# 16. Machine Learning Introduction

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#### Outline

- Objectives
- Machine Learning
- Supervised Learning
- Unsupervised Learning
- Other Machine Learning Algorithms
- Approach to making predictions
- Using SKLearn
- Refresher on probability
- Evaluating Classification Problems
- Evaluating Regression Problems

#### **Objectives**

- Understand broadly the concept of machine learning
- Understand the difference between supervised and unsupervised learning
- Understand data pre-processing

#### Machine Learning

- Basically using mathematical approaches to make predictions
- These predictions may be based on some prior knowledge captured in the data
- Where we use labels within the data to determine what unlabeled data points are we call that supervised learning
- Where there are no labels but we want to determine differences in the data we call that unsupervised learning
- Machine learning is the basis for deep learning

#### Supervised Learning

- These are based on labels that exist
- The variables we want to predict are called labels while those used in predicting are called features
  - Labels vs Features
  - Dependent variables vs Independent variables
  - Outcome vs Exposure
- When the label is categorical we use classification algorithms
- When the label is numeric we use regression algorithms

#### Unsupervised Learning

- These are not based on labels
- The algorithms attempt to find similarities and differences in the data
- Examples
  - Clustering algorithms
  - Dimensionality reduction algorithms (Principal Component Analysis and Linear Discriminant Analysis)

### Other Machine Learning Algorithms

- Ensemble algorithms
  - Random Forest Regression (bagging algorithm)
  - Random Forest Classification (bagging Algorithm)
  - XGBoost (boosting algorithm)
- Recommender systems
  - Facebook friend suggestion
  - Netflix movie recommendation
- More!

#### Approach to making predictions

- Usually you need to have thee parts of the data
  - Training set
  - Validation set
  - Test set
- The training-validation-testing lifecycle
- Be careful of overfitting

## Using SKLearn

- The parent module for machine learning
- Lets refer to the features as x and the labels as y
- Preprocessing
  - Import module: from sklearn.cross\_validation import train\_test\_split
  - 2. Create the split:
     xtrain, xtest, ytrain, ytest = train\_test\_split(x, y, test\_size=0.75)
- Using the algorithm
  - 1. Import the required model from a family from sklearn.family import Model
  - 2. Instantiate the model
    model = Model()
  - 3. Fit the model:
     model.fit(xtrain, ytrain)
  - 4. Predict:
    model.predict(xtest)

## Refresher on probability

	Actual True	Actual False
Predited True	a	b
Predicted False	c	d

- $P(PredictedTrue \cap ActualTrue) = \mathbf{a} = True \ Positive$
- $P(PredictedTrue \cap ActualFalse) = \mathbf{b} = False \ Positive$
- $P(PredictedFalse \cap ActualTrue) = \mathbf{c} = False \ Negative$
- $P(PredictedFalse \cap ActualFalse) = \mathbf{d} = True \ Negative$

## **Evaluating Classification Problems**

Table 2: Confusion Matrix

	Actual True	Actual False
Predited True	a	b
Predicted False	$^{\mathrm{c}}$	d

- Proportion of correct predictions =  $\frac{a+d}{a+b+c+d}$  = Accuracy
- $P(PredictedTrue|ActualTrue) = \frac{a}{a+c} = Recall$
- $P(ActualTrue|PredictedTrue) = \frac{a}{a+b} = Precision$

• Harmonic mean of Recall and  $Precision = 2 * \frac{Recall * Precision}{Recall + Precision} = F1 Score$ 

# **Evaluating Regression Problems**

- Regression problems are those supervised learning problems in which the label is numeric (continuous)
- Mean Absolute Error (MAE) =  $\frac{1}{n} \sum_{i=1}^{n} y_i \hat{y}_i$
- Mean Square Error (MSE) =  $\frac{1}{n} \sum_{i=1}^{n} (y_i \hat{y}_i)^2$
- Root Mean Square Error (RMSE) =  $\sqrt{\frac{1}{n} \sum_{i}^{n} (y_i \hat{y}_i)^2}$

# Review of objectives

- Understand broadly the concept of machine learning
- Understand the difference between supervised and unsupervised learning
- Understand data pre-processing

## Q&A

There's no gist of the day!

Thanks for being part of the conversation!