Performance Evaluation Metrics

Applied Machine Learning for Educational Data Science

${\rm true}$

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In the next few weeks, we will talk about different approaches to build predictive models. We need some performance metrics to evaluate how useful a model is by itself, or how well it performs compared to alternative models. We will consider the performance metrics in two main categories: 1) metrics for predicting a continuous outcome, 2) metrics for predicting a binary outcome.

We will use two hypothetical small datasets (N = 30) for introducing these concepts. The first hypothetical dataset will have a continuous outcome and predictions. The second hypothetical dataset will have a binary outcome and predictions.

You can load these two hypothetical datasets using the code below:

1. Evaluation Metrics for Continuous Outcomes

Explained Variance (R^2)

Error related indices
Mean Absolute Error
Median Absolute Error
Mean Squared Error
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Squared Log Error
Mean Absolute Percentage Error
Baseline performance considerations
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