



Using Machine Learning to Identify Careless Web Respondents

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Web surveys

- Web surveys are a popular mode of data collection
- Known data quality issues
 - When used in a single-mode design, coverage error due to exclusion of people without internet access/or device
 - Nonprobability web panels worse inference than probability web panels
 - Nonresponse error when used both in a single-mode or in a multimode design
 - Measurement error when used both in a single-mode or in a multimode design)
 - Two types of web respondents of concern

Two types of web respondents of concern

- Fraudulent respondents (Kennedy et al. 2021; Puleston, 2019)
 - Bot
 - People living outside targeted area (or fake respondents)
 - Duplicate IPs/Multi-completers
 - Ghost respondents
- Careless respondents (Kennedy et al., 2021; Puleston, 2019; Jones et al., 2015)
 - Also called inattentive/insincere/bogus/satisficing respondents
 - Respondents do not read questions carefully, do not spend time and effort to carefully answer questions, multitask, not motivated
 - Focus of the talk

Impact of careless respondents





Careless Rs

Identifying careless respondents

During and/or after data collection

- Attention checks/instructional manipulation checks/traps (Gummer et al., 2021)
- Speeding (Conrad et al., 2017)
- Low-incidence question, inconsistent answers (Jones et al., 2015)
- Proxy indicators of data quality examined alone or together
 - Straightlining/nondifferentiated answers, extreme responses, midpoint, acquiescence, missing data rate
- Open-ended questions (Kennedy et al., 2021)
- Response entropy (Tawa, 2021)

This talk

- We explored using machine learning to identify careless respondents
 - Four unsupervised clustering methods
 - K-means clustering
 - Hierarchical clustering
 - Density-based spatial clustering of applications with noise (DBSCAN)
 - Mean shift clustering
- How does each clustering method work?
- Do they converge?

Data

- National Study of Social, Economic, and Health Experiences (NSSEHE)
 - Tracks changes in opinions, life style, and health of Americans
 - Experiments to investigate mechanisms account for panel conditioning
- A sample of 8000 registered voters in two states
 - Invited to participate in four waves of web surveys through mailings, emails, and text messages
- Fourth wave data collection between February 2023 to March 2023
 - a total of 947 completes at a response rate of 71.4%

Data (2)

- Variables used in clustering methods
 - Whether or not R failed the trap questions
 - Whether or not R reported multitasking
 - Whether or not R answered too fast
 - Item nonresponse rate
 - Extreme response rate
 - Middle response rate
 - Response entropy

Identifying careless respondents

Variables used	Cases flagged
Whether or not R failed trap questions	5% failed at least one trap question
Whether or not R reported multitasking	25% reported multitasking
Whether or not R answered too fast	5% fastest
Item nonresponse rate	7% with item nonresponse rate $>=5\%$
Extreme response rate	8% with extreme response rate>=50%
Middle response rate	1% with middle response rate $>=50\%$
Response entropy	10% with largest and smallest 5%



K-Means Clustering

Variables used	Cluster 1 (n=482)	Cluster 2 (n=465)
Whether or not R failed trap questions	5.2%	5.6%
Whether or not R reported multitasking*	30.2%	20.3%
Whether or not R answered too fast	5.6%	4.5%
Item nonresponse rate>=5%	5.8%	7.8%
Extreme response rate>=50%*	0%	16.1%
Middle response rate>=50%*	2.1%	0%
Response entropy too large or too small	10.0%	10.3%

**p*<.05

Hierarchical clustering

Variables used	Cluster 1 (n=48)	Cluster 2 (n=899)
Whether or not R failed trap questions	8.3%	5.2%
Whether or not R reported multitasking	25.0%	25.3%
Whether or not R answered too fast*	100%	0%
Item nonresponse rate>=5%	10.4%	6.6%
Extreme response rate>=50%	10.4%	7.8%
Middle response rate>=50%	2.1%	1.0%
Response entropy too large or too small	12.5%	10.0%

**p*<.05

DBSCAN clustering

Variables used	Cluster 1 (n=98)	Cluster 2 (n=849)
Whether or not R failed trap questions*	52.0%	0%
Whether or not R reported multitasking	25.0%	25.3%
Whether or not R answered too fast*	49.0%	0%
Item nonresponse rate>=5%*	13.3%	6.0%
Extreme response rate>=50%	11.2%	7.5%
Middle response rate>=50%	2.0%	0.9%
Response entropy too large or too small	10.2%	10.1%

**p*<.05

Mean shift clustering

Variables used	Cluster 1 (n=23)	Cluster 2 (n=924)
Whether or not R failed trap questions^	13.0%	5.2%
Whether or not R reported multitasking	31.8%	25.1%
Whether or not R answered too fast	4.4%	5.1%
Item nonresponse rate>=5%*	21.7%	6.4%
Extreme response rate>=50%	13.0%	7.8%
Middle response rate>=50%	0%	1.1%
Response entropy too large or too small*	60.9%	8.9%

*p<.05; ^ p<.10

Convergence of four clustering methods

- 85% careless Rs were flagged by only one method
- 10% flagged by two methods
- 5% by three methods
- 1 case flagged by all four methods
- Hierarchical part of DBSCAN



Convergence of four clustering methods

- 51% careless Rs were flagged by only one method
- 48% flagged by two methods
- 1 case flagged by all three methods
- Hierarchical part of DBSCAN





Conclusions and Discussion

- Clustering methods can be used to identify careless respondents
 - Different number of respondents were identified
 - K-Means>DBSCAN>Hierarchical>Mean Shift
 - Overlaps between DBSCAN and Hierarchical
 - Different variables determined clusters
 - K-Means: Multitasking, ERS, MRS
 - Hierarchical: speeding
 - DBSCAN: trap, speeding, item nonresponse
 - Mean Shift: trap, item nonresponse, response entropy

Conclusions and Discussion (2)

More research needed to evaluate and validate the methods

- Number of clusters?
 - K-Means and hierarchical methods produce solutions of more than 2 clusters
- Are clusters meaningful and different from each other?
- Who are in each cluster?
- Can we use these clusters to inform detection of careless respondents during data collection?
- Can we predict careless respondents based on these clusters?



Thank you

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