

#### **Session 7**

FIT5202 Big Data Processing

K-Means Clustering Model Selection



# Week 7 Agenda

- Part A
- Session 5 Review
- K-means Clustering
  - Shilouette Score
- Tutorial Instructions
  - Use case : Identify if 3 hackers were involved

- Part B
- Model Selection
  - Hyperparameter Tuning
  - Cross Validation
    - K-fold Cross Validation
  - TrainValidationSplit
- Model Persistence
  - Saving and Loading a Model



# **K-Means Clustering**

Finds groups (or clusters) of data

A cluster comprises a number of "similar" objects

A member is closer to another member within the same group than to a member of a different group

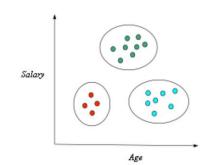
Groups have no category or label

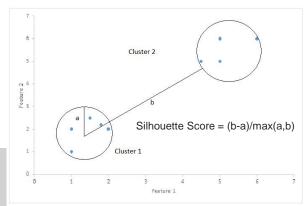
Unsupervised learning

Animation Demo (https://www.naftaliharris.com/blog/visualizing-k-means-clustering)

**Silhouette Score** (-1 to 1): calculates the goodness of a clustering technique

- 1 means Clusters are well apart from each other and clearly distinguishes
- 0 means clusters are not clearly distinguished, the distance between the clusters is not significant
- -1 means clusters are assigned in the wrong way





#### **Model Selection**

#### Hyperparameter Tuning

- Hyper-parameters are not model parameters: they cannot be trained from the data
- Hyperparameter tuning: choosing a set of optimal hyperparameters for a learning algorithm
- model.getParamMap() to get the list of hyperparameters for the model

```
from pyspark.ml.classification import DecisionTreeClassifier

# Extracts the number of nodes in the decision tree and the tree depth in the model and stores
dt = DecisionTreeClassifier(featuresCol = 'features', labelCol = 'label', maxDepth = 3)
dtModel = dt.fit(train)
```

```
Hyperparameters
           n iter
  test size
             max depth
random state
                n neighbors
 alpha
                     gamma
  n components
                 metric
      kernel
             n folds
        penalty
                   CV
```



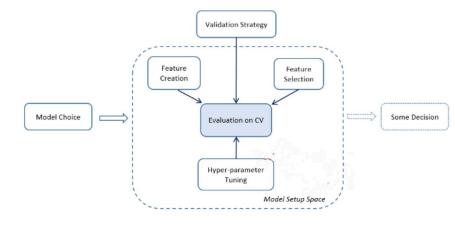
#### **Model Selection**

All models are wrong; some are useful (George E.P. Box)

- Finding the best model or parameters
- Tuning can be done for individual Estimators or the entire Pipeline

Model selection for Mlib has the following tools:

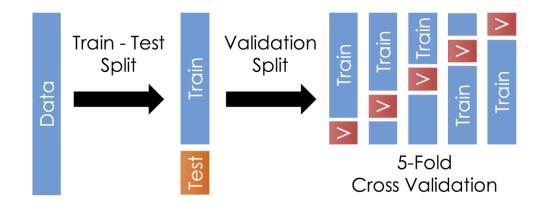
- 1. CrossValidator
- 2. TrainValidationSplit





#### **Cross Validation (K-Fold)**

- Splitting dataset into a set of folds, which are used as separate training and test datasets.





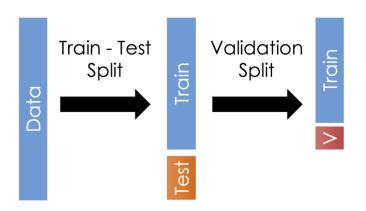
## **Cross Validation (Decision Tree)**

```
from pyspark.ml.tuning import ParamGridBuilder, CrossValidator, CrossValidatorModel
from pyspark.ml.evaluation import BinaryClassificationEvaluator
# Create ParamGrid for Cross Validation
dtparamGrid = (ParamGridBuilder()
             .addGrid(dt.maxDepth, [2, 5, 10, 20, 30])
             .addGrid(dt.maxBins, [10, 20, 40, 80, 100])
             .build())
dtevaluator = BinaryClassificationEvaluator(rawPredictionCol="rawPrediction")
dtcv = CrossValidator(estimator = pipeline,
                        estimatorParamMaps = dtparamGrid,
                        evaluator = dtevaluator,
                        numFolds = 3)
dtcvModel = dtcv.fit(train)
hestModel= dtcvModel.bestModel
print('Best Param (regParam): ', bestModel.stages[-1]. java obj.paramMap())
Best Param for DT: {
       DecisionTreeClassifier ba35db4d44b0-featuresCol: features,
       DecisionTreeClassifier ba35db4d44b0-labelCol: label,
       DecisionTreeClassifier ba35db4d44b0-maxBins: 20.
       DecisionTreeClassifier_ba35db4d44b0-maxDepth: 20
```



## **TrainValidationSplit**

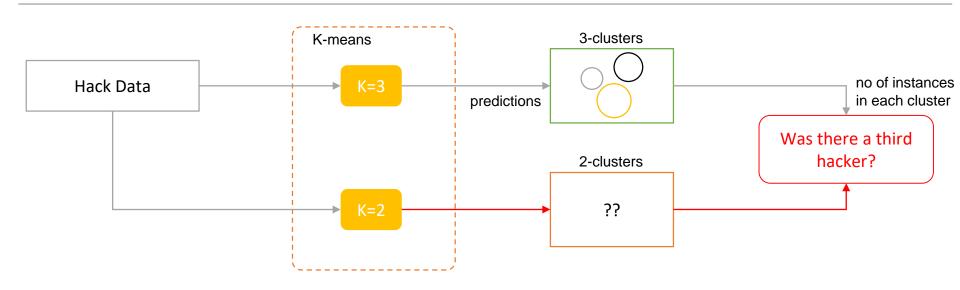
- Creates a single dataset pair
- Only evaluates each combination of parameter once as opposed to k-times in case of CrossValidator
- Less expensive but not reliable if the training dataset is not large enough





# Use case: Was there a third hacker?

**Assumption**: Hackers trade off attacks equally





#### **Thank You!**

See you next week.