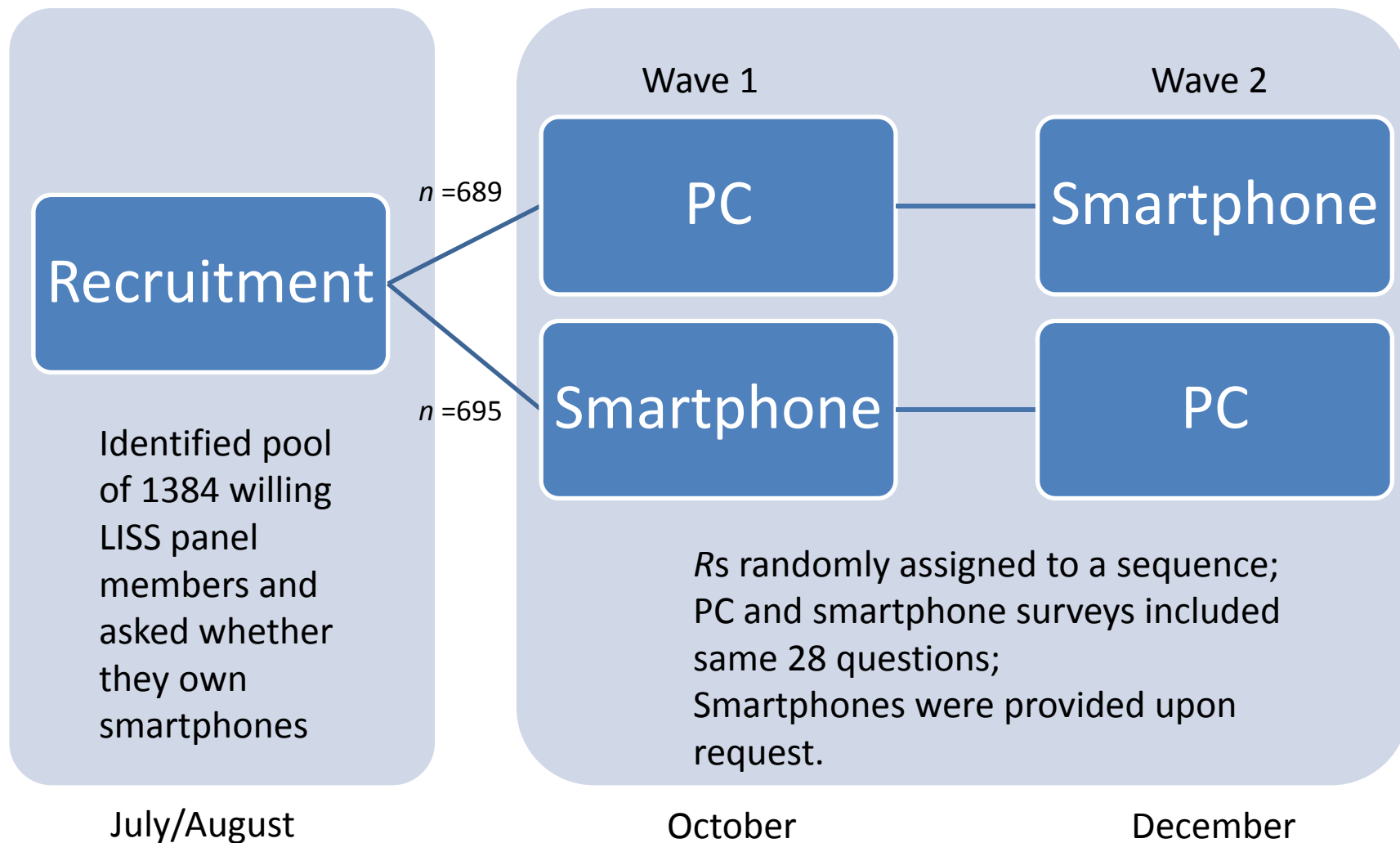
- Estimates non-coverage bias in a mobile Web survey for a selected group of survey variables
 - Usually don't know about uncovered population
 - Overcame this by conducting parallel PC Web survey
 - Know about those with and without smartphones; consider those without their own smartphones to be uncovered
- Compares non-coverage bias relative to other sources of error – nonresponse and measurement
- Considers the impact on nonresponse and measurement errors of providing phones to all respondents



Mode Experiment

July – December 2013

LISS panel: probability Web panel in the Netherlands





Implementation

- Sent two emails to panel members
 - instructions about device to use
 - “normal” invitation email with URL
- To start mobile survey
 - could click URL in email
 - Or if received smartphone for study, then could click on bookmark on the phone’s home screen
- Two email reminders near the end of month
- Normal cash incentive

Items

- Several measures that one might want to measure in smartphone survey:
 - health: exercise; social life; binge drinking
 - technology: tablet use; TV viewing
 - travel: drunk driving; eating out
- These variables lend themselves to passive measurement or apps with prompts
 - E.g., health survey + Bluetooth monitoring; TV viewing app; travel survey + GPS

Predictions: Non-coverage

Y: survey variable of interest

P: propensity to be covered

Z: other survey variables

- Relation between survey variable and coverage can be represented by 3 models

1) separate causes

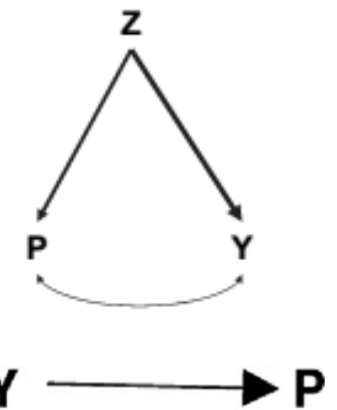
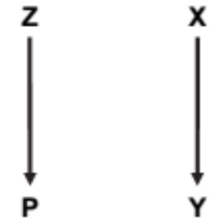
- Travel

2) common cause

- Health (affected by age, which affects smartphone adoption)

3) Survey variable cause

- Technology (social media use directly affects smartphone adoption)



Source: Groves (2006)



Assumptions

- No carryover effects: response in wave 2 not affected by wave 1
- Benchmark is full population of panel members who agreed to participate in the experiment and completed the PC Web survey; deviations from benchmark regarded as error
 - *relative* biases



Estimating Non-coverage Bias

- Compare mean based on benchmark ($\bar{y}_{\text{benchmark}}$) to mean based on those with smartphone (\bar{y}_{covered})
 - Coverage deviation: $B_{\text{cov}} = \bar{y}_{\text{benchmark}} - \bar{y}_{\text{covered}}$
 - Assumes self-reported ownership is accurate and smartphones are not shared

Estimating Nonresponse Bias

- Compare mean based on those with smartphone (\bar{y}_{covered}) to mean based on those with completed smartphone survey (\bar{y}_{respond})
 - Nonresponse deviation: $B_{\text{nr}} = \bar{y}_{\text{covered}} - \bar{y}_{\text{respond}}$

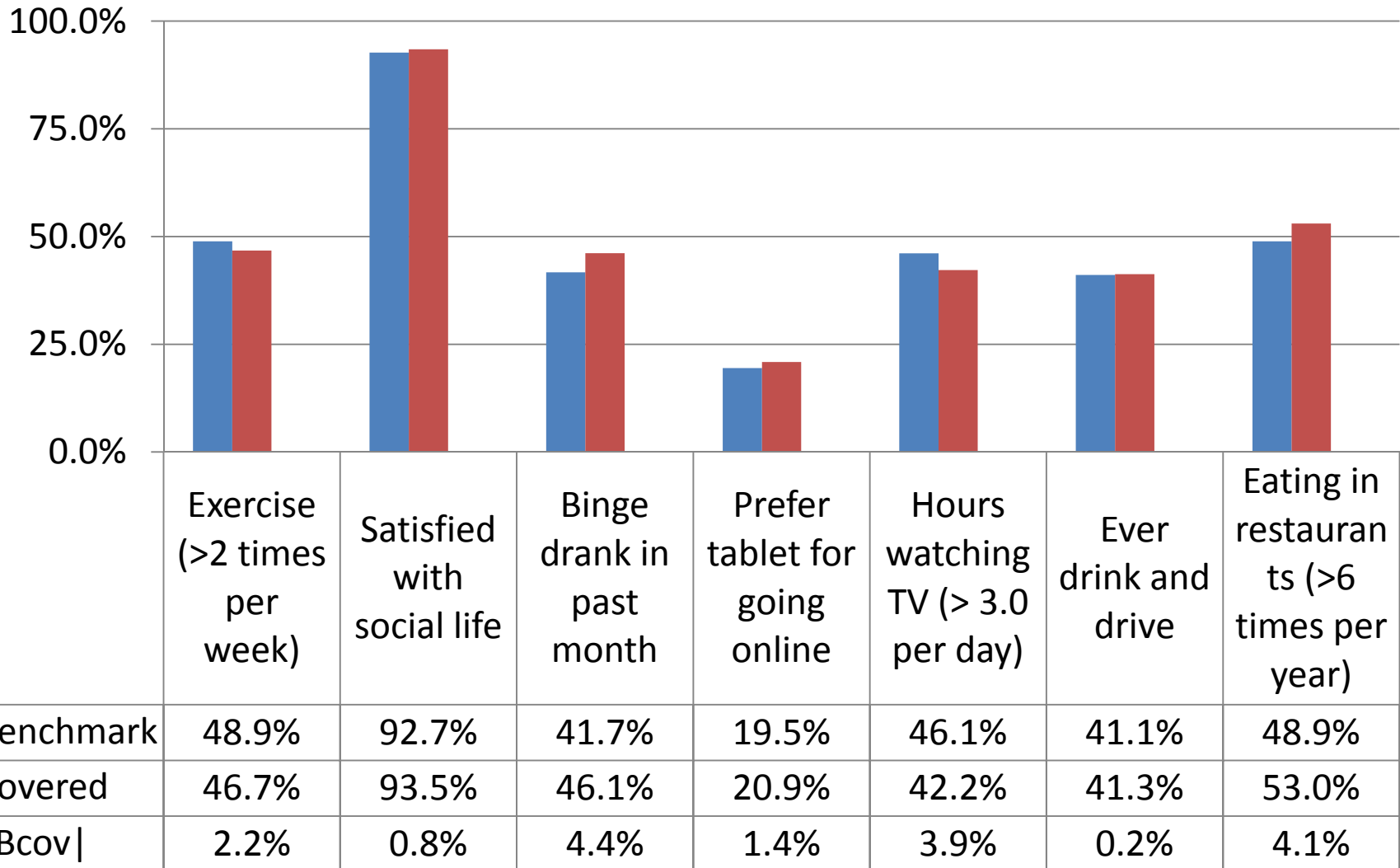
Estimating Measurement Bias

- Difference between means based on PC Web responses (\bar{y}_{respond}) and means based on smartphone survey responses ($\bar{y}_{\text{smartphone}}$) for the same respondents
 - Measurement deviation: $B_{\text{meas}} = \bar{y}_{\text{respond}} - \bar{y}_{\text{smartphone}}$

Subgroups

	Sample Size	Rate
Benchmark	1180	--
Covered	843	71.4%
Respond	614	72.8%
Respond (smartphone answers)	614	--

Results: Non-coverage errors



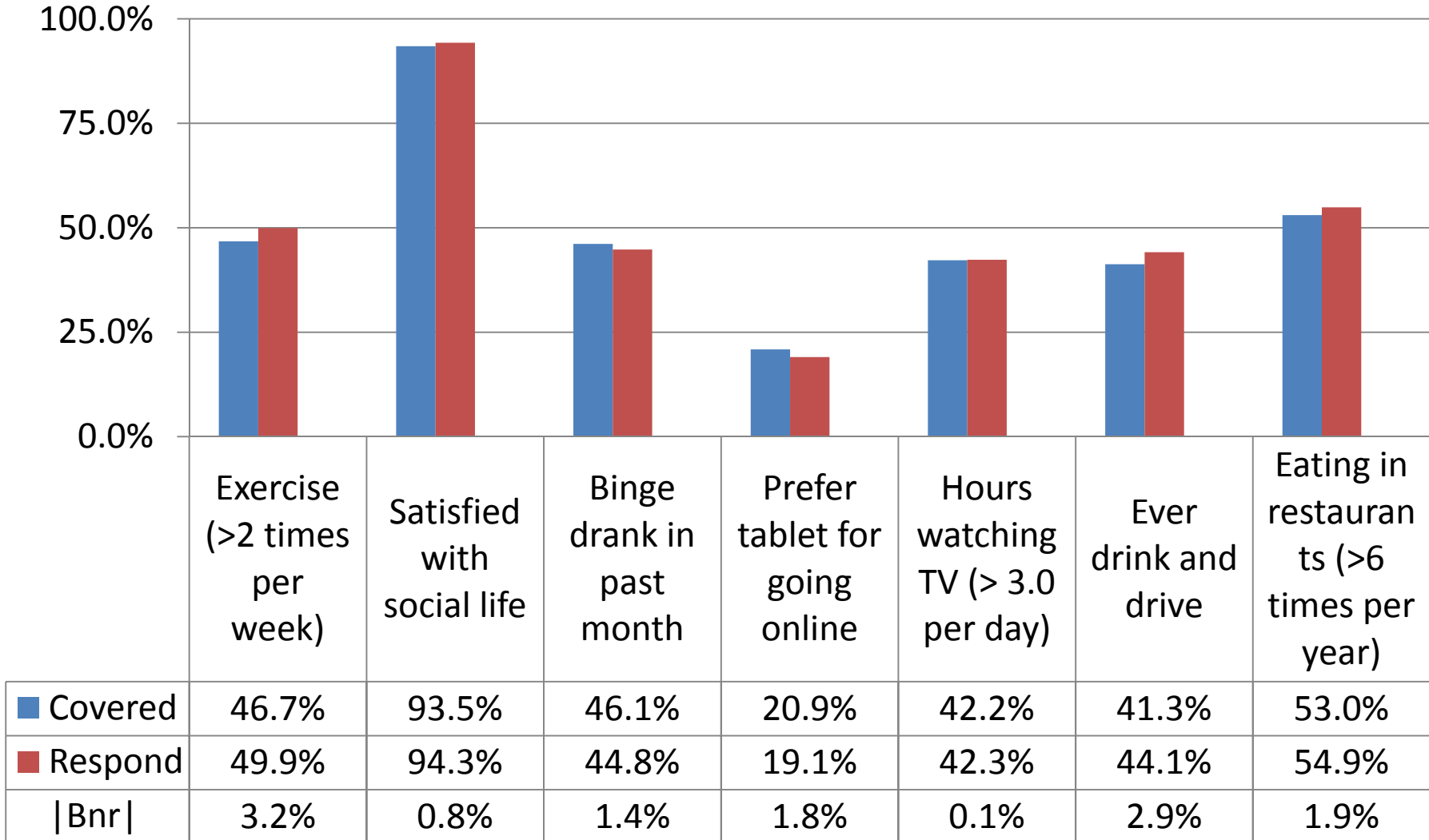
Average absolute $B_{cov} = 2.4\%$



Logistic regression to predict non-coverage

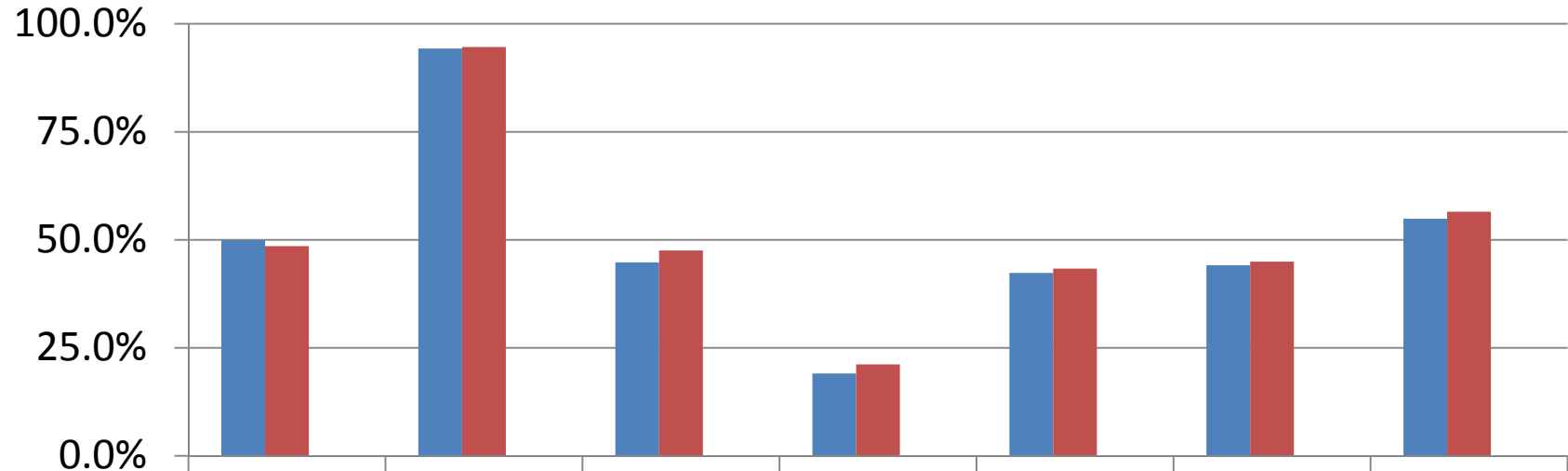
- DV: covered vs. not covered
- IVs: demographics commonly used for weighting adjustments (age, gender, education, marital status, and urbanicity); survey variables of interest
- Results: two survey variables are significant predictors
 - Tablet use; eating out
 - Appear to fall under Groves’s “survey variable cause” model

Nonresponse Errors



Average absolute B_{nr} = 1.7%

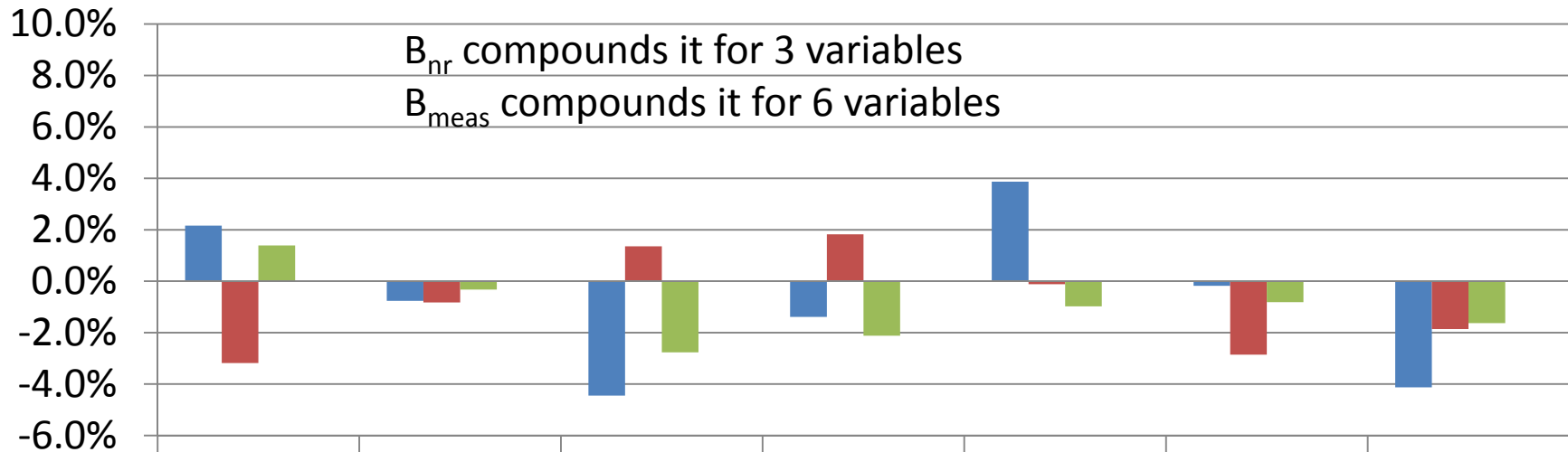
Measurement Errors



	Exercise (>2 times per week)	Satisfied with social life	Binge drank in past month	Prefer tablet for going online	Hours watching TV (> 3.0 per day)	Ever drink and drive	Eating in restaurants (>6 times per year)
■ Respond	49.9%	94.3%	44.8%	19.1%	42.3%	44.1%	54.9%
■ Respond: smartphone	48.5%	94.6%	47.6%	21.2%	43.3%	45.0%	56.5%
Bmeas	1.4%	0.3%	2.8%	2.1%	1.0%	0.8%	1.6%

Average absolute B_{meas} = 1.4%

Non-coverage errors not consistently offset by other sources of errors



	Exercise (>2 times per week)	Satisfied with social life	Binge drank in past month	Prefer tablet for going online	Hours watching TV (> 3.0 per day)	Ever drink and drive	Eating in restaurants (>6 times per year)
B_{cov}	2.2%	-0.8%	-4.4%	-1.4%	3.9%	-0.2%	-4.1%
B_{nr}	-3.2%	-0.8%	1.4%	1.8%	-0.1%	-2.9%	-1.9%
B_{meas}	1.4%	-0.3%	-2.8%	-2.1%	-1.0%	-0.8%	-1.6%
$ B_{total} $	0.4%	1.9%	5.9%	1.7%	2.8%	3.8%	7.6%

Average absolute $B_{total} = 3.4%$



Summary so far

- Absolute non-coverage bias larger than absolute nonresponse or measurement biases
 - 2.4% vs. 1.7% vs. 1.4% on average
- Two survey variables directly related to non-coverage
- Other errors don't consistently offset it



Estimating errors when phones are provided

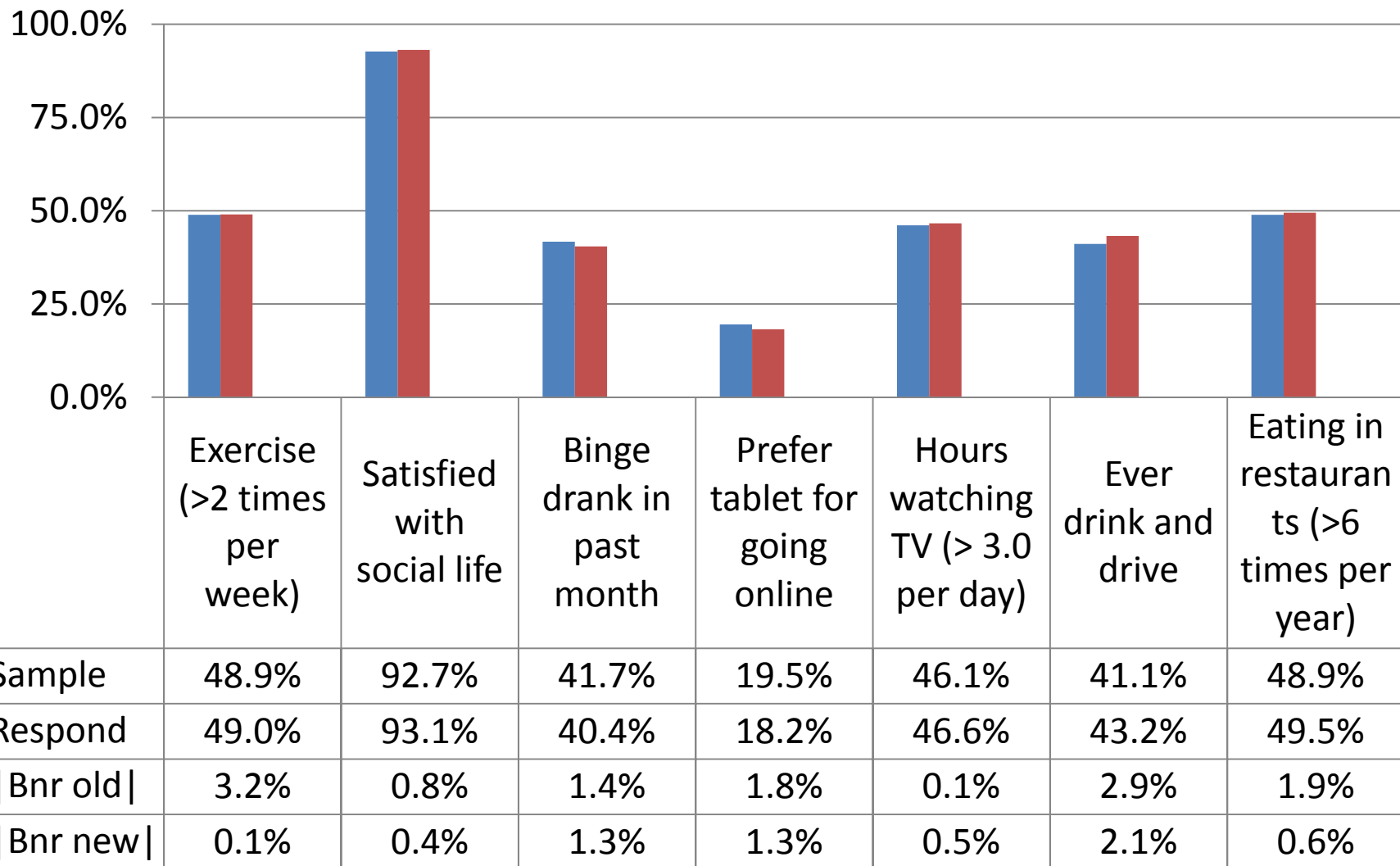
- Coverage issues are (by definition) eliminated
 - But additional respondents using unfamiliar phones could inflate other errors
 - unwillingness to accept or use loaner phone may increase nonresponse error

Subgroups (With Phones Provided)

	Old Sample Size	Old Rate	New Sample Size	New Rate
Benchmark	1180	--	1180	--
Covered	843	71.4%	1180	--
Respond	614	72.8%	918	77.8%
Respond (smartphone answers)	614	--	918	--



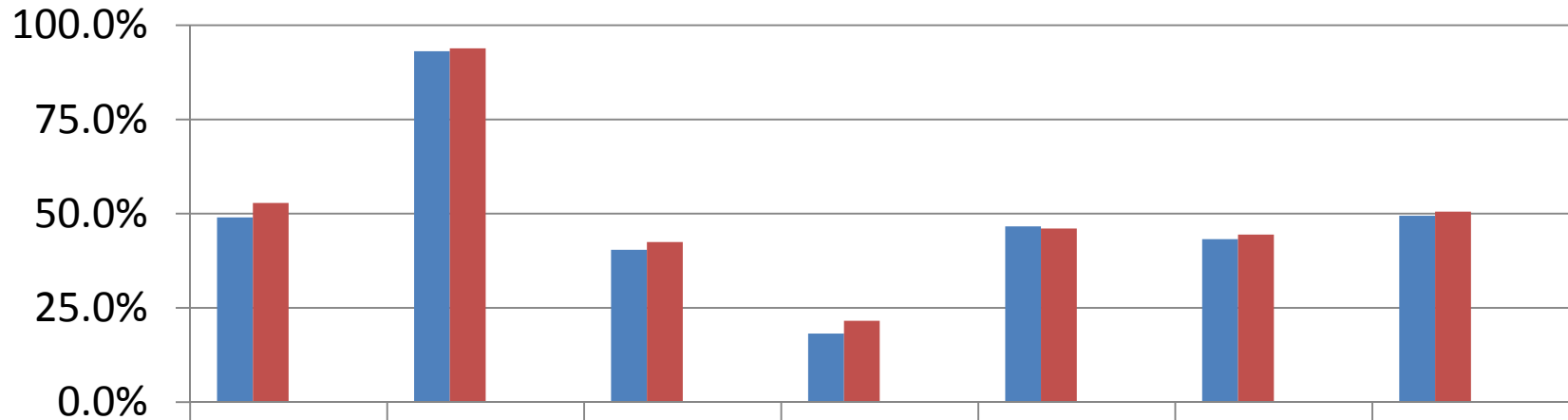
No negative effect on nonresponse errors



Average absolute B_{nr} old= 1.7%; **Average absolute B_{nr} new= 0.9%**



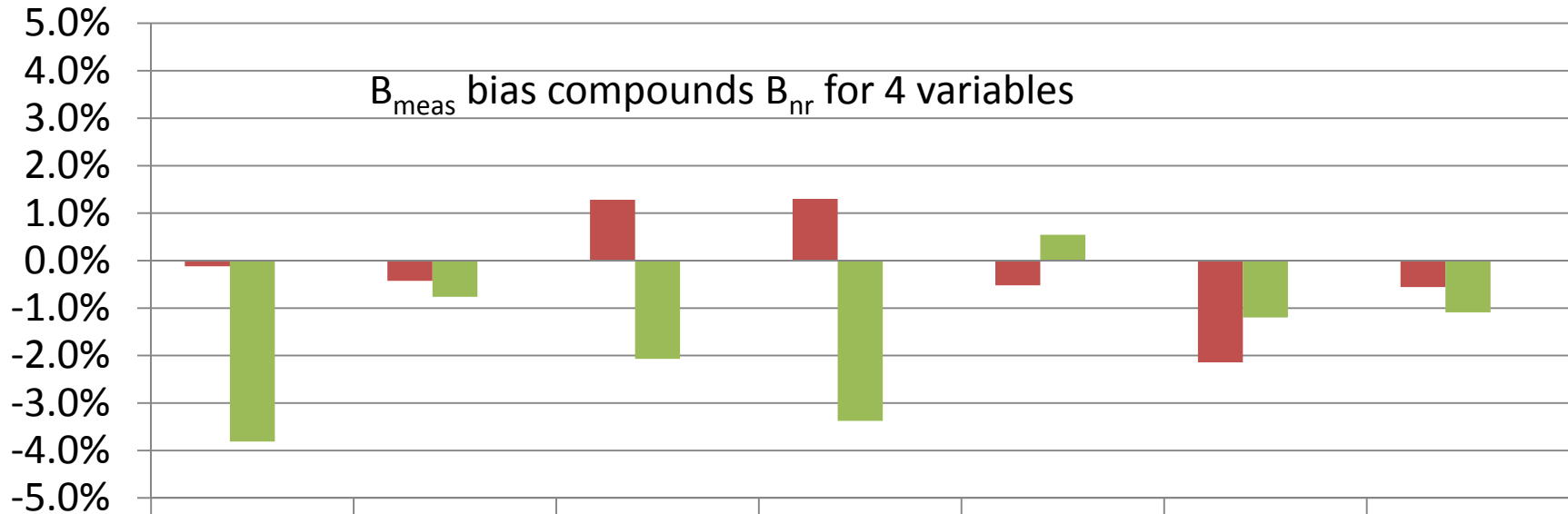
Modest increase in measurement errors



■ Respond	49.0%	93.1%	40.4%	18.2%	46.6%	43.2%	49.5%
■ Respond: smartphone	52.8%	93.9%	42.5%	21.6%	46.1%	44.4%	50.5%
B _{meas old}	1.4%	0.3%	2.8%	2.1%	1.0%	0.8%	1.6%
B _{meas new}	3.8%	0.8%	2.1%	3.4%	0.5%	1.2%	1.1%

Average $B_{meas\ old}$ = 1.4; Average $B_{meas\ new}$ = 1.8%

As before, biases don't consistently cancel out



	Exercise (>2 times per week)	Satisfied with social life	Binge drank in past month	Prefer tablet for going online	Hours watching TV (> 3.0 per day)	Ever drink and drive	Eating in restaurants (>6 times per year)
■ Bnr	-0.1%	-0.4%	1.3%	1.3%	-0.5%	-2.1%	-0.6%
■ Bmeas	-3.8%	-0.8%	-2.1%	-3.4%	0.5%	-1.2%	-1.1%
Btotal	3.9%	1.2%	0.8%	2.1%	0.0%	3.3%	1.6%

Average B_{total} old= 3.4% ; Average B_{total} new= 1.9%



Summary

- Large non-coverage error biases relative to other sources of error
 - Cannot be eliminated by weighting
 - Two survey variables directly related to non-coverage
 - Furthermore, non-coverage errors not consistently offset by other errors
- Suggests that limiting Web surveys to mobile Web users only is risky for general population surveys
- Unless phones are provided
 - Eliminates coverage issues without large effect on other sources of error



Remaining Questions

- For this analysis:
 - Are bias estimates statistically significant?
 - Plan is to use bootstrapping to add confidence intervals around bias estimates
 - Present results for more survey variables
- For future work:
 - error in a cross sectional survey
 - mean square error (MSE)
 - weighting adjustments
- Biases have different causes. What are they?
 - E.g., are the observed measurement errors due to socially desirable reporting, satisficing, or what?

THANK YOU!