Assessment of a Modified Sandwich Estimator for Generalized Estimating Equations with Application to Opioid Poisoning in **MIMC-IV ICU Patients**

Introduction

We demonstrated in simulations with sample sizes of 100 subjects and an autoregressive covariance structure with Longitudinal regression models for correlated binary higher correlation settings (0.10 and 0.15) that all sandwich estimators produced coverage probabilities that fell below outcomes are frequently fit using generalized estimating 95%. This was not observed in our earlier simulations with low correlation values. As the sample sizes dropped under equations (GEE). The Liang and Zeger sandwich estimator these same correlation conditions, the Liang-Zeger continued to perform abysmally while the Rogers-Stoner and Pan is often used in GEE to produce unbiased standard error estimators adjusted. As the sample sizes decreased under a 0.10 correlation with 10% and 5% outcome prevalences, the estimation for regression coefficients in large sample coverage probabilities of the Liang-Zeger continued to deteriorate, while the Rogers-Stoner and Pan estimators settings, even when the covariance structure is recovered, almost achieving 95% coverage probabilities at 40 subjects and lower. misspecified. The sandwich estimator performs optimally in balanced designs when the number of participants is **Coverage Probabilities** large with few repeated measurements. The sandwich Prevalence: 0.10 Prevalence: 0.05 1.00 estimator's asymptotic properties do not hold in small sample and rare-event settings. Under these conditions, the sandwich estimator underestimates the variances and 0.95 is biased downwards. The goal of this research is to construct a hybrid sandwich estimator able to correctly 0.90 model the variances in rare outcome and finite sample situations. Only a handful of statisticians have attempted 0.85 improving the performance of the sandwich estimator Prevalence: 0.30 Prevalence: 0.50 1.00 under these conditions. Here, the performance of a modified sandwich estimator is compared to the traditional Liang-Zeger estimator and alternative forms 0.95 proposed by authors Morel, Pan, and Mancl-DeRouen. Each estimator's performance was assessed with 95% 0.90 coverage probabilities for the regression coefficients using simulated data under various combinations of sample sizes 0.85 and outcome prevalence values with independence and 90 autoregressive correlation structures. Number of Subjects

Materials and Methods

We created a hybrid sandwich estimator and compared it to others by Pan, Mancl-DeRouen, Morel, and the traditional Liang-Zeger. The comparisons were done via simulations with real-world EHR SpO2 measurements from the MIMIC-IV ICU dataset after patient repositioning.

Literature Cited

Rogers P, Stoner J. Modification of the Sandwich Estimator in Generalized Estimating Equations with Correlated Binary Outcomes in Rare Event and Small Sample Settings. Am J Appl Math Stat. 2015;3(6):243-251. doi:10.12691/ajams-3-6-5

Pan W. On the Robust Variance Estimator in Generalised Estimating Equations. Biometrika, 2001;88(3), 901–906. http://www.jstor.org/stable/2673458

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Results and Discussion

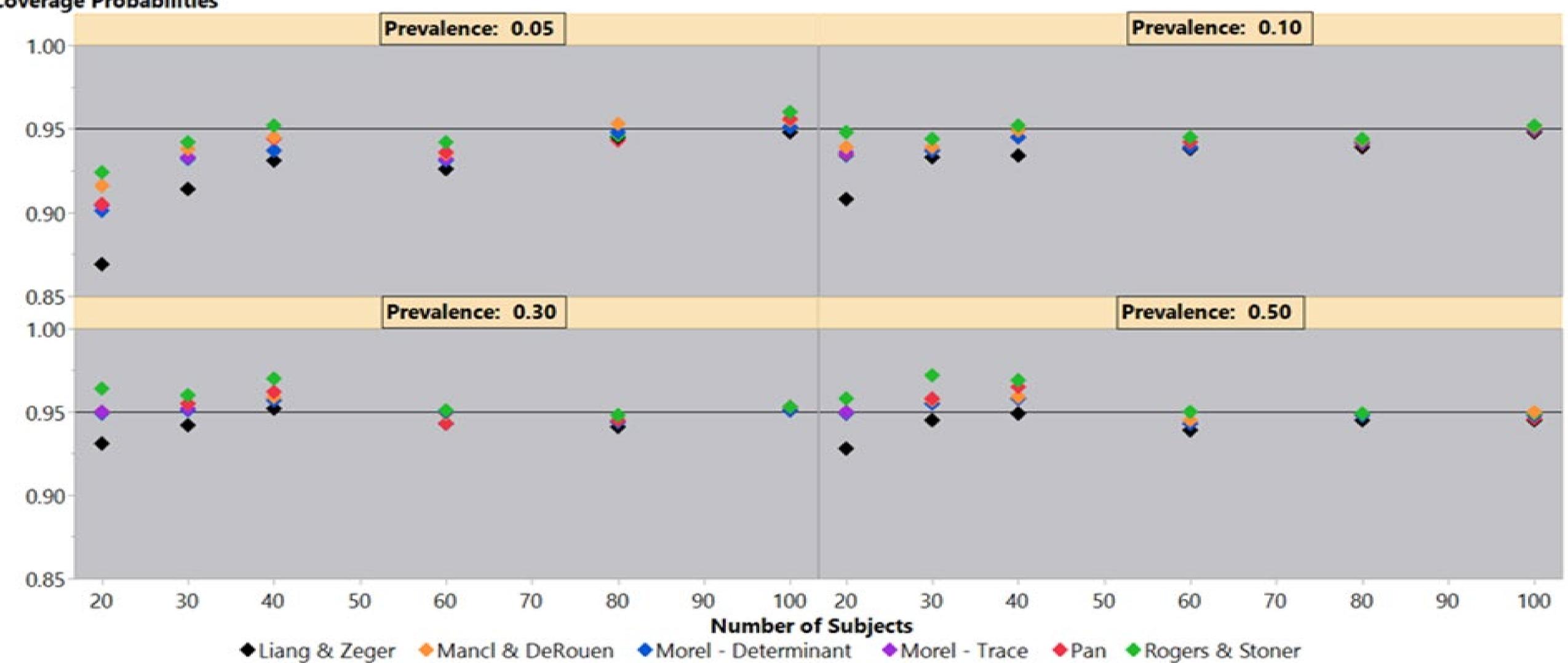


Figure 1. Coverage probabilities when estimating the regression coefficient β1 under a simulated independence covariance structure for 5% through 50% prevalence values.

Table 1. Estimated regression coefficients, odds ratios (OR), standard errors, 95% confidence intervals (CI) from analysis of repositioning events in recorded ICU SpO₂ measurements within the first 24 hours for 20 patients.

| Estimated Regression Coefficients Under an Autoregressive Correlation Structure | | | | | | |
|---|--------------------------|---------------------|--------|----------------|---------|---------|
| Method | $\hat{oldsymbol{eta}}_1$ | $SE(\hat{\beta}_1)$ | OR | 95% CI for OR | Z-Score | p-value |
| Liang-Zeger | -0.2712 | 0.0748 | 0.7624 | 0.6583, 0.8828 | -3.6238 | 0.0002 |
| Mancl-DeRouen | -0.2712 | 0.1413 | 0.7624 | 0.5779, 1.0057 | -1.9195 | 0.0549 |
| Morel – (Trace) | -0.2712 | 0.1047 | 0.7624 | 0.6209, 0.9361 | -2.5900 | 0.0095 |
| Morel-(Determinant) | -0.2712 | 0.0998 | 0.7624 | 0.6268, 0.9272 | -2.7156 | 0.0066 |
| Pan | -0.2712 | 0.2183 | 0.7624 | 0.4969, 1.1696 | -1.2423 | 0.2141 |
| Rogers-Stoner | -0.2712 | 0.2283 | 0.7624 | 0.4872, 1.1928 | -1.1877 | 0.2349 |



Conclusion

In our previously published research on simulations involving only low values of correlation, we concluded that, the choice of estimators matters, and alternatives to the Liang-Zeger estimator should be considered. This research further assessed the qualities of our hybrid sandwich estimator, reaffirming its superior performance over that of the Liang-Zeger estimator in simulations involving finite samples and low outcome prevalence. The Liang-Zeger sandwich estimator's performance suffered as the sample sizes dropped below 60 subjects under correlation settings as low as .01, when outcome prevalence values were less than 30%. This drop-off in performance was exacerbated further by lower outcome prevalence values, smaller sample sizes, and higher correlation settings. The real-world ICU practice of patient repositioning to improve opioid-induced respiratory depression showed the danger in hypothesis testing for low-prevalence outcomes with small sample sizes in GEE models with binary outcomes. Our approach provides a method for modeling rare events in finite samples on the effects of medications, drugs, and poisons.

Additional Information

Please see https://physionet.org/for more information. Contact Paul.Rogers@fda.hhs.gov for further questions or comments.



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*Julie Ann Stoner (1972-2020) Dedicated to the memory of Julie Stoner, a committed public health warrior. Deceased 18 June 2020.

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