

1. The Akaike Information Criterion (AIC) is derived from a metric that measures how close a candidate model is from the "true" model. 3 / 3 points
- True
- False
2. The AIC for linear regression can be written as 3 / 3 points
- $$AIC = 2(p + 1) + n (\log(RSS) - \log(n)),$$
- where RSS is the residual sum of squares for the model, and  $p$  is the number of predictors in the model.
- True
- False
3. The AIC strikes a balance between: 4 / 4 points
- The number of units in the dataset and the number of predictors in the dataset.
- The residual sum of squares and the number of predictors in the model
- Model fit and model complexity
- The residual sum of squares and the number of parameters in the model
4. The Bayes Information Criterion (BIC) strikes a balance between: 4 / 4 points
- The residual sum of squares and the number of predictors in the model
- The residual sum of squares and the number of parameters in the model
- Model fit and complexity
- The number of units in the dataset and the number of predictors in the dataset.
5. If AIC has been chosen as the best model selection criterion, then given several candidate models, one should choose the model with the highest AIC. 3 / 3 points
- True
- False
6. If BIC has been chosen as the best model selection criterion, then given several candidate models, one should choose the model with the lowest BIC. 3 / 3 points
- True
- False
7. If the adjusted  $R^2$  has been chosen as the best model selection criterion, then given several candidate models, one should choose the model with the lowest adjusted  $R^2$ . 3 / 3 points
- True
- False
8. For a dataset with  $n > 8$  rows, the AIC favors smaller models. 3 / 3 points
- True
- False
9. For a dataset with  $n > 8$  rows, the penalty for adding parameters is higher for BIC than AIC. 3 / 3 points
- True
- False
10.  $R^2$  and  $R_a^2$  are the same when comparing models with the same number of predictors. 3 / 3 points
- True
- False