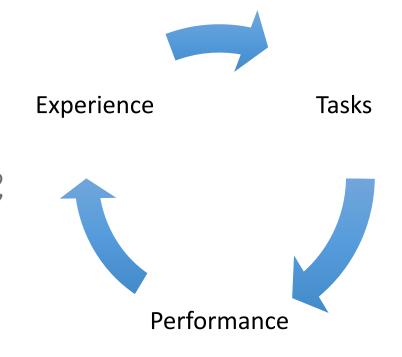
DIAG 2016 Machine Learning: Classification Techniques in R AJG Project

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What is Machine Learning?

"A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P, if its performance at tasks in T, as measured by P, improves with experience E."



- The humans provide : definition of Tasks (T) and measure of Performance (P)
- The machine leverages Experience (E) to get better at T as measured by P
- ML is also known as 'Statistical Learning'

What is Classification

- In machine learning terms, categorizing data points is a **classification** task.
 - Binary Classification involves two classes e.g. 0 or 1
 - Multi-class Classification involves multiple classes
- A classifier algorithm (procedure) implements a 'loss function' (performance) that is optimized using the training data (experience)
- This is known as Supervised Learning as the data is labelled
 - Unsupervised learning is used in clustering (e.g. Market Segmentation)

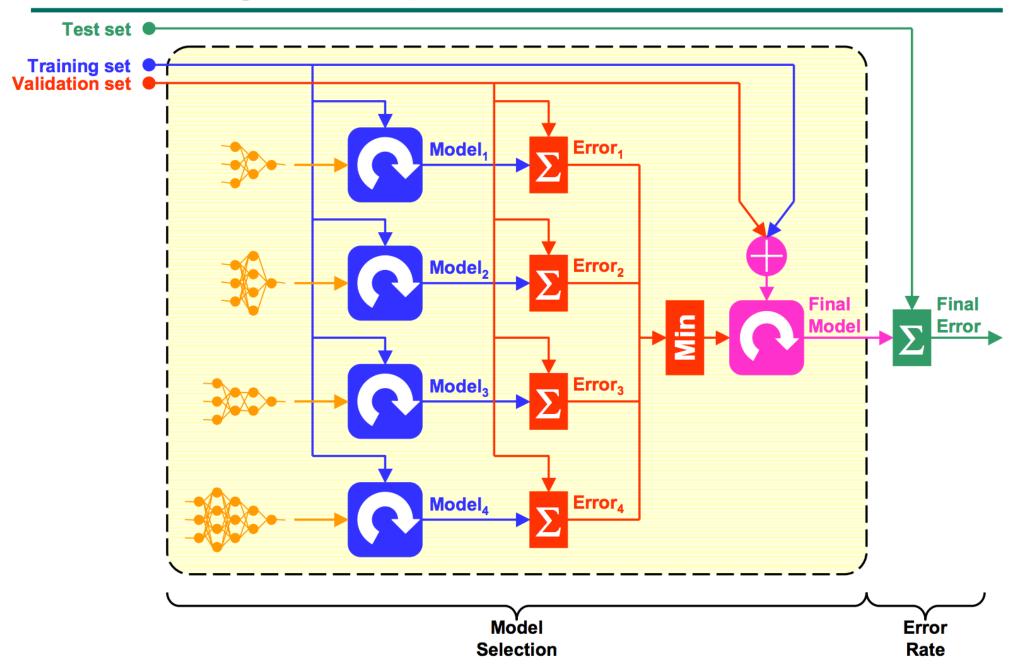
Different Kinds of Classification Methods

- Logistic Regression
- Classification and Regression Trees (CART)
 - Also known as 'Decision Trees' in some domains
- SVM
- Random Forest
- Etc.

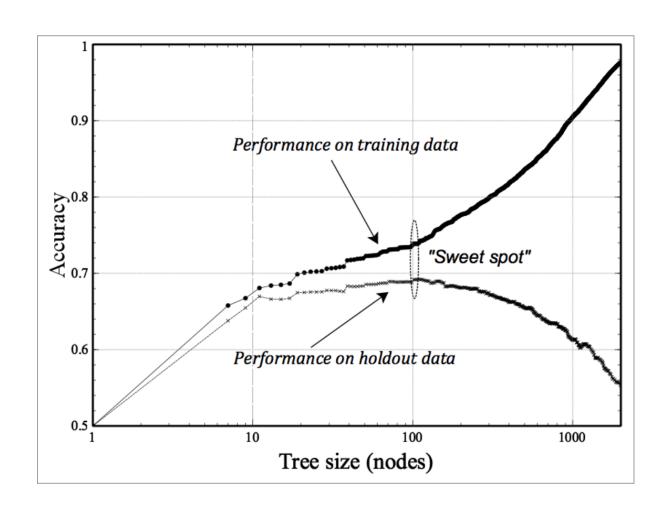
Overfitting is the WORST THING EVER

- Overfitting is when your model 'memorizes' the data set and uses the data to predict itself! That is USELESS
- To avoid this, data sets are split into three:
 - Training Set is a set of labelled data used to train a Machine Learning Model
 - Validation Set is a set of labelled data used to measure the Model Performance
 - Test Set is a set of un-labelled data used to make predictions via the trained and validated model

Three-way data splits



Tune to max(perf) && min(overfitting)



How to measure model performance

Confusion Matrix!

	Predicted:	Predicted:
n=165	NO	YES
Actual:		
NO	50	10
Actual:		
YES	5	100

- true positives (TP): These are cases in which we predicted yes (they have the disease), and they do have the disease.
- true negatives (TN): We predicted no, and they don't have the disease.
- false positives (FP): We predicted yes, but they don't actually have the disease. (Also known as a "Type I error.")
- false negatives (FN): We predicted no, but they actually do have the disease. (Also known as a "Type II error.")
- http://www.dataschool.io/simple-guide-to-confusion-matrix-terminology/

Model Performance

- Accuracy: Overall, how often is the classifier correct?
 - (TP+TN)/total = (100+50)/165 = 0.91
- Misclassification Rate: Overall, how often is it wrong?
 - (FP+FN)/total = (10+5)/165 = 0.09
 - equivalent to 1 minus Accuracy
 - also known as "Error Rate"
- True Positive Rate: When it's actually yes, how often does it predict yes?
 - TP/actual yes = 100/105 = 0.95
 - also known as "Sensitivity" or "Recall"
- False Positive Rate: When it's actually no, how often does it predict yes?
 - FP/actual no = 10/60 = 0.17
- Specificity: When it's actually no, how often does it predict no?
 - TN/actual no = 50/60 = 0.83
 - equivalent to 1 minus False Positive Rate

Interactive Tutorial

- Lets do the Kaggle 'Titanic Challenge'
 - You can do it yourself later on www.kaggle.com
- Predict which Passengers survived the Titanic disaster

Questions?

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