



# Welcome to the course!

Sergey Fogelson VP of Analytics, Viacom



## Before we get to XGBoost...

- Need to understand the basics of
  - Supervised classification
  - Decision trees
  - Boosting



## Supervised learning

- Relies on labeled data
- Have some understanding of past behavior



### Supervised learning example

Does a specific image contain a person's face?



- Training data: vectors of pixel values
- Labels: 1 or 0



## Supervised learning: Classification

• Outcome can be binary or multi-class



### Binary classification example

• Will a person purchase the insurance package given some quote?





### Multi-class classification example

Classifying the species of a given bird

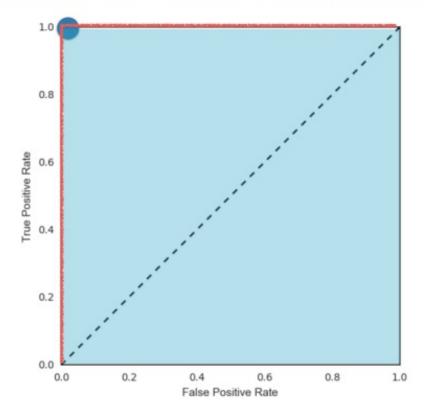




## AUC: Metric for binary classification models

### Area under the ROC curve (AUC)

Larger area under the ROC curve = better model





### Accuracy score and confusion matrix

Confusion matrix

Actual: Spam Email

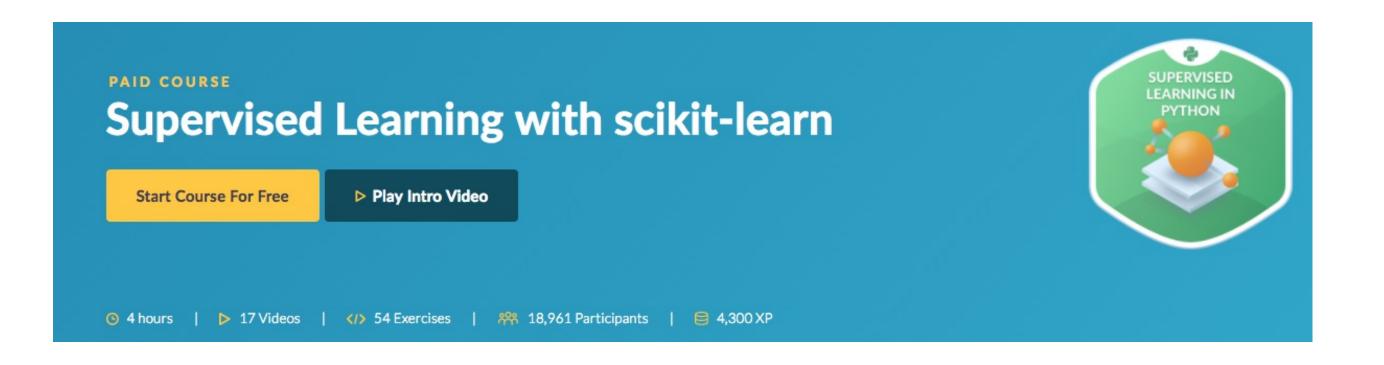
Actual: Real Email

Predicted: Spam Email	Predicted: Real Email
True Positive	False Negative
False Positive	True Negative

ullet Accuracy:  $rac{tp+tn}{tp+tn+fp+fn}$ 



### Supervised learning with scikit-learn





### Other supervised learning considerations

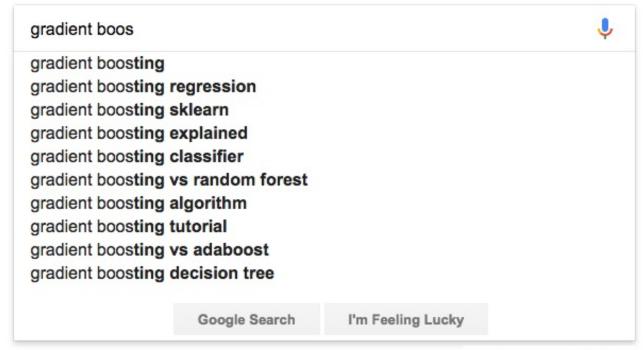
- Features can be either numeric or categorical
- Numeric features should be scaled (Z-scored)
- Categorical features should be encoded (one-hot)



### Ranking

• Predicting an ordering on a set of choices





Report inappropriate predictions



### Recommendation

- Recommending an item to a user
- Based on consumption history and profile
- Example: Netflix





# Let's get to work!





# Introducing XGBoost

Sergey Fogelson VP of Analytics, Viacom



### What is XGBoost?

- Optimized gradient-boosting machine learning library
- Originally written in C++
- Has APIs in several languages:
  - Python
  - R
  - Scala
  - Julia
  - Java



### What makes XGBoost so popular?

- Speed and performance
- Core algorithm is parallelizable
- Consistently outperforms single-algorithm methods
- State-of-the-art performance in many ML tasks



### Using XGBoost: A Quick Example

```
In [1]: import xgboost as xgb
In [2]: import pandas as pd
In [3]: import numpy as np
In [4]: from sklearn.model selection import train test split
In [5]: class data = pd.read csv("classification data.csv")
In [6]: X, y = class_data.iloc[:,:-1], class_data.iloc[:,-1]
In [7]: X_train, X_test, y train, y test= train test split(X, y,
        test size=0.2, random state=123)
In [8]: xg cl = xgb.XGBClassifier(objective='binary:logistic',
        n estimators=10, seed=123)
In [9]: xg cl.fit(X train, y train)
In [10]: preds = xg cl.predict(X test)
In [11]: accuracy = float(np.sum(preds==y test))/y test.shape[0]
In [12]: print("accuracy: %f" % (accuracy))
accuracy: 0 78333
```





# Let's begin using XGBoost!



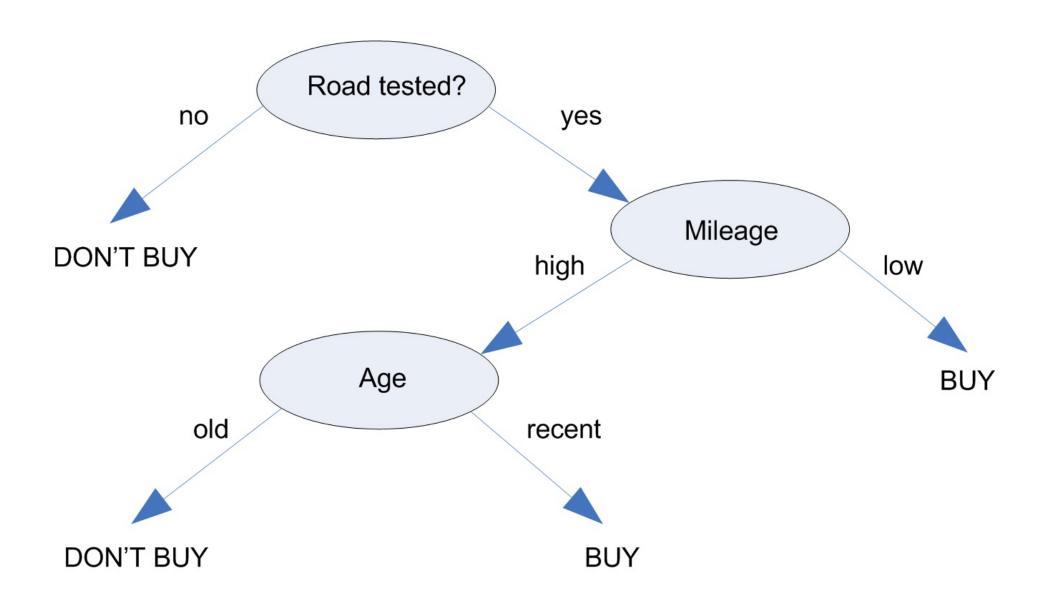


# What is a decision tree?

Sergey Fogelson Instructor



## Visualizing a decision tree





#### Decision trees as base learners

- Decision trees are generally used as base learners in XGBoost
- Base learner Individual learning algorithm in an ensemble algorithm

- Composed of a series of binary questions
- Predictions happen at the "leaves" of the tree



#### Decision trees and CART

- Constructed iteratively (one decision at a time)
  - Until a stopping criterion is met
- Individual decision trees are low-bias, high-variance models
  - Tendency to overfit
- XGBoost uses CART Classification and Regression Tree
  - Each leaf always contains a real-valued score
  - Can later be converted into categories





# Let's practice!





## What is Boosting?

Sergey Fogelson VP of Analytics, Viacom



## Boosting overview

- Not a specific machine learning algorithm
- Concept that can be applied to a set of machine learning models
  - "Meta-algorithm"
- Ensemble meta-algorithm used to convert many weak learners into a strong learner



### Weak learners and strong learners

- Weak learner: ML algorithm that is slightly better than chance
  - Example: Decision tree whose predictions are slightly better than
     50%
- Boosting converts a collection of weak learners into a strong learner
- Strong learner: Any algorithm that can be tuned to achieve good performance

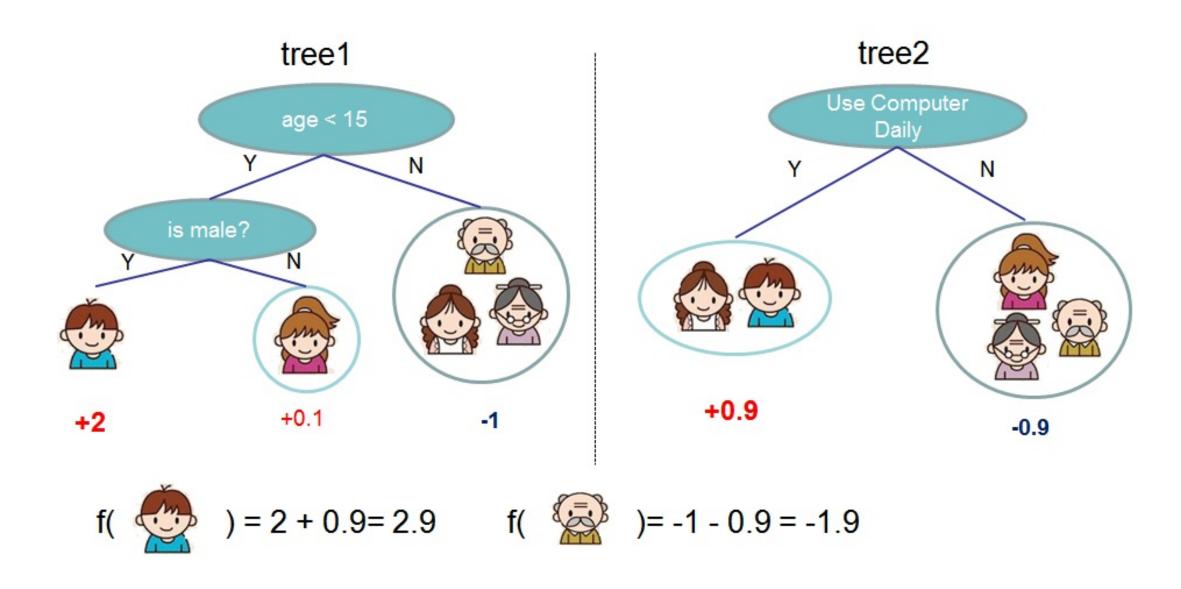


## How boosting is accomplished

- Iteratively learning a set of weak models on subsets of the data
- Weighing each weak prediction according to each weak learner's performance
- Combine the weighted predictions to obtain a single weighted prediction
- ... that is much better than the individual predictions themselves!



### Boosting example





### Model evaluation through cross-validation

- Cross-validation: Robust method for estimating the performance of a model on unseen data
- Generates many non-overlapping train/test splits on training data
- Reports the average test set performance across all data splits



### Cross-validation in XGBoost example





# Let's practice!





# When should I use XGBoost?

Sergey Fogelson VP of Analytics, Viacom



#### When to use XGBoost

- You have a large number of training samples
  - Greater than 1000 training samples and less 100 features
  - The number of features < number of training samples
- You have a mixture of categorical and numeric features
  - Or just numeric features



#### When to NOT use XGBoost

- Image recognition
- Computer vision
- Natural language processing and understanding problems
- When the number of training samples is significantly smaller than the number of features





# Let's practice!