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# THE PALGRAVE HANDBOOK OF SURVEY RESEARCH

Edited by  
David L. Vannette & Jon A. Krosnick



# The Palgrave Handbook of Survey Research

David L. Vannette • Jon A. Krosnick  
Editors

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ISBN 978-3-319-54394-9

ISBN 978-3-319-54395-6 (eBook)

<https://doi.org/10.1007/978-3-319-54395-6>

Library of Congress Control Number: 2017953116

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Cover illustration: Anna Berkut / Alamy Stock Photo

Printed on acid-free paper

This Palgrave Macmillan imprint is published by Springer Nature  
The registered company is Springer International Publishing AG  
The registered company address is: Gewerbestrasse 11, 6330 Cham, Switzerland



# Contents

## Section 1 Conventional Survey Research

- 1 Assessing the Accuracy of Survey Research** 3  
*Jon A. Krosnick*
- 2 The Importance of Probability-Based Sampling Methods for Drawing Valid Inferences** 7  
*Gary Langer*
- 3 Sampling for Single and Multi-Mode Surveys Using Address-Based Sampling** 13  
*Colm O'Muircheartaigh*
- 4 Evidence About the Accuracy of Surveys in the Face of Declining Response Rates** 19  
*Scott Keeter*
- 5 Sampling to Minimize Nonresponse Bias** 23  
*J. Michael Brick*
- 6 Cross-National Issues in Response Rates** 29  
*Vasja Vehovar and Koen Beullens*
- 7 Choosing a Mode of Survey Data Collection** 43  
*Roger Tourangeau*

<b>8</b>	<b>Mixed-Mode and Mixed-Device Surveys</b>	<b>51</b>
	<i>Edith Desiree de Leeuw and Vera Toepoel</i>	
<b>9</b>	<b>The Use and Effects of Incentives in Surveys</b>	<b>63</b>
	<i>Eleanor Singer</i>	
<b>10</b>	<b>Methods for Determining Who Lives in a Household</b>	<b>71</b>
	<i>Kathleen Targowski Ashenfelter</i>	
<b>11</b>	<b>Harmonization for Cross-National Comparative Social Survey Research: A Case Study Using the “Private Household” Variable</b>	<b>79</b>
	<i>Jürgen H. P. Hoffmeyer-Zlotnik and Uwe Warner</i>	
<b>12</b>	<b>Answering for Someone Else: Proxy Reports in Survey Research</b>	<b>87</b>
	<i>Curtiss Cobb</i>	
<b>13</b>	<b>Improving Question Design to Maximize Reliability and Validity</b>	<b>95</b>
	<i>Jon A. Krosnick</i>	
<b>14</b>	<b>Cognitive Interviewing in Survey Design: State of the Science and Future Directions</b>	<b>103</b>
	<i>Gordon Willis</i>	
<b>15</b>	<b>Survey Interviewing: Departures from the Script</b>	<b>109</b>
	<i>Nora Cate Schaeffer</i>	
<b>16</b>	<b>How to Improve Coding for Open-Ended Survey Data: Lessons from the ANES</b>	<b>121</b>
	<i>Arthur Lupia</i>	
<b>17</b>	<b>Applying Human Language Technology in Survey Research</b>	<b>129</b>
	<i>Mark Liberman</i>	
<b>18</b>	<b>Maintaining Respondent Trust and Protecting Their Data</b>	<b>135</b>
	<i>Roger Tourangeau</i>	
<b>19</b>	<b>Tackling Panel Attrition</b>	<b>143</b>
	<i>Peter Lynn</i>	

<b>20</b>	<b>Respondent Attrition Versus Data Attrition and Their Reduction</b>	<b>155</b>
	<i>Randall J. Olsen</i>	
<b>21</b>	<b>Best Practices for Creating Survey Weights</b>	<b>159</b>
	<i>Matthew DeBell</i>	
 <b>Section 2 Opportunities to Expand Data Collection</b>		
<b>22</b>	<b>New Kinds of Survey Measurements</b>	<b>165</b>
	<i>Carli Lessof and Patrick Sturgis</i>	
<b>23</b>	<b>The Role of Surveys in the Era of “Big Data”</b>	<b>175</b>
	<i>Mario Callegaro and Yongwei Yang</i>	
<b>24</b>	<b>Getting the Most Out of Paradata</b>	<b>193</b>
	<i>Frauke Kreuter</i>	
<b>25</b>	<b>Open Probability-Based Panel Infrastructures</b>	<b>199</b>
	<i>Marcel Das, Arie Kapteyn and Michael Bosnjak</i>	
<b>26</b>	<b>Collecting Interviewer Observations to Augment Survey Data</b>	<b>211</b>
	<i>Brady T. West</i>	
<b>27</b>	<b>“Just One More Thing”: Using Leave-Behind Measurement Supplements to Augment Survey Data Collection</b>	<b>217</b>
	<i>Michael W. Link</i>	
<b>28</b>	<b>Ecological Momentary Assessment in Survey Research</b>	<b>221</b>
	<i>Arthur A. Stone</i>	
<b>29</b>	<b>Biomarkers in Representative Population Surveys</b>	<b>227</b>
	<i>David Weir</i>	
<b>30</b>	<b>Measuring Past Events Using Calendar Aids</b>	<b>235</b>
	<i>Robert F. Belli</i>	
<b>31</b>	<b>Collecting Social Network Data</b>	<b>241</b>
	<i>Tobias H. Stark</i>	

**Section 3 Linking Survey Data with External Sources**

- 32 Methods of Linking Survey Data to Official Records** 257  
*Joseph W. Sakshaug*
- 33 Linking Individual-Level Survey Data to Consumer File Records** 263  
*Josh Pasek*
- 34 Linking Survey Data to Administrative Records in a Comparative Survey Context** 267  
*Annelies G. Blom and Julie Korbmacher*
- 35 Linking Survey Data with the Catalist Commercial Database** 275  
*Robert Blaemire*
- 36 Challenges and Opportunities in Collecting Election Administration Data** 289  
*Michael P. McDonald*
- 37 Challenges with Validating Survey Data** 293  
*Matthew K. Berent*
- 38 The Promise of Collaborative Data Sharing Across Research Sectors** 297  
*Robert M. Groves*

**Section 4 Improving Research Transparency and Data Dissemination**

- 39 The Importance of Data Curation** 303  
*Steven Ruggles*
- 40 Improving the Usability of Survey Project Websites** 309  
*David L. Vannette*
- 41 The Role of Transparency in Maintaining the Legitimacy and Credibility of Survey Research** 315  
*Arthur Lupia*

<b>42 Evidence-Based Survey Operations: Choosing and Mixing Modes</b>	<b>319</b>
<i>Michael Bosnjak</i>	
<b>43 Best Practices for Survey Research</b>	<b>331</b>
<i>David L. Vannette</i>	
<b>Section 5 Detailed Chapters</b>	
<b>44 Reasons for Optimism About Survey Research</b>	<b>347</b>
<i>Jon A. Krosnick</i>	
<b>45 Probability Versus Non-Probability Methods</b>	<b>351</b>
<i>Gary Langer</i>	
<b>46 Address-Based and List-Based Sampling</b>	<b>363</b>
<i>Colm O'Muircheartaigh</i>	
<b>47 The Impact of Survey Non-response on Survey Accuracy</b>	<b>373</b>
<i>Scott Keeter</i>	
<b>48 Optimizing Response Rates</b>	<b>383</b>
<i>J. Michael Brick</i>	
<b>49 Data Collection Mode</b>	<b>393</b>
<i>Roger Tourangeau</i>	
<b>50 Survey Incentives</b>	<b>405</b>
<i>Eleanor Singer</i>	
<b>51 Building Household Rosters Sensibly</b>	<b>417</b>
<i>Kathy Ashenfelter</i>	
<b>52 Proxy Reporting</b>	<b>427</b>
<i>Curtiss Cobb</i>	
<b>53 Questionnaire Design</b>	<b>439</b>
<i>Jon A. Krosnick</i>	
<b>54 Cognitive Evaluation of Survey Instruments</b>	<b>457</b>
<i>Gordon Willis</i>	

<b>55</b>	<b>Interviewer Deviations from Scripts</b>	<b>465</b>
	<i>Nora Cate Schaeffer</i>	
<b>56</b>	<b>Coding Open Responses</b>	<b>473</b>
	<i>Arthur Lupia</i>	
<b>57</b>	<b>What HLT Can Do for You (and Vice Versa)</b>	<b>489</b>
	<i>Mark Liberman</i>	
<b>58</b>	<b>Confidentiality, Privacy, and Anonymity</b>	<b>501</b>
	<i>Roger Tourangeau</i>	
<b>59</b>	<b>Panel Attrition</b>	<b>509</b>
	<i>Randall Olsen</i>	
<b>60</b>	<b>Computation of Survey Weights</b>	<b>519</b>
	<i>Matthew DeBell</i>	
<b>61</b>	<b>Paradata</b>	<b>529</b>
	<i>Frauke Kreuter</i>	
<b>62</b>	<b>Interviewer Observations</b>	<b>537</b>
	<i>Brady T. West</i>	
<b>63</b>	<b>Leave-Behind Measurement Supplements</b>	<b>549</b>
	<i>Michael Link</i>	
<b>64</b>	<b>Ecological Momentary Assessment and Experience Sampling</b>	<b>563</b>
	<i>Arthur Stone</i>	
<b>65</b>	<b>Biomarkers</b>	<b>573</b>
	<i>David Weir</i>	
<b>66</b>	<b>Specialized Tools for Measuring Past Events</b>	<b>587</b>
	<i>Robert F. Belli</i>	
<b>67</b>	<b>Linking Survey Data to Official Government Records</b>	<b>597</b>
	<i>Joseph W. Sakshaug</i>	

<b>68</b>	<b>Linking Knowledge Networks Web Panel Data with External Data</b>	<b>607</b>
	<i>Josh Pasek</i>	
<b>69</b>	<b>History and Promise and Blending Survey Data with Government Records on Turnout</b>	<b>621</b>
	<i>Michael P. McDonald</i>	
<b>70</b>	<b>Improving Information Quality and Availability Through Interactions Between Government and Academic, and Industry Survey Research Sectors</b>	<b>627</b>
	<i>Robert M. Groves</i>	
<b>71</b>	<b>Metadata and Preservation</b>	<b>635</b>
	<i>Steven Ruggles</i>	
<b>72</b>	<b>Usability of Survey Project Websites</b>	<b>645</b>
	<i>David L. Vannette</i>	
<b>73</b>	<b>Research Transparency and the Credibility of Survey-Based Social Science</b>	<b>655</b>
	<i>Arthur Lupia</i>	
	<b>Index</b>	<b>667</b>



# 37

## Challenges with Validating Survey Data

Matthew K. Berent

Self-reports of voting behavior in many studies, including the American National Election Studies (ANES), are often about 20 percent higher than the official statistics on turnout. This has been a remarkably consistent finding since the 1960s, meaning that as actual turnout increases or decreases there are similar increases or decreases in self-reports of voting. Many people take this information to imply that the estimates generated by self-report survey data are not representative of the population because they don't match the "true" gold standard value provided by the government.

However, there are a number of different ways that this discrepancy could arise that warrants consideration. For example, a respondent might not answer the question because perhaps they believe voting behavior is private and sensitive and are unwilling to report their answer to the survey question, which leads to nonresponse and inaccuracy in the measure that is unrelated to the representativeness of the survey sample.

The second explanation is that there could be survey effects, meaning that participating in the survey could influence the voting behavior that the survey is attempting to measure. Many voting behavior studies involve very in-depth interviews before the election, and then an interview after the election. It is reasonable to expect that, after having answered a long battery of questions about politics before an election, respondents are made aware of

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their political attitudes that were perhaps not as salient before the interview. This increased salience could then influence some respondents to go vote when they might not have otherwise. The result is that these surveys could inadvertently be biasing their own sample to over-represent voters.

This raises a third potential mechanism for higher levels of voting behavior in surveys than in the general population, it is possible that survey response propensity is positively associated with the propensity to vote, meaning that the same types of people who are willing to respond to a long interview about politics may be the same types of people who go out and vote; this would be another way that the sample of survey respondents could be biased from the general population, not demographically but behaviorally.

Lastly, it is entirely possible that respondents misreport their voting behavior, meaning that their answers are factually incorrect. This could be caused by respondents misinterpreting the question, misremembering the behavior, reporting based on their typical voting behavior rather than their specific behavior in the focal election, or intentional lying which might be due to social desirability bias if the respondent believes that it would reflect poorly on them to report not voting in the election to the interviewer.

Misreporting is the most commonly cited cause of the discrepancy between survey data and the official records, leading to the typical conclusion that official records should be used instead of self-reports. However, the approach of using official records is not completely straightforward. To employ this method, researchers must obtain the official records of turnout histories, match each respondent to his or her official record, and then determine the “correct” turnout status for each respondent.

There are two primary problems with the validation task. The first is that over 200 million people are currently eligible to vote in U.S. elections, thus the matching task is not trivial simply on the basis of the size of the databases involved. Second, the federal government does not aggregate individual voting records, meaning that researchers must collect the records from each individual state and there is substantial variability between states in the availability, accuracy, and content of these records.

Researchers are increasingly turning to commercial vendors of political and voter file databases to conduct the matching between survey data and the official voter records. However, these researchers face another set of challenges when working with these databases. For example, vendors vary in their level of transparency, which means that researchers need to identify and specify the level of uncertainty that they are willing to accept with regard to the provenance and quality of their data. Vendors are also typically unwilling

to provide complete details about their matching algorithms, which makes it impossible to estimate the reliability or validity of the matches.

Some researchers opt to attempt in-house matching by obtaining the government registration and turnout records from a sample of states. Publicly available computer applications, such as LinkPlus from the Centers for Disease Control, can then be used to match survey respondents to their official government records. This approach requires more work on the part of the researcher but the benefits are complete control over the data cleaning and matching processes.

Recent evidence from an ANES in-house matching project indicates that two main factors are contributing to the discrepancy between self-reports and official records of voting behavior (Berent et al. 2016). The first is a downward bias in government records that occurs when the records incorrectly identify some respondents as having not turned out when in fact they did. These are all cases where a record match cannot be found, not cases where the self-report and government record data disagree explicitly. The second factor is an upward bias in the self-reports that occurs because people who participate in surveys are more likely to vote. These biases are additive, not offsetting, and seem to account for nearly all of the discrepancies between the self-report data and government records.

Areas for future research:

- Developing a better set of tools for researchers to conduct their own transparent validation studies without needing to use commercial vendors
- Identifying and understanding the correlates of survey participation and turnout to better understand the potential mechanism that drives the higher levels of turnout among survey respondents

## Reference and Further Reading

Berent, M. K., Krosnick, J. A., & Lupia, A. (2016). Measuring Voter Registration and Turnout in Surveys Do Official Government Records Yield More Accurate Assessments? *Public Opinion Quarterly*, 80 (3), 597–621.

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