

Option 1: Separate Validation Set

 If you have "enough" data and plan to do some model tuning, you should really partition your data into three parts — Training, Validation and Test sets. There is no general rule for how you should partition the data and it will depend on how strong the signal in your data is, but an example could be: 50% Train,

25% Validation and 25% Test.

 The validation set is used strictly for model tuning (via validation of models with different parameters) and the test set is used to make a final estimate of the generalization error.

Training Set vs. Validation Set vs. Test Set

Validation is for Model **Tuning**







Validation Test Train

Option 2: Cross-Validation

Training Set vs. Test Set

- Partition the original data (randomly) into a training set and a test set. (e.g. 70/30)
- Train a model using the training set and evaluate performance (a single time) on the test set.



K-fold Cross-validation

- Train and test K models using separate folds.
- Average the model performance over the K test sets.
- Report cross-validated metrics.



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Train Validation Test

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