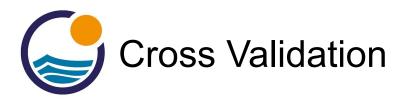


Cross Validation

Train | Test Split



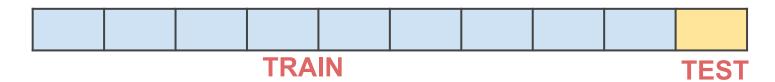


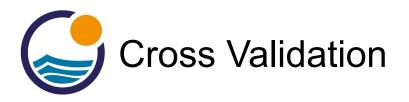
Begin with entire data set





Split into two sets Train and Test

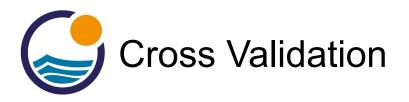




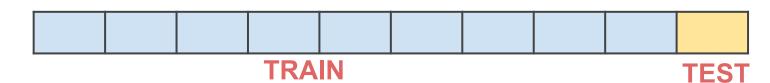
Train model then evaluate error on Test.

											ERROR
TRAIN								TEST			





We "allow" ourselves model adjustments.



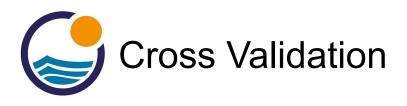




Cross Validation

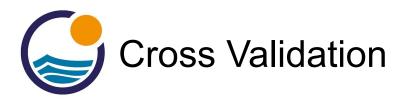
Train | Validation | Test Split





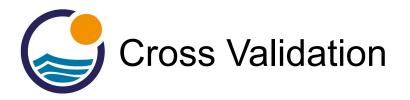
- We just saw that Train | Test split method has a disadvantage of not having a portion of data that can report a performance metric on truly "unseen" data.
- If we want a truly fair and final set of performance metrics, we should get these metrics from a final test set that we do not allow ourselves to adjust on.



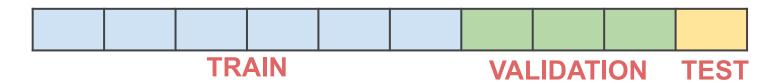


Begin with entire data set

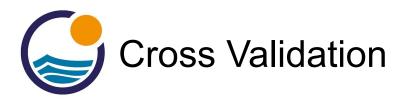




Split into: Train, Validation, and Test.

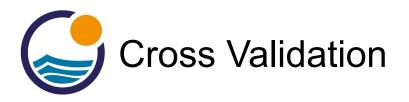






Set aside Test set for final metrics.

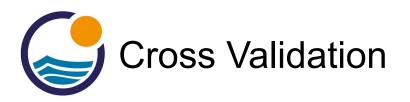




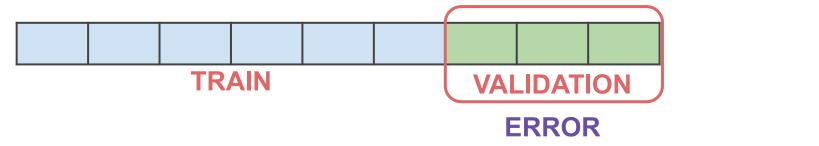
Fit model on Train set.





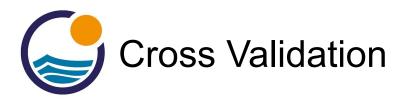


Evaluate performance on validation set.

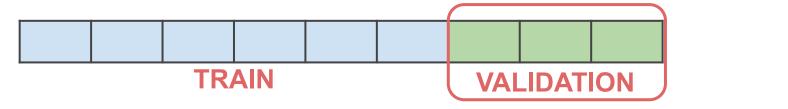




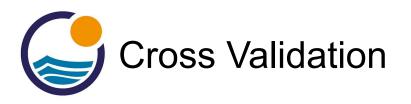
PIERIAN 🍪 DATA



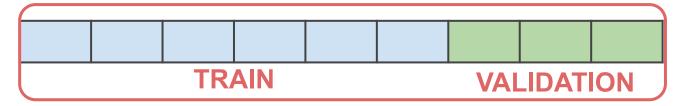
Adjust hyperparameters as needed...



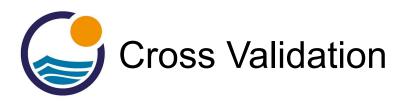




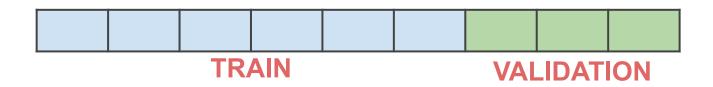
Train, validate, and adjust as necessary.



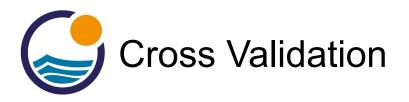




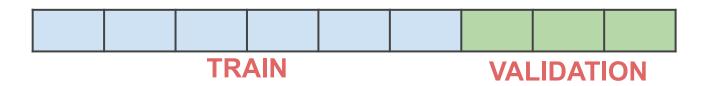
Perform final evaluation on Test set.





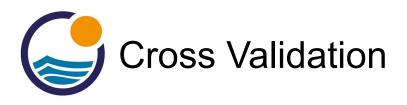


Do not adjust! Report this as final metric.



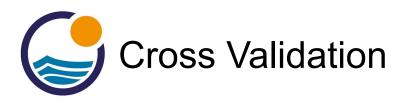






- Recall the entire reason to not adjust after the final test data set is to get the fairest evaluation of the model.
- The model was not fitted to the final test data and the model hyperparameters were not adjusted based off final test data.
- This is truly never before seen data!





- To achieve this in Python with Scikit-Learn we simply perform the train_test_split() function call twice.
 - Once to split off larger training set.
 - Second time to split remaining data into a validation set and test set.

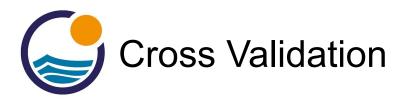




Cross Validation

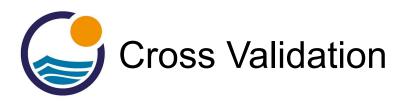
Using the cross_val_score function





Start with entire data set:





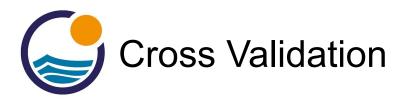
Split data Training data and Test Data:



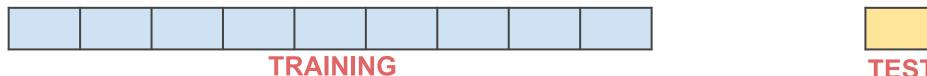


Split data Training data and Test Data:

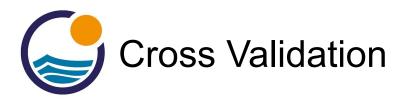




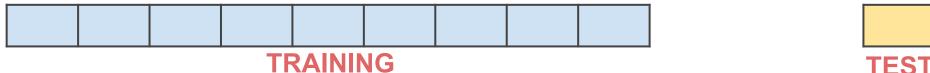
Remove Test data for final evaluation:



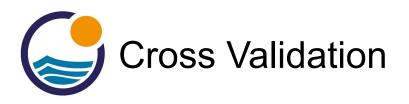




Choose K-Fold Split Value for Training Data



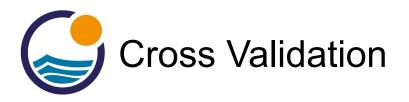




Recall larger K means more computation!



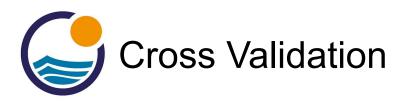
TES1



Choose K-Fold Split Value for Training Data



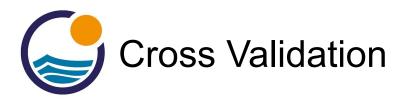
PIERIAN 🍪 DATA



• Here K = 5

TRAINING

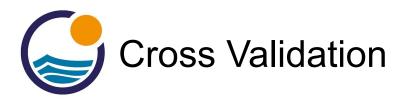




Train on K-1 folds and Validate on 1 Fold

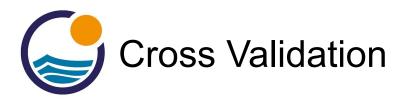






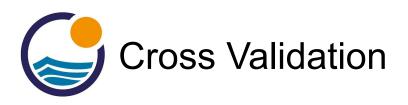
Obtain an error metric for this fold:

				ERROR 1	
TRAINING			VALIDATE		TEST



Repeat for another combination

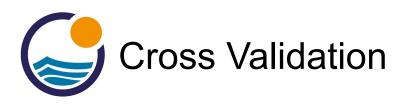
			ERROR 1	
			ERROR 2	TEST



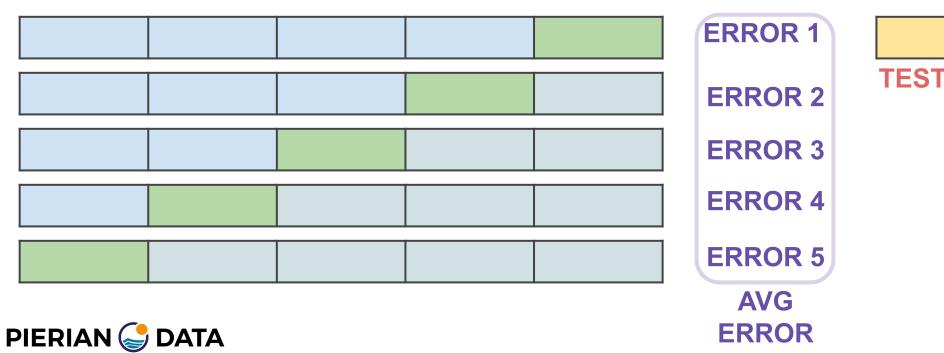
Continue for all fold combinations

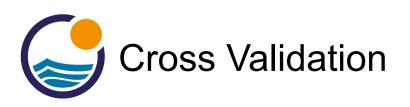
		ERROR 1	
		ERROR 2	TEST
		ERROR 3	
		ERROR 4	
		ERROR 5	





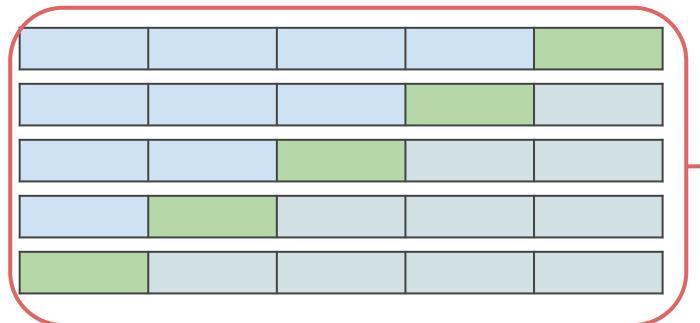
Use mean error for parameter adjustments



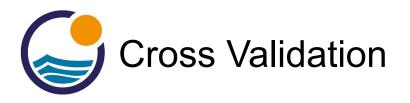


Get final metrics from final test set.

TEST







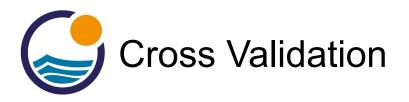
- The cross_val_score function uses a model and training set (along with a K and chosen metric) to perform all of this for us automatically!
- This allows for K-Fold cross validation to be performed on any model.



Cross Validation

Using the cross_validate function





 The cross_validate function allows us to view multiple performance metrics from cross validation on a model and explore how much time fitting and testing took.



Grid Search





- Often more complex models have multiple adjustable hyperparameters.
- A grid search is a way of training and validating a model on every possible combination of multiple hyperparameter options.





- Scikit-Learn includes a GridSearchCV class capable of testing a dictionary of multiple hyperparameter options through cross-validation.
- This allows for both cross-validation and a grid search to be performed in a generalized way for any model.

