## A Monte Carlo study on methods for handling class imbalance

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## Method

## **Data Generating Process**

Two class data were simulated by adapting the caret::twoClassSim R function (Kuhn, 2008):

- Two multivariate normal predictors (A and B) are generated. A and B are correlated at r = .65. These two variables contributed to the log-odds by 4A + 4B + 2AB.
- Another variable,  $J \sim U(-1,1)$ , was generated. This variable further added to the log-odds by  $J^3 + 2 \times \exp(-6 \times (J 0.3)^2)$ .
- Two more variables,  $K \sim U(0,1)$  and  $L \sim U(0,1)$ , were generated and contributed to the log-odds by  $2 \times \sin(K \times L)$ .
- For each data set, a number X was selected, where  $X \sim N(50,7)$ . Another number, Y, was selected, where  $Y \sim N(.15, .033)$ .  $Z = X (X \times Y)$  variables were generated from a N(0,1) distribution. Each of these Z variables further added to the log-odds in a simple additive fashion, where coefficients were (a) of alternating signs and (b) evenly spaced from 2.50 to 0.25.
- $\frac{Y}{2}$  variables were generated from a N(0,1) distribution and did not contribute to the log-odds.
- The log-odds for each case were converted to probabilities. For each data set, a positive (i.e., minority) class proportion, M, was sampled from N(.03,.007). Probabilities were sorted from lowest to highest. The difference between the probability for the 1-Mth highest probability and M was calculated, and this constant was added to the probability for each case.
- Lastly, the number of cases in each data set were randomly drawn from a distribution N(40000, 5000). 500 data sets were generated, and sixteen combinations of sampling techniques and algorithms were fit to each of these data sets.

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Algorithms

Metrics

Results