

Appendix 1. Strategies for Assessing Model Performance and Avoiding Resubstitution Bias.

Data-Splitting

Advantages
Model performance on test data is unbiased estimate of how the model will perform on new data from the same population.

Disadvantages
When the size of the test dataset is small, estimates of model performance are highly variable. Statistically inefficient because the test data are only used for validation and not model selection or fitting⁷

***Key: Only "one look" at the test data**

Cross Validation (ex. 10-fold)

Advantages
Within each iteration, the model performance is unbiased. Variation of data-splitting that avoids "wasting data"

Disadvantages
Model selection procedure must be prescribed and automated. No flexibility to explore data, exercise judgment and refine procedures. Each iteration typically produces a different model - unclear what model to report. Computationally Intensive.

***Key: Cross Validation can only be performed once**

Bootstrapping

Advantages
Avoids "wasting data" - full dataset is used to fit model

Disadvantages
Model selection procedure must be prescribed and automated. No flexibility to explore data, exercise judgment and refine procedures. Computationally intensive

Step 1. Apply model selection procedure to data and compute performance measure S
 Step 2. For each bootstrap dataset, apply model selection procedure from step 1 and compute performance measure. Take the final, fitted model from the bootstrap dataset and apply it to the original dataset and compute a performance statistic S . The difference in performance statistics for the fitted model from the bootstrap dataset and the fitted model from the original dataset is the optimism. Repeat the procedure many times, say $n=1000$, so that there are 1000 estimates of the optimism.
 Step 3. Subtract the average optimism from the apparent performance statistic S to get an optimism-corrected S value.