

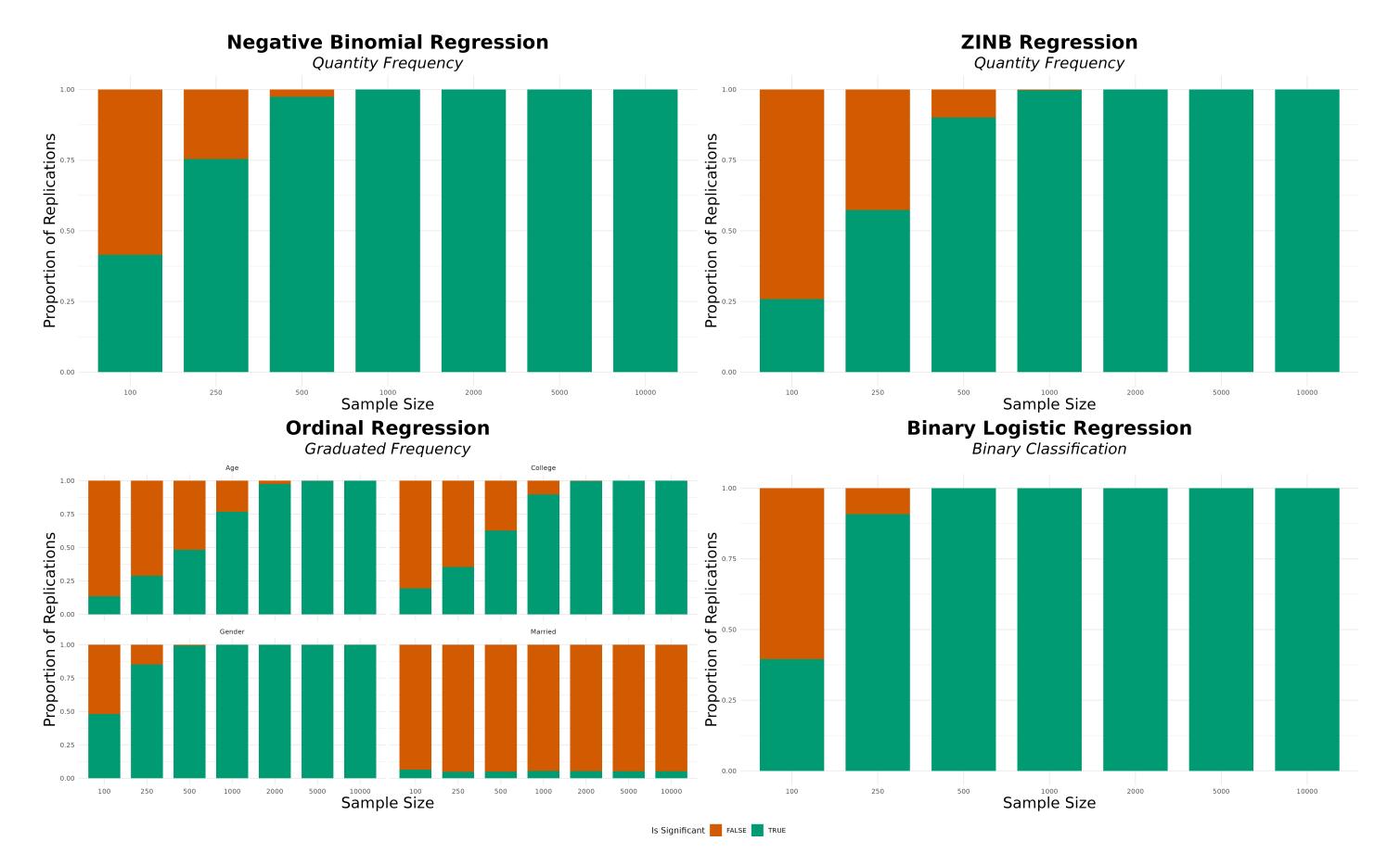
The Statistical Power of Alcohol Consumption Measures

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Background

- Understanding alcohol use epidemiology is essential for public health surveillance, policy, and intervention design.
- However, while self-report surveys remain the primary tool for measuring consumption, their reliability depends heavily on



Results

methodological precision.

• Using simulated data, we systematically evaluate the statistical power of three widely used alcohol consumption measures to determine which best detects associations in epidemiological research.

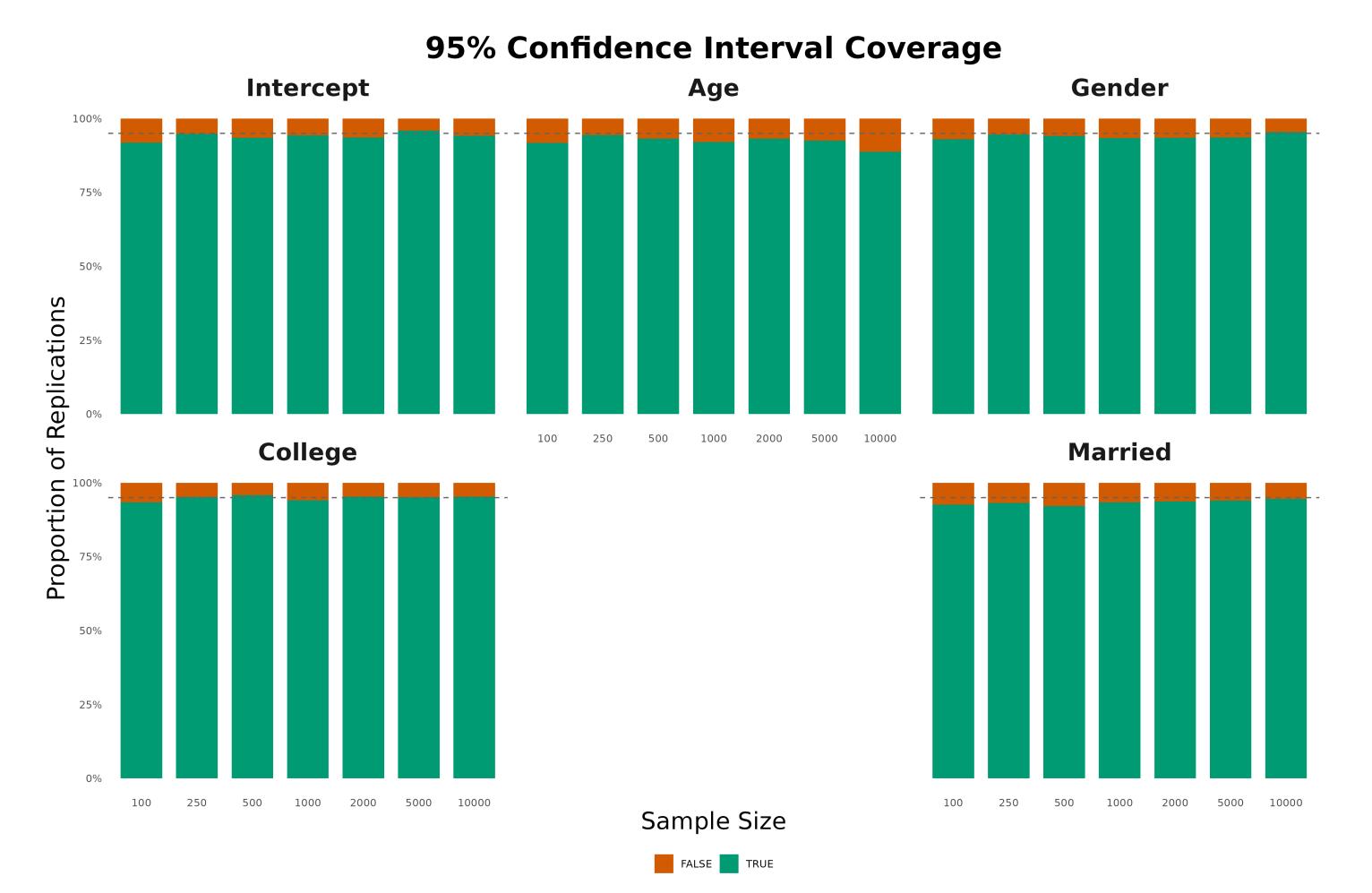
Measure	Response
Quantity frequency	Count data
Graduated frequency	Ordinal data
Binary classification	Binary data

Methods

We conducted a simulation study to compare the statistical power of different alcohol measurement approaches:

- Data Simulation
 - Generated a large-scale dataset reflecting real-world alcohol consumption distributions.

Figure 1: Proportion of Significant Wald Values



- Simulated **7000 random samples** across varying sample sizes $n \in \{100, 250, 500, 1000, 5000, 10000\}$
- Modelling Approach
 - Fitted four generalized linear models based on response variable type:
 - Negative binomial (quantity-frequency)

 $\begin{aligned} Y_i \sim \mathrm{NB}(\mu_i, \theta) \\ \log(\mu_i) = \boldsymbol{\beta}^\top X_i \end{aligned}$

- Zero-inflated negative binomial (quantity-frequency)

$$Y_i \sim \begin{cases} \pi + (1 - \pi) \cdot \operatorname{NB}(0 \mid \mu_i, k) & \text{ if } y = 0 \\ (1 - \pi) \cdot \operatorname{NB}(y \mid \mu_i, k) & \text{ if } y > 0 \end{cases}$$

$$\log(\mu_i) = \boldsymbol{\beta}^\top X_i$$

Figure 2: Proportion of 95% CI Coverage

Conclusion

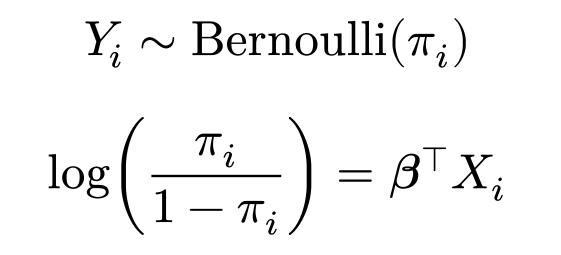
- With small sample sizes, binary data is the most robust approach for measuring alcohol consumption. However, with larger sample sizes, count data performs equivalently.
- Given these findings, we recommend the use of binary classifications when working with small sample sizes due to their robust performance.
- For large sample sizes, count data can offer greater precision

$$\log\!\left(\frac{\pi}{1-\pi}\right) = \boldsymbol{\gamma}^{\!\top} \boldsymbol{z}_i$$

- Ordinal logistic regression (graduated frequency)

$$\log \! \left(\frac{P(Y_i \leq \ell)}{P(Y_i > \ell)} \right) = \zeta_\ell - \beta^\top X_i$$

- Binary logistic regression (binary classification)



- by capturing more detailed variations in consumption levels.
- These findings have important implications for public health research and policy, where accurate measurement is critical.
 Future studies should explore how these methods perform in
 - broader populations, including high-risk groups and cross-

cultural settings, to ensure generalizability.

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