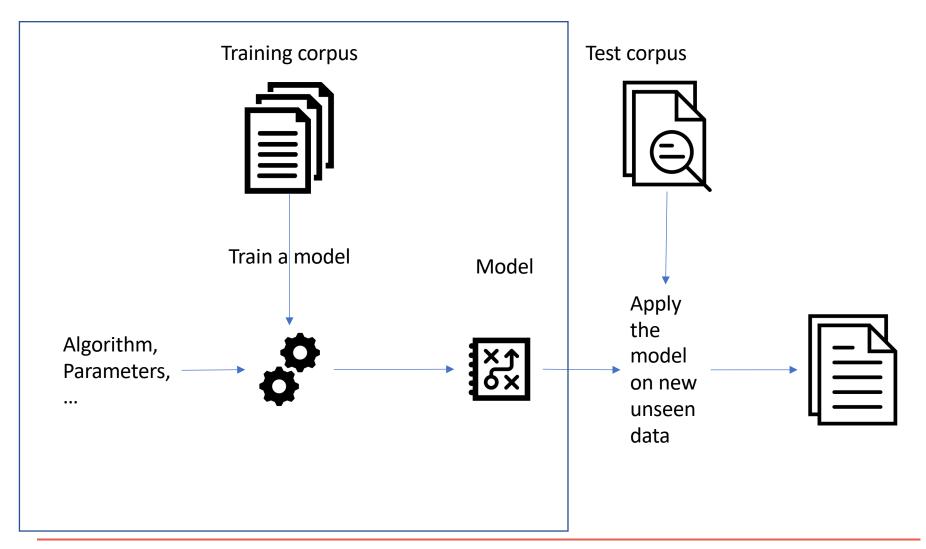


How do we know how good the model is?

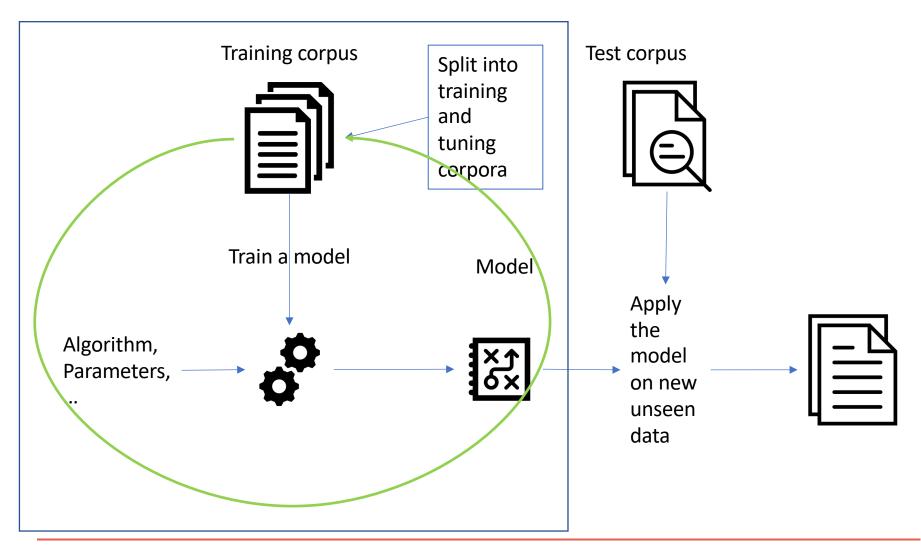


Tuning

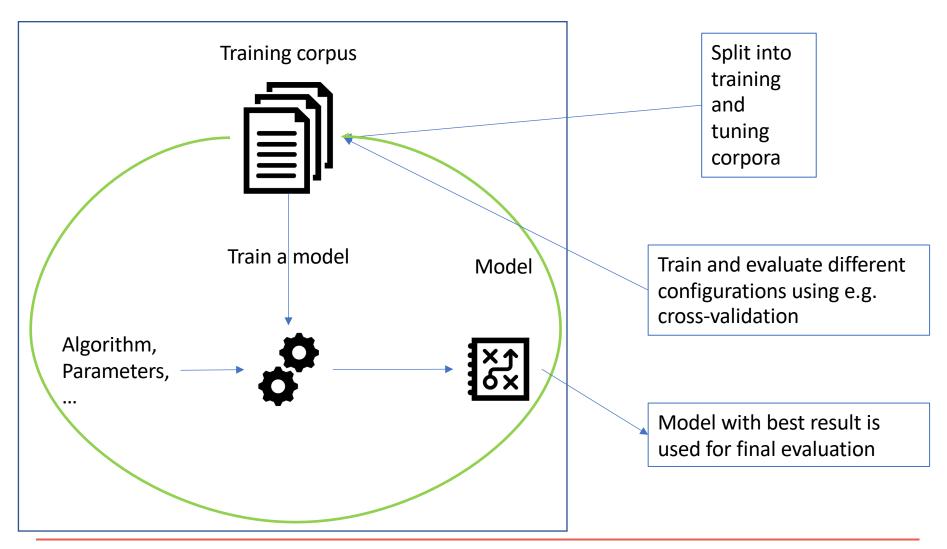
- Supervised machine learning algorithms require
 - parameters
 - different features and combinations
- How can we develop a model that we believe will work well on unseen data?









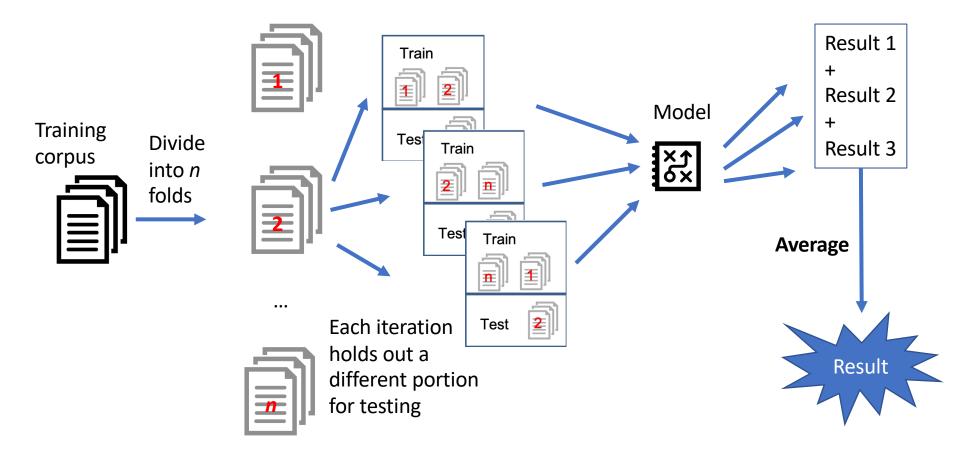




Cross-validation

- Splits the training data into n portions
- Each iteration, one nth is used for testing, the rest for training
- All results averaged

Cross-validation





Final evaluation

- Gold standard
 - To evaluate model performance, we need test data with the 'right answers'
 - This has to be different data than the training data!

Intrinsic evaluation

	Gold standard value = positive	Gold standard value = negative	
Predicted value = positive	True positive (TP)	False positive (FP)	PPV, precision: TP TP+FP
Predicted value = negative	False negative (FN)	True negative (TN)	
	TPR, Sensitivity, Recall: TP TP+FN	TNR, Specificity: TN TN+FP	F-Score: 2× PPV×TPR PPV+TPR
	Accuracy: TP+TN TP+TN+FP+FN		

Intrinsic evaluation

- Micro and macro average
 - Micro average is computed on all instances
 - Macro average is computed independently for each class and then averaged
 - Problematic if there is big class imbalance

Practicals

- We will use sklearn, nltk, spacy and jupyter notebooks today
 - You can try different machine learning algorithms, and you will work on evaluation in different ways.
- Other tools you can try
 - GATE has support for most supervised learning algorithms and allows for easy experimentation with other language features
 - Weka
 - •



Thank you!

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