

Discussion Data Driven Insights: The Utility and Policy Building of Integrated Data from Federal Statistical Agencies

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FCSM 2023

October 26, 2023: 10.30 AM

NATIONAL CENTER FOR SCIENCE AND ENGINEERING STATISTICS NATIONAL SCIENCE FOUNDATION

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Linking Data

- A powerful and efficient mechanism for producing policy-relevant information
 - Brings together information to create a new, richer resource
 - Allows for the construction of longitudinal data with passive follow-up



Factors to Consider in Data Linkage

- Linkage eligibility (e.g., consent, sufficient personally identifiable information)
- Linkage error
- Analytic considerations
 - o Data quality
 - Disclosure risks
 - Coverage
 - o Data limitations and inference
 - o Timeliness



Three Diverse Linkage Talks

- National Center for Science and Engineering Statistics
 - Linking the Policy and Utility of Linked Data in Science and Engineering (Finamore)
- National Center for Health Statistics
 - Does the Decade Matter? Examining the Impact of Using Geocodes from Different Decades in the Analysis of Merged Survey and Contextual Data (Parker)
 - Using Linked Data to Train and Validate Machine Learning Prediction Models (Davy)



Linking the Policy and Utility of Linked Data in Science and Engineering

- Recent NCSES data linkage efforts
 - Development of a policydriven research and linkage framework
 - Insights on the way forward in this critical field of data integration
 - What lessons can be learned from others?





Does the Decade Matter?



- Geocoded survey data allow researchers to merge survey data with geographic information from other sources to add contextual information
- Geocodes change with each decennial Census
- Time period of the geocodes should be carefully considered because they may impact results due to real change over time and/or administrative changes



Using Linked Data to Train and Validate Machine Learning Prediction Models

- When data linkage is not possible, machine learning (ML) prediction models can be used to predict outcomes, such as morbidity and mortality
- ML models require high quality, accurate training data and a validation source
- ML models perform reasonably well for predicting all-cause mortality





Takeaways and Common Themes

- Establishing a robust linkage program requires many analytic and practical considerations
- By augmenting information, linked survey and administrative information facilitate richer analyses
- Quality assessments of linked data are important for inference and training datasets



Discussion Points

- What are key policies that need to be considered when establishing a robust linkage program?
- Are linkages, whether entity to entity or geocoded data, being considered to replace information that could be collected through survey questions?
- If linked data were to be used as a training data set for a ML model how often would they need to be updated?
 - What skills are needed to keep the training dataset up to date?



Discussion Points cont.

- Are considerations being given to utilize linkages to inform data collection (e.g., adaptive design)?
- What information is missing from the surveys discussed that would benefit from linking to another source to support evidence-based policymaking?



Final Thoughts

- Continue to identify and integrate the data needed to inform key policy questions
- Utilize innovative technologies and data sources
- Evaluate alternative data sources for linkages and uses of linked data





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