

# Assessing Utility of Synthetic Data for Two Studies: Applications to the Survey of Doctoral Recipients and Census Transportation Planning Products

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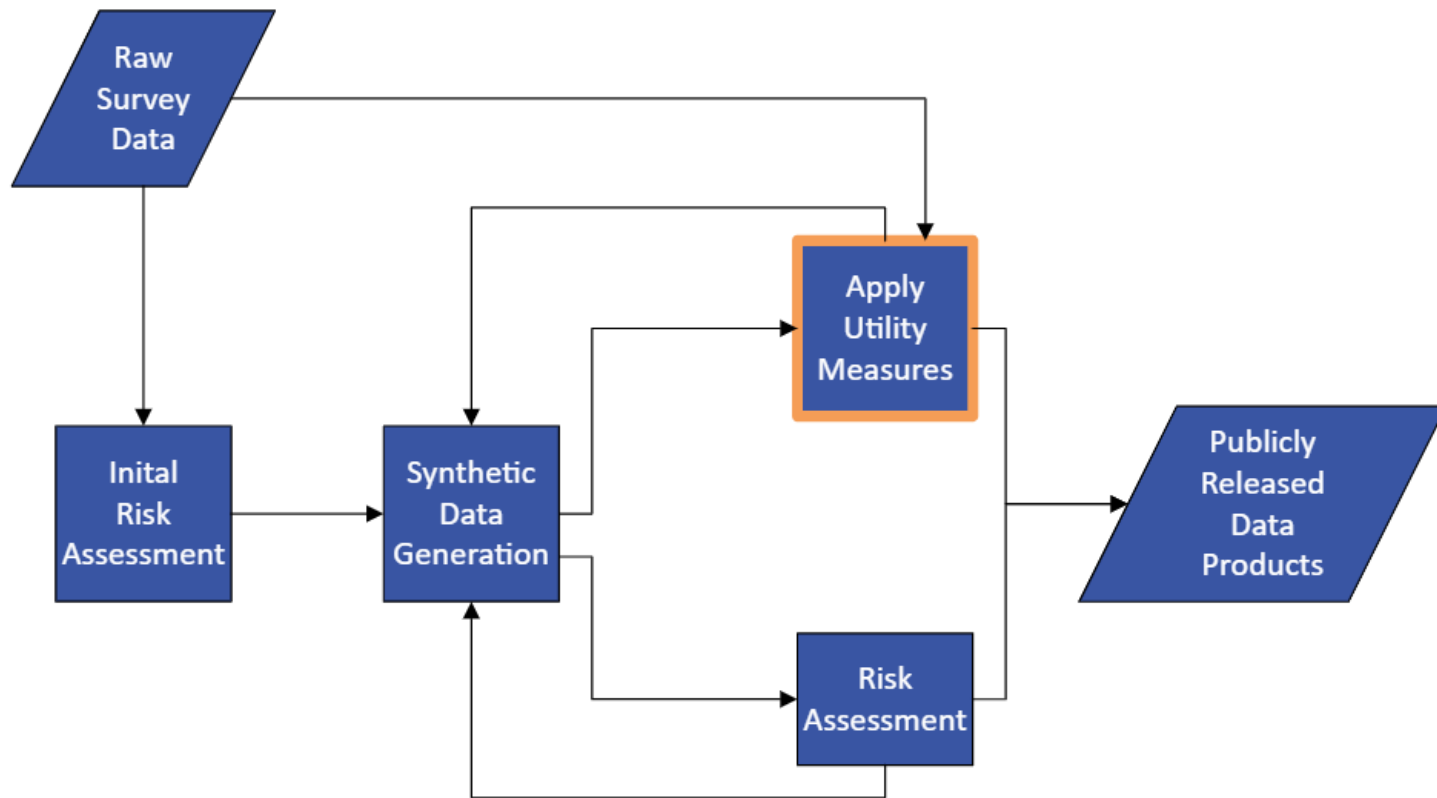
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# Synthetic Data

- Synthetic data generation is gaining traction as a method for avoiding disclosure in publicly released data products
- Conclusions reached using the synthetic data must be similar to those reached using the original data
- A careful balance must be struck between utility and risk of disclosure
- This presentation focuses on the utility side of the risk-utility tradeoff

# Synthetic Data Generation Process



# Utility Measures

- We compiled a list of several utility measures, which we sorted into five broad groups:
  - QC Checks
  - Weighted Frequency Checks
  - Measures of Association
  - Dataset-Wide Checks
  - Equity-Focused Measures
- These measures were applied to data synthesized for the Survey of Doctoral Recipients (SDR), sponsored by NCSES, and the Census Transportation Planning Products (CTPP), sponsored by Census Bureau and American Association of State Highway and Transportation Officials

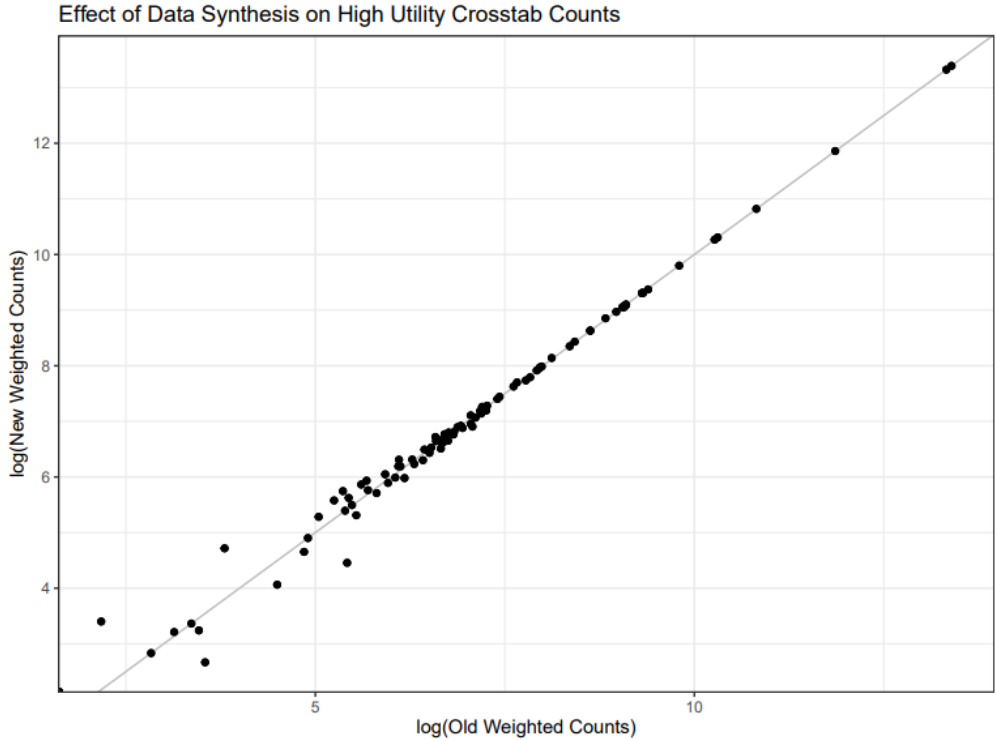
# QC Checks

- Percent of records changed for each variable
- Percent of variables changed for each record
- Logic checks: differences in missing value patterns/combinations

# Weighted Frequency Checks

- Distances between variables
  - Summary statistics of individual differences for continuous variables
  - Hellinger distance for categorical variables
- High-utility crosstabs
- Confidence interval overlap
- Indicators of whether synthesized estimate is within the original confidence interval

# Example: High-Utility Crosstabs

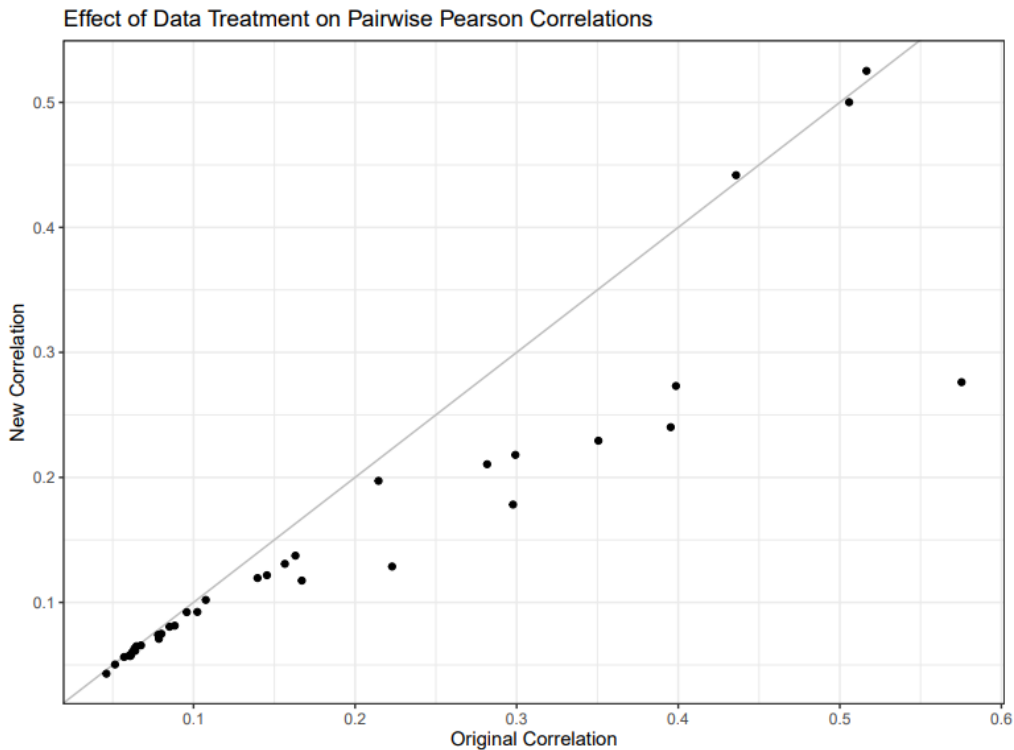


# Measures of Association

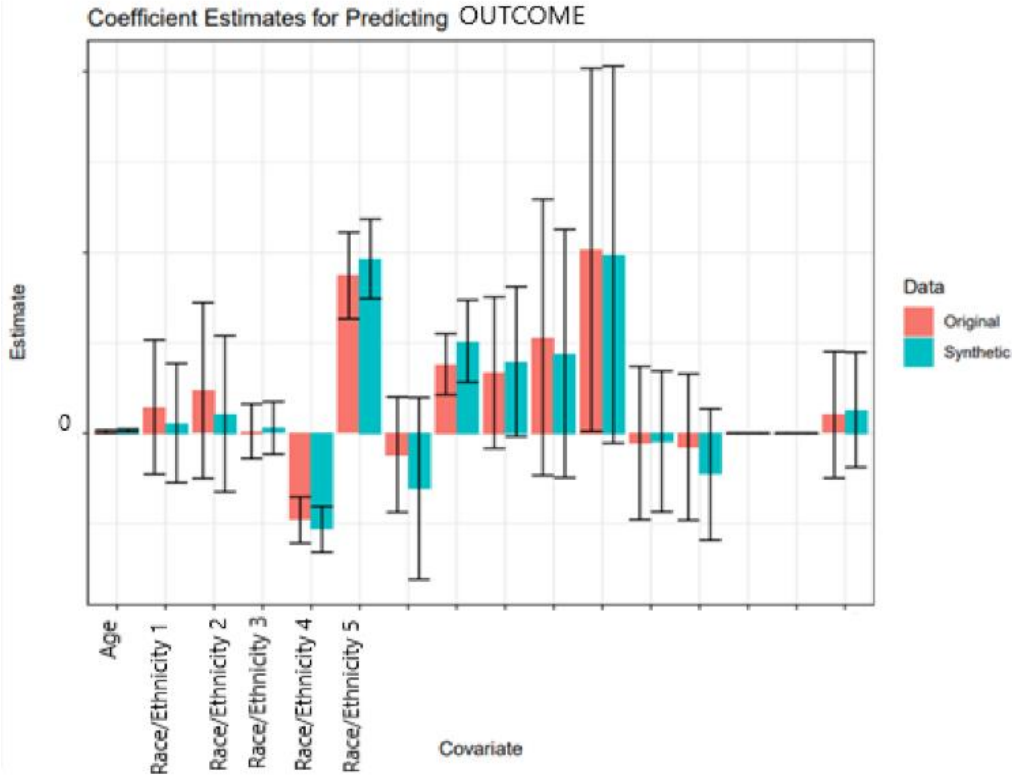
- Pairwise associations
- Significance of regression coefficients
- U-Statistic (Woo et al., 2009, Snoke et al., 2018)



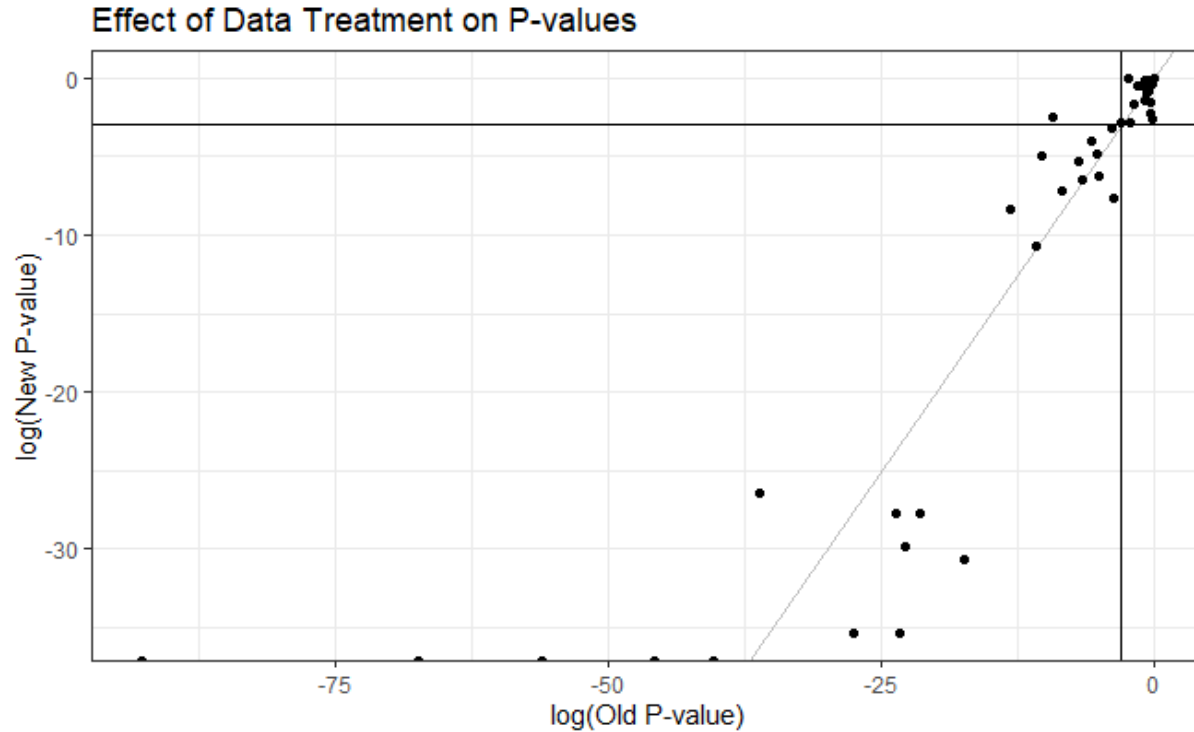
# Example: Pairwise Associations for an Initial Diagnostic Run



# Example: Regression Coefficients



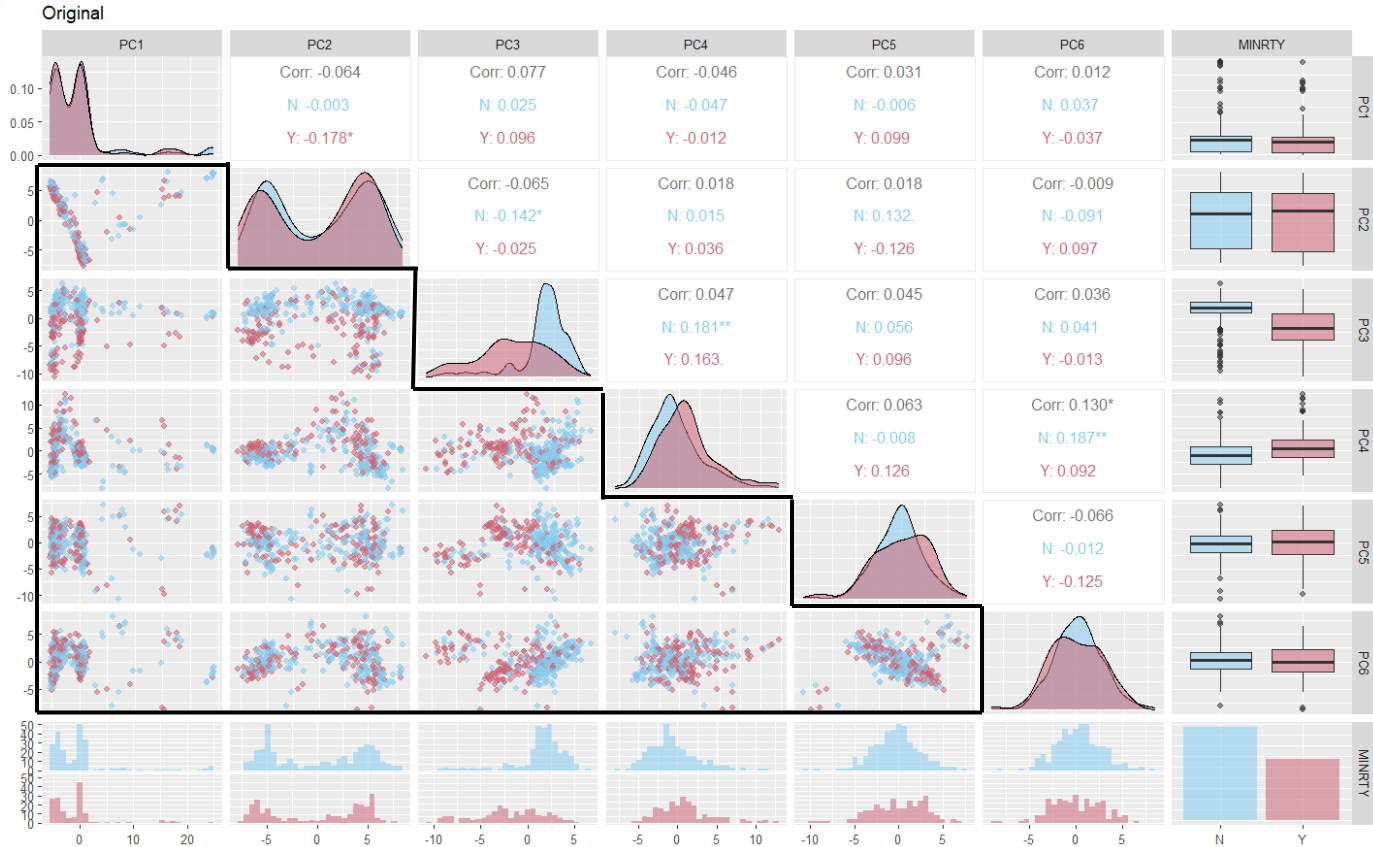
# Example: Significance of Regression Coefficients



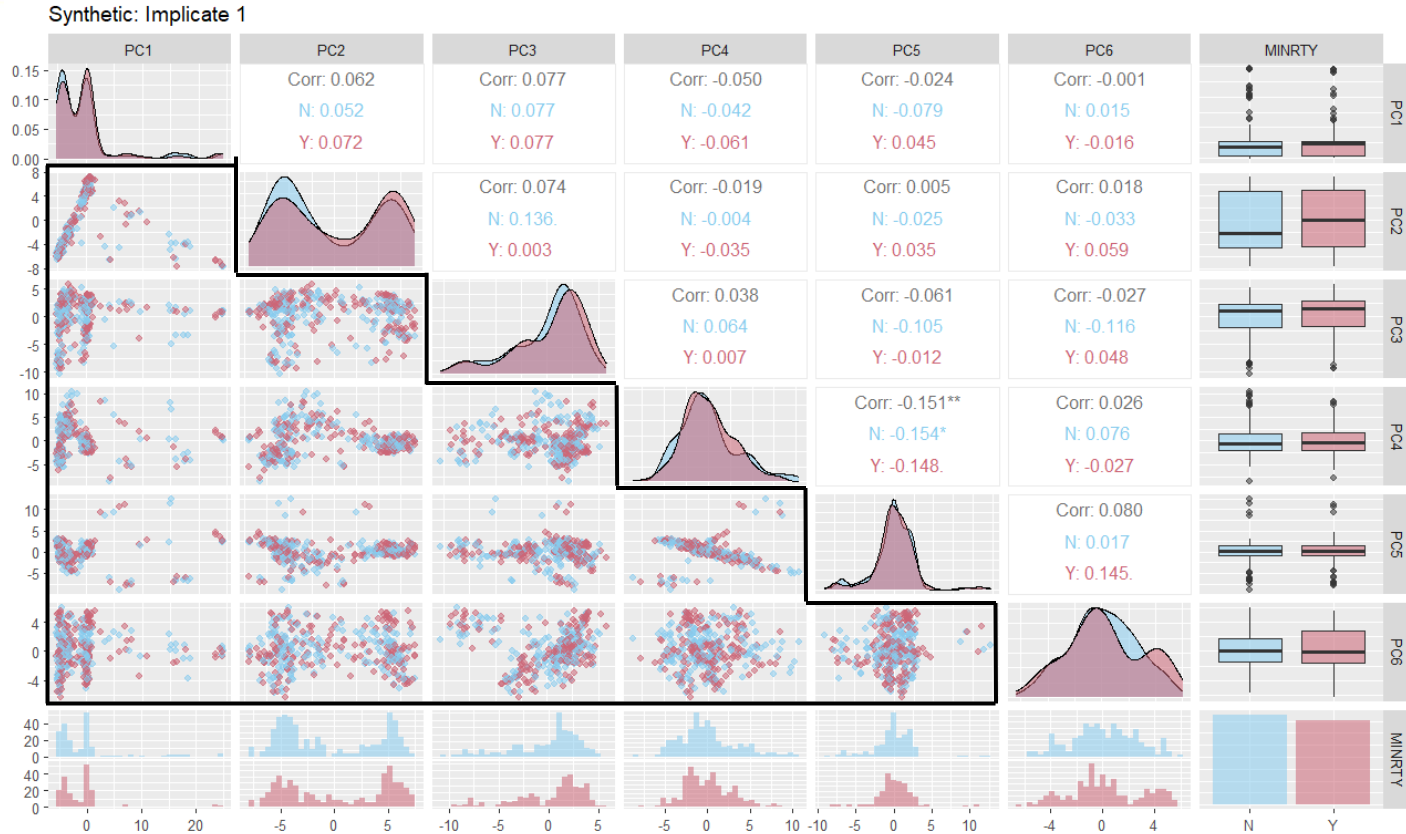
# Dataset-Wide Checks

- Perform Principal Component Analysis (PCA) on the original and synthetic data sets to ensure patterns hold across the entire data set

# Example: PCA on Original Data



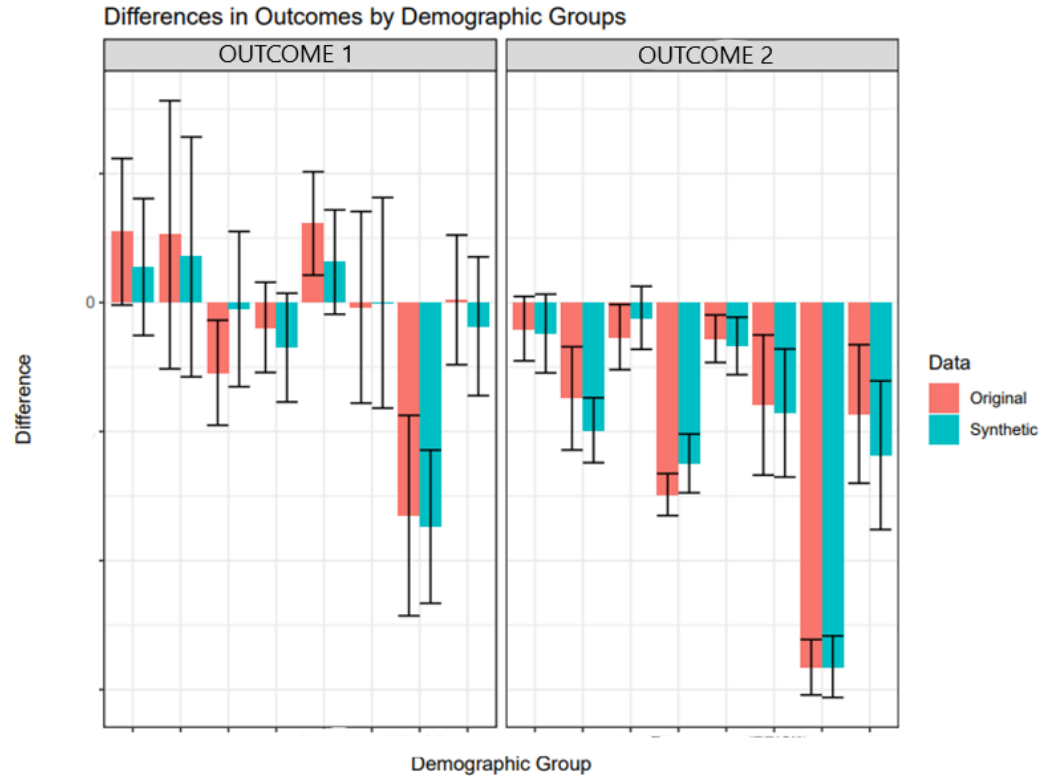
# Example: PCA on Synthetic Data



# Equity-Focused Measures

- Apply earlier utility measures, but subset to specific demographic groups
- Compare differences in outcomes between subgroups

# Example: Differences in Outcomes by Demographic Groups





# Conclusion

- The utility measures are important in determining whether a proposed publicly released synthesized data set is suitable for analysts' needs
  - Iterative process of synthetic data generation
- Results used in combination with risk assessment to ensure an appropriate balance between respondents' privacy and data usefulness
- Take into consideration minority and vulnerable demographic groups to ensure equitable analytic results

# Thank you!

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Photos are for illustrative purposes only. All persons depicted, unless otherwise stated, are models.

# References

- Joshua Snoke, Gillian M. Raab, Beata Nowok, Chris Dibben, Aleksandra Slavkovic, General and Specific Utility Measures for Synthetic Data, *Journal of the Royal Statistical Society Series A: Statistics in Society*, Volume 181, Issue 3, June 2018, Pages 663–688, <https://doi.org/10.1111/rssa.12358>
- Mi-Ja Woo, Jerome P. Reiter, Anna Oganian, Alan F. Karr, Global Measures of Data Utility for Microdata Masked for Disclosure Limitation. *Journal of Privacy and Confidentiality*, Volume 1, Issue 1, 2009, Pages 111-124, <https://doi.org/10.29012/jpc.v1i1.568>