### Supplementary Material

#### Wallisch C, Dunkler D, Rauch G, de Bin R, Heinze G Selection of variables for multivariable models: opportunities and limitations in quantifying model stability by resampling, 2020.

# Content



### Supplementary Material S1: Equivalence of subsampling with  $m = 0.5N$  and the bootstrap to obtain the sampling distribution of a regression coefficient

Consider  $\hat{\beta}$  the vector of estimates of regression coefficients from analysis of the original sample of size N. Let  $\hat{\beta}^{(b)}$  be the vector of estimates of regression coefficients from analysis of the bth resample. If the nonparametric bootstrap (resampling  $N$  observations with resampling) was applied, the following is an estimate of the variance of  $\hat{\beta}_j$ , the *j*th regression coefficient:<sup>1</sup>

$$
\frac{1}{B}\sum_{b=1}^B \left(\hat{\beta}_j^{(b)} - \hat{\beta}_j\right)^2
$$

If the jackknife is used (resampling N data sets each consisting of  $N - 1$  observations, resampling without replacement), then the variance estimator is given by

$$
\frac{N-1}{N}\sum_{i=1}^N(\hat{\beta}_j^{(i)}-\hat{\beta}_j)^2
$$

Here  $\hat{\beta}_j^0$  $(i)$  is the regression coefficient of interest from the *i*th subsample, i.e., from the subsample where observation  $i$  has been left out. The formula can be rewritten to

$$
\frac{N-1}{1} \cdot \frac{1}{N} \sum_{i=1}^{N} (\hat{\beta}_j^{(i)} - \hat{\beta}_j)^2
$$

where  $\frac{1}{N}\sum_{i=1}^{N}(\hat{\beta}_j^{(i)} - \hat{\beta}_j)^2$  denotes the expected squared deviation of a resampled regression coefficient from its original-sample counterpart. Note that the multiplicator  $A/L = (N - 1)/1$  is the ratio of included observations per resample divided by left-out observations per resample.

Now consider a general subsampling strategy of  $m$  observations without replacement. The expected value of leaving out  $N-m$  observations is given by  $\frac{1}{B}\sum_{b=1}^{B} (\hat{\beta}_j^{(b)} - \hat{\beta}_j)^2$ , where  $B = {N \choose m}$  $\binom{n}{m}$  could denote the number of different ways to draw  $m$  observations out of  $N$ . (One can approximate the full set of possible resamples by a random sample of, say,  $B = 1000$  resamples.) Considering that now  $A/L = m/(N - m)$ , the variance of  $\hat{\beta}_j$  with this subsampling scheme is well approximated by

$$
\frac{m}{N-m}\cdot\frac{1}{B}\sum_{b=1}^{B}\left(\hat{\beta}_j^{(b)}-\hat{\beta}_j\right)^2
$$

Thus, with  $m = N - m$ , or  $m = 0.5N$ , the distribution of the subsampled regression coefficient estimates the sampling distribution of the regression coefficient, just like the distribution of the bootstrapped regression coefficients.

# Supplementary Material S2: Distribution and correlation structure of design variables  $x_j$ ,  $j = 1, ..., 17$

**Supplementary Figure 1:** Histograms of simulated design covariates  $X_1, ..., X_{17}$  from 1,000 simulated observations.



Supplementary Figure 2: Correlation network graph of all continuous (circle) and binary (square) design covariates. Numbers printed close to the edges are empirical correlation coefficients observed in a simulation of 1,000 observations. Edges are shown for a pair of covariates if the absolute value of their correlation coefficient exceeded 0.10. Widths of edges are proportional to correlation.



# Supplementary Material S3: Simulation study: Monte Carlo errors

At a sample size of  $N=150$ , the Monte Carlo errors of the mean estimated MSF (x100) and of the RMSE of the estimated MSF (x100) were 0.01 in the bootstrap and in any subsampling approach. The following tables show the MCE of summary statistics of all other stability measures.



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# Supplementary Material S4: Simulation study: Variable inclusion frequencies (VIF) for BE(AIC), BE(0.05) and Lasso

**Supplementary Figure 3:** Mean estimated VIF by subsampling with  $m = 0.5N$  (*'5'*, blue),  $m =$  $0.632N$  ('6', purple),  $m = 0.8N$  ('8', red), by bootstrap ('B', yellow), and their estimands ('X', black). Variables are ranked by partial R².



Supplementary Figure 4: Root mean squared error (RMSE) of estimated VIF by subsampling with  $m = 0.5N$  ('5', blue),  $m = 0.632N$  ('6', purple),  $m = 0.8N$  ('8', red), by bootstrap ('B', yellow), and the omission/ selection strategy ('O', black). The omission/selection strategy sets the VIF estimate to 0 or 1 according to omission or selection in the model fitted on a simulated data set. Variables are ranked by partial R².



### Supplementary Material S5: Simulation study: Model selection frequency for the correct model for BE(AIC), BE(0.05) and Lasso

Supplementary Table 2: Estimands, mean and RMSE of estimates of MSF for subsampling with  $m = 0.5N$ ,  $m = 0.632N$ , and  $m = 0.8N$  (S<sub>0.5</sub>, S<sub>0.632</sub>, S<sub>0.8</sub>) and bootstrap (B) estimators. All numbers multiplied by 100.

### BE(AIC)



### BE(0.05)



#### Lasso



# Supplementary Material S6: Simulation study: Relative conditional bias (RCB)

Supplementary Figure 5: Relative conditional bias. Upper rows: median estimate; middle rows: median deviation to the estimand; lower rows: median absolute deviation to the estimand; red and blue indicate high and low multiple correlation of a variable with others; solid and dashed lines represent predictors and non-predictors. The line width is proportional to absolute effect size.



A. RCB for BE(AIC) estimated by subsampling with  $m = 0.5N$  (S<sub>0.5</sub>),  $m = 0.632N$  (S<sub>0.632</sub>),  $m =$  $0.8N$  (S<sub>0.8</sub>), and by bootstrap

B. RCB estimated by bootstrap for BE(AIC), BE(0.05) and Lasso



### Supplementary Material S7: Simulation study: Root mean squared difference (RMSD) ratio

Supplementary Figure 6: Root mean squared difference ratio. Upper rows: median estimate; middle rows: median deviation to the estimand; lower rows: median absolute deviation to the estimand; red and blue indicate high and low multiple correlation of a variable with others; solid and dashed lines represent predictors and non-predictors. The line width is proportional to absolute effect size.





B. RMSD ratio estimated by bootstrap for BE(AIC), BE(0.05) and Lasso



# Supplementary Material S8: Simulation study: Results for BE(AIC) in logistic regression

### Additional information to the simulation setup for logistic regression

The simulation setup for logistic regression was analogous to the setup for linear regression described in the manuscript up to the definition of the linear predictor  $\eta$ . The binary outcome  $Y_B$  was drawn from a Bernoulli distribution with event probability  $\pi = [1 + \exp(-\eta + 8.47)]^{-1}$  which yielded a Nagelkerke's R<sup>2</sup> of 0.18 and an expected event rate of approximately 0.5.<sup>2</sup>

### Supplementary Figure 7: Variable inclusion frequencies (VIF) in logistic regression

Left column: mean estimated VIF by subsampling with  $m = 0.5N$  ('5', blue),  $m = 0.632N$  ('6', purple),  $m = 0.8N$  ('8', red), by bootstrap ('B', yellow), and their estimands ('X', black). Right column: root mean squared error (RMSE) of estimated VIF by subsampling with  $m = 0.5N$ ('5', blue),  $m = 0.632N$  ('6', purple),  $m = 0.8N$  ('8', red), by bootstrap ('B', yellow), and the omission/ selection strategy ('O', black). The omission/selection strategy sets the VIF estimate to 0 or 1 according to omission or selection in the model fitted on a simulated data set. Variables are ranked by partial R².



Supplementary Table 3: Model selection frequency for the correct model in logistic regression Estimands, mean and RMSE of estimates for subsampling with  $m = 0.5N$ ,  $m = 0.632N$ , and  $m =$  $0.8N$  (S<sub>0.5</sub>, S<sub>0.632</sub>, S<sub>0.8</sub>) and bootstrap (B) estimators. All numbers multiplied by 100.



### Supplementary Figure 8: Relative conditional bias (left column) and root mean squared difference ratio (right column) estimated by bootstrap in logistic regression

Upper row: median estimate; middle row: median deviation to the estimand; lower row: median absolute deviation to the estimand; red and blue indicate high and low multiple correlation of a variable with others; solid and dashed lines represent predictors and non-predictors. The line width is proportional to absolute effect size.



### Supplementary Material S9: Simulation study: Results for BE(AIC) in Cox regression

#### Additional information to the simulation setup for Cox regression

The simulation setup for Cox regression was analogous to the setup for linear regression described in the manuscript up to the definition of the linear predictor  $\eta$ . To obtain a time-to-event outcome, Weibull distributed survival times T were drawn from  $\left(-\log(U)/\frac{1}{\epsilon}\right)$  $/(\frac{1}{5} \exp(\eta - 3.5)))$ <sup>1/3</sup>,<sup>3</sup> whereby *U* was standard uniformly distributed. The follow-up times  $C$  were drawn from a uniform distribution  $U(0.001, 3.851)$ . The observable survival time was defined as  $Y_T = min(T, C)$  and the status indicator  $\delta$  was 1 if  $T < C$  and 0 otherwise. This setup yielded a censoring rate of approximately 0.5 and a Schemper-Henderson V of 0.22.<sup>4</sup>

#### Supplementary Figure 9: Variable inclusion frequencies (VIF) for Cox regression

Left column: mean estimated VIF by subsampling with  $m = 0.5N$  ('5', blue),  $m = 0.632N$  ('6', purple),  $m = 0.8N$  ('8', red), by bootstrap ('B', yellow), and their estimands ('X', black). Right column: root mean squared error (RMSE) of estimated VIF by subsampling with  $m = 0.5N$ ('5', blue),  $m = 0.632N$  ('6', purple),  $m = 0.8N$  ('8', red), by bootstrap ('B', yellow), and the omission/ selection strategy ('O', black). The omission/selection strategy sets the VIF estimate to 0 or 1 according to omission or selection in the model fitted on a simulated data set. Variables are ranked by partial R².





Supplementary Table 4: Model selection frequency for the correct model for Cox regression Estimands, mean and RMSE of estimates for subsampling with  $m = 0.5N$ ,  $m = 0.632N$ , and  $m =$  $0.8N$  (S<sub>0.5</sub>, S<sub>0.632</sub>, S<sub>0.8</sub>) and bootstrap (B) estimators. All numbers multiplied by 100.

Supplementary Figure 10: Relative conditional bias (left column) and root mean squared difference ratio (right column) estimated by bootstrap for Cox regression

Upper row: median estimate; middle row: median deviation to the estimand; lower row: median absolute deviation to the estimand; red and blue indicate high and low multiple correlation of a variable with others; solid and dashed lines represent predictors and non-predictors. The line width is proportional to absolute effect size.



# Supplementary Material S10: Example: Correlation structure of variables and pairwise inclusion frequencies

Supplementary Figure 11: Correlation network graph of all continuous (circle) and binary (square) covariates. Numbers printed close to the edges are empirical correlation coefficients. Edges are shown for a pair of covariates if the absolute value of their correlation coefficient exceeded 0.15. Widths of edges are proportional to correlation.





## Supplementary Table 5: Pairwise inclusion frequencies of variables estimated by subsampling with  $m = 0.5N$

" $(+)$ " and " $(-)$ " indicate if a pair of variables were selected significantly more or less often than by chance applying a significance level of 0.01. This was determined by conducting Χ²-tests for independence.

BMI, body mass index; BP, blood pressure; chol, cholesterol; HDL, high-density lipoprotein

# References

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