

CIML Summer Institute 2022

Spark Hands-On Exercises



Spark Hands-On

Mai H. Nguyen, Ph.D.

Setup

- **Login to Expanse**

- Open terminal window on local machine
- `ssh login.expanse.sdsc.edu -l <xdtr_account>`

- **Pull latest from repo**

- `git pull`
- URL:

<https://github.com/ciml-org/ciml-summer-institute-2022>

Server Setup for PySpark - Portal

- **Expanse Portal**

- <https://portal.expanse.sdsc.edu>

- **Parameters**

- Account: sds184
- Time limit (min): 240
- Number of cores: 4
- Memory required per node: 16 GB
- GPUs: 0
- Singularity image:
/cm/shared/apps/containers/singularity/ciml/2022/pyspark-latest.sif
- Environment module: singularitypro
- Reservation: ciml-day3
- Working directory: home
- Type: JupyterLab

Server Setup for PySpark - Command Line

- **In terminal window**

- jupyter-shared-spark

- Alias for: `galileo launch --account ${CIML_ACCOUNT} --reservation ${CIML_RESERVATION_CPU} --partition shared --cpus 4 --memory 16 --time-limit 04:00:00 --env-modules singularitypro --sif /cm/shared/apps/containers/singularity/ciml/2022/pyspark-latest.sif --bind /exppanse,/scratch,/cm --quiet`

- **To check queue**

- `squeue -u $USER`

Data Setup

- In terminal window in Jupyter Lab, do the following
- Go to your home directory
 - cd
 - pwd # Should see /home/\$USER
- Get data
 - mkdir data
 - cp /cm/shared/examples/sdsc/ciml/2022/BookReviews_5M.txt.zip .
 - unzip BookReviews_5M.txt.zip
 - cp /cm/shared/examples/sdsc/ciml/2022/minute_weather.csv.zip .
 - unzip minute_weather.csv.zip
 - ls # Should see BookReviews_5M.txt & minute_weather.csv

PySpark Scaling Hands-On

- **Data**
 - BookReviews_5M.txt
 - Source : <https://jmcauley.ucsd.edu/data/amazon/>
- **Notebook**
 - pyspark_demo.ipynb
 - Location: <ciml_dir>/3.6_spark
- **To do**
 - Change number of cores: 1, 2, 4
 - Note difference in execution times

GETTING EXECUTION TIMES

- In notebook, execution time is printed out in cell before Spark session is stopped (next to last cell)
- Need to restart the kernel and run all cells without stopping to get accurate execution time:
 - Run -> Restart Kernel and Run All Cells
- Find mean and standard deviation of execution times over 3 runs for
 - 1 core, 2 cores, and 4 cores

```
import pyspark
from pyspark.sql import SparkSession
```

```
conf = pyspark.SparkConf().setAll([
    ('spark.master', 'local[2]'),
    ('spark.app.name', 'PySpark Demo)])
spark = SparkSession.builder.config(conf=conf).getOrCreate()
```

Specify number of cores.
“*” uses all available cores



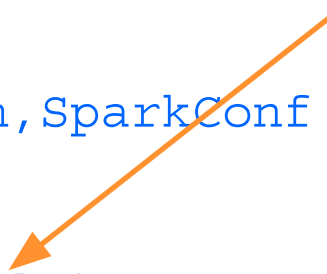
SPARK SESSION

```
import pyspark
from pyspark.sql import SparkSession, SparkConf

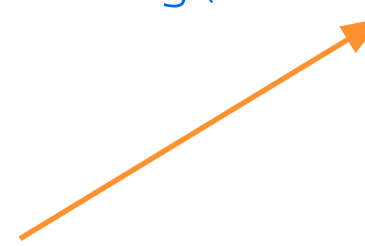
conf = SparkConf().setAll([
    ('spark.master', 'local[*]'),
    ('spark.app.name', 'PySpark Demo')])

spark = SparkSession.builder.config(conf=conf).getOrCreate()
```

Use * to use all available cores, or integer value to specify number of cores to use



Configuration parameters for Spark session



Get existing Spark session or create new one



SPARK PROGRAM STRUCTURE

- **Start Spark session**
 - `spark = SparkSession.builder.config(conf=conf).getOrCreate()`
- **Create distributed dataset**
 - `df = spark.read.csv("data.csv",header="True")`
- **Apply transformations**
 - `new_df = df.filter(col("dept") == "Sales")`
- **Perform actions**
 - `df.collect()`
- **Stop Spark session**
 - `spark.stop()`

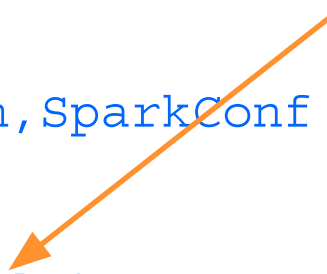
START SPARK SESSION

```
import pyspark
from pyspark.sql import SparkSession, SparkConf

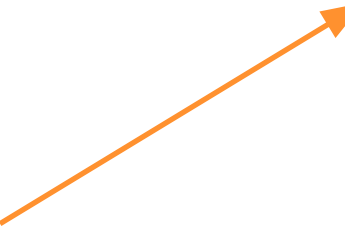
conf = SparkConf().setAll([
    ('spark.master', 'local[*]'),
    ('spark.app.name', 'PySpark Demo')])

spark = SparkSession.builder.config(conf=conf).getOrCreate()
```


Use * to use all available cores, or integer value to specify number of cores to use



Configuration parameters for Spark session



Get existing Spark session or create new one



LOAD DATA

```
df = spark.read.text("file.txt")
```

```
df = spark.read.csv("file.csv",  
                    header=True  
                    inferSchema=True).cache()
```

Indicates whether
column headers exist



Automatically infer
data types of columns



Cache data



LOAD DATA

- Loading data from local file system

```
df = spark.read.text("file://<path>/<file>.txt")
```

```
df = spark.read.csv("file://<path>/<file>.csv",  
                    header=True).cache()
```

- Loading data from HDFS

```
df = spark.read.text("hdfs://<path>/<file>.txt") \  
    .cache()
```

```
df = spark.read.csv("hdfs://<path>/<file>.csv",  
                    header=True,  
                    inferSchema=True).cache()
```

Indicates whether
column headers exist



Automatically infer
data types of columns



Cache data



CHAINING

```
text_file = sc.textFile("file.txt")

counts = text_file.flatMap(lambda line: line.split(" ")) \
    .map(lambda word: (word, 1)) \
    .reduceByKey(lambda a, b: a + b)

counts.saveAsTextFile("output.txt")
```

**Chaining: Making multiple
method calls on same object**



**Line continuation
indicator**



RDD Wordcount

DROP ROWS WITH NULLS

- **Drop rows with null values**

```
df.dropna()  
df.dropna(how='any')  
df.dropna(how='all')
```

- **Check number of rows before and after dropping rows**

```
df.count()
```

CREATE FEATURE VECTOR COLUMN

- **Create feature vector column**
 - Combines given list of columns into single vector column
 - To feed data to machine learning models

```
from pyspark.ml.feature import VectorAssembler
```

```
features = ['air_temp', 'relative_humidity']  
assembler = VectorAssembler(inputCols=features,  
                             outputCol='featureVector')
```

```
features_df = assembler.transform(df)  
features_df.show()
```

air_temp	relative_humidity
62.96	63.9



air_temp	relative_humidity	featureVector
62.96	63.9	[62.96, 63.9]

New column
appended to
features_df

PARTITION DATA

- Partition available data into train and test data sets

```
train_df, test_df = df.randomSplit(0.8, 0.2), seed=<seed>)
```

Percentage of
samples for train
dataset



Percentage of
samples for test
dataset



SCALE DATA

- **Scale input data values**

- Standardize values to have zero mean and unit standard deviation
- Each feature is scaled separately
- Create scale transformer using train data, then apply to train/test data

```
from pyspark.ml.feature import StandardScaler
scaler = StandardScaler(inputCol="features_unscaled",
                        outputCol="features_scaled",
                        withStd=True, withMean=True)
scalerModel = scaler.fit(train_df)
```

```
train_df = scalerModel.transform(train_df)
test_df  = scalerModel.transform(test_df)
```

BUILD MODEL

- **Build decision tree classifier**

- Create model
- Use fit() to train model

```
from pyspark.ml.classification import DecisionTreeClassifier
```

```
dt = DecisionTreeClassifier(  
    featuresCol='featureVector',  
    labelCol='label',  
    predictionCol='prediction')
```

Can specify name of
columns for features,
labels, and predictions



```
dt_model = dt.fit(<train>)
```

BUILD MODEL

- **Build k-means model**

- Create model
- Use fit() to train model

```
from pyspark.ml.clustering import KMeans
```

```
kmeans = KMeans(  
    k=<num_clusters>, seed=<seed_value>)
```

```
model = kmeans.fit(<data>)
```

APPLY MODEL

- **Apply trained model**
 - Use transform()

```
predictions = <model>.transform(<data>)
```

Will have <data> and
new 'prediction'
column appended at
the end



Trained
model



Input data



EVALUATE CLASSIFICATION MODEL

- **Evaluator for classification model**
 - Calculates F1, precision, recall, accuracy

```
from pyspark.ml.evaluation import MulticlassClassificationEvaluator

mc_evaluator = MulticlassClassificationEvaluator(
    predictionCol='prediction',
    labelCol='label',
    metricName='f1')

mc_evaluator.evaluate(<predictions>)
```

Column with predictions

Column with labels

Evaluation metric

Contains predictions and labels

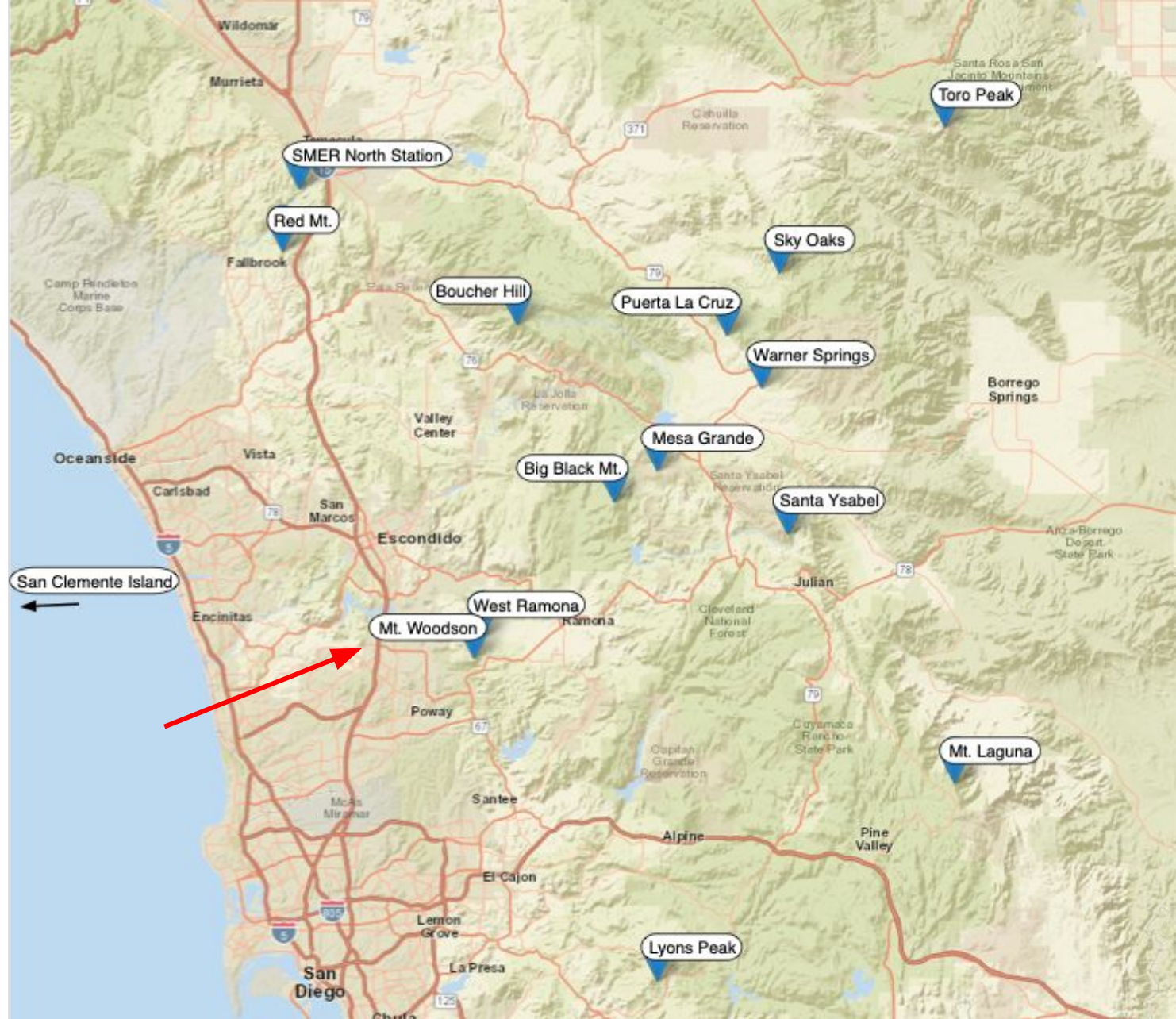
PySpark Cluster Analysis Hands-On

- **Data**
 - Weather station measurements
- **Task**
 - Perform cluster analysis to identify different weather patterns
- **Approach**
 - Spark k-means
- **Notebook**
 - Location: <ciml_dir>/3.6_spark
 - pyspark-clustering.ipynb
 - pyspark-clustering-wOutput.ipynb
 - pyspark-clustering-soln.ipynb

Dataset Description

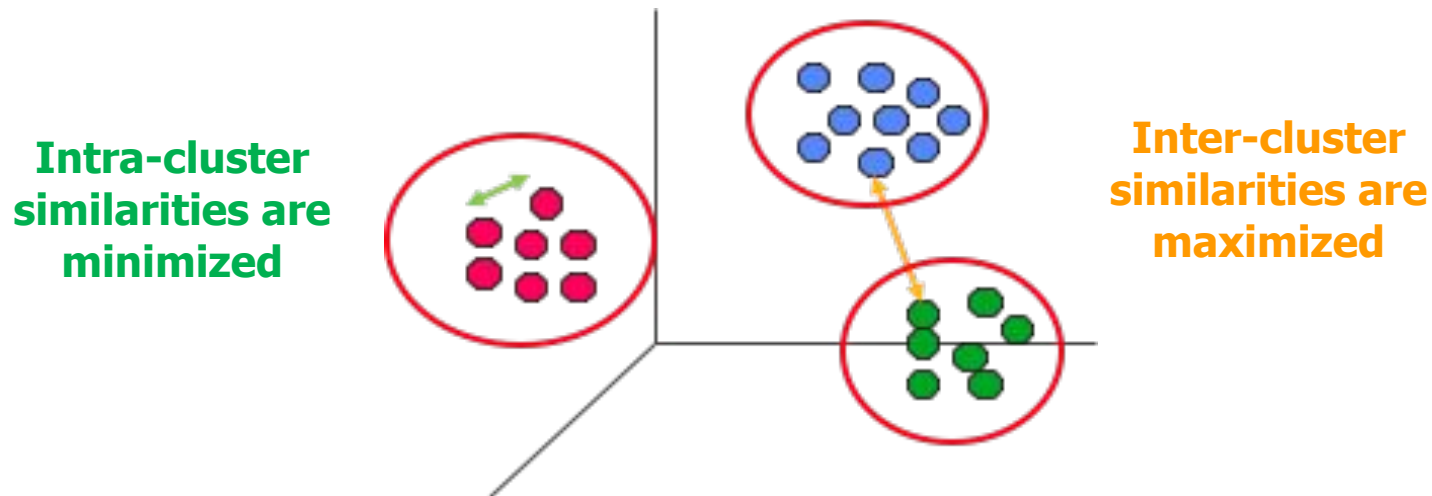
- **Measurements from weather station on Mt. Woodson, San Diego**
- **Air temperature, humidity, wind speed, wind direction, etc.**
- **Three years of data: Sep. 2011 - Sep. 2014**
 - minute_weather.csv: measurement every minute
- **Source**
 - <http://hpwren.ucsd.edu>

Map of HPWREN Weather Stations



Cluster Analysis

- **Cluster analysis divides data into groups**
 - Grouping is based on some similarity measure.
 - Samples within a cluster are more similar to each other than to samples in other clusters.



<http://www-users.cs.umn.edu/~kumar/dmbook/index.php>

k-Means Clustering

- **Partitional**
 - Clusters are divided into non-overlapping subsets
- **Centroid-Based**
 - Cluster represented by central vector
- **Simple, classic clustering technique**
 - Data points are grouped into k clusters
 - Cluster defined by cluster mean

- **Algorithm**

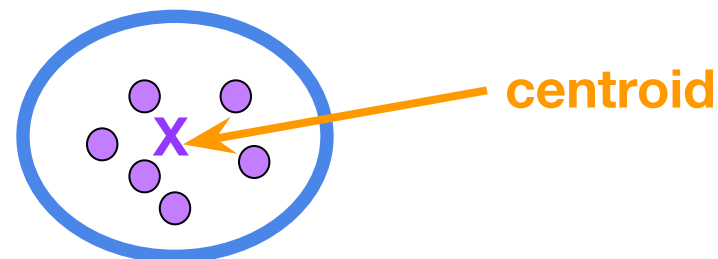
Select k initial *centroids* (cluster centers)

Repeat

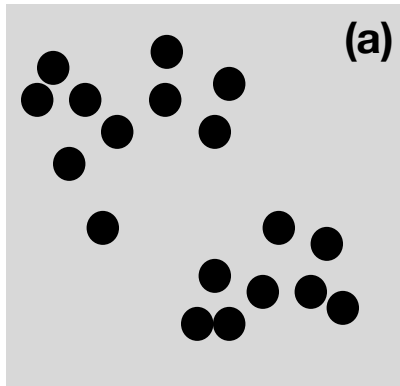
Assign each sample to closest centroid

Calculate mean of cluster to determine new centroid

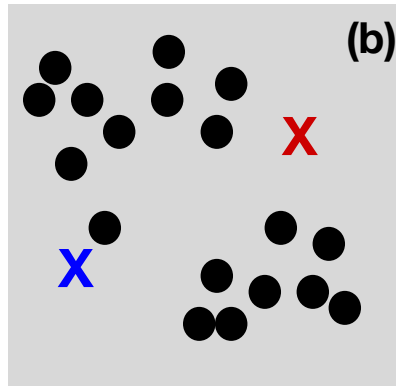
Until some stopping criterion is reached



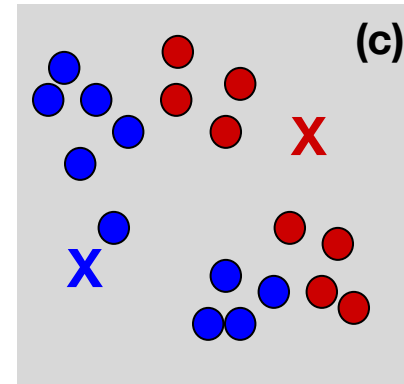
k-Means Clustering Illustration



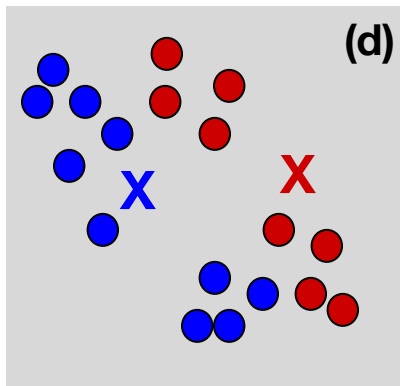
Original samples



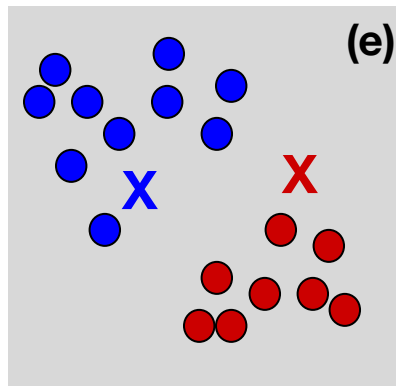
Initial Centroids



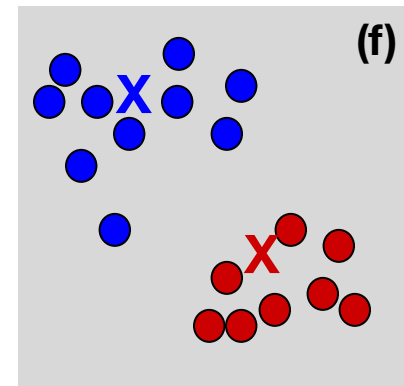
Assign Samples



Re-calculate Centroids



Assign Samples

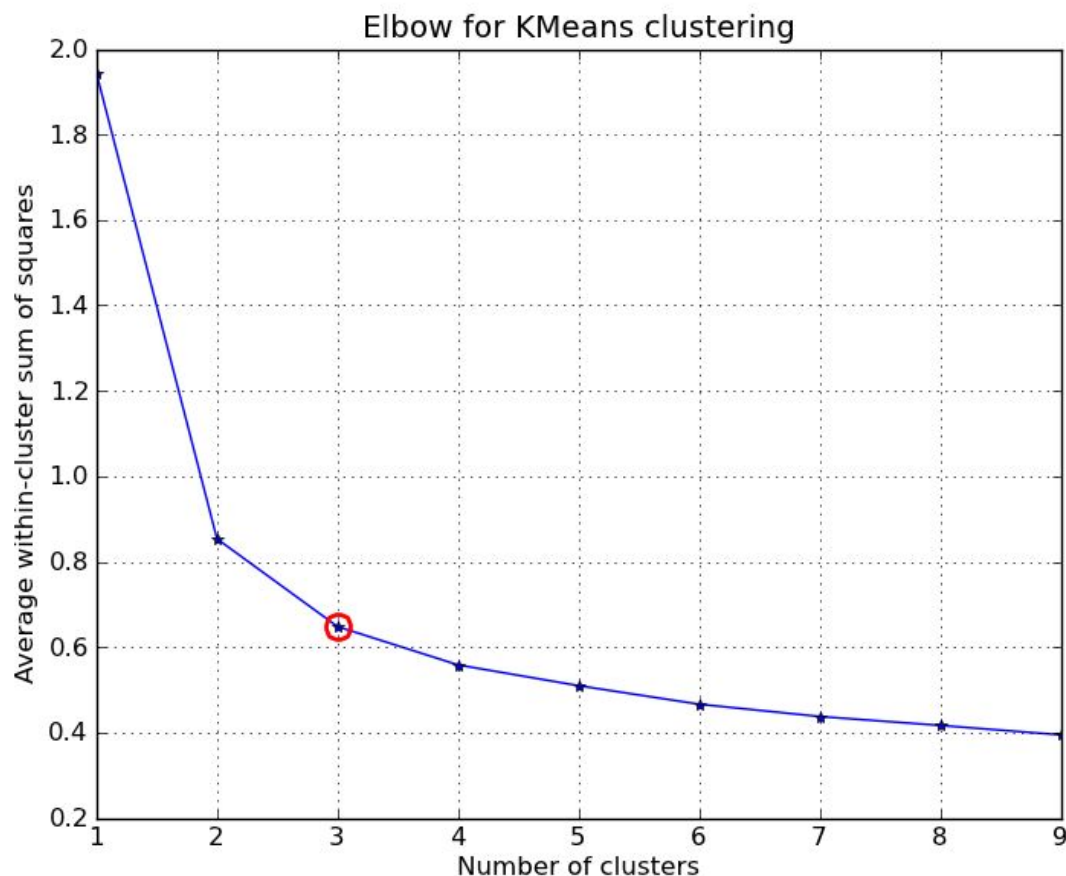


Re-calculate
Centroids

Choosing Number of Clusters (k)

- **Elbow method**

- Plot cluster evaluation metric (e.g., WSSE) vs. different values for k
- “Elbow” in plot suggests value(s) for k



<http://stackoverflow.com/questions/6645895/calculating-the-percentage-of-variance-measure-for-k-means>

Evaluating Clustering Results

- **Within-Cluster Sum of Squared Error (WSSE)**
- For each sample, error is distance to centroid.
Then, **WSSE** is computed as:

$$WSSE = \sum_{i=1}^K \sum_{x \in C_i} \|x - m_i\|^2$$

x : data sample in cluster C_i

m_i : cluster centroid (i.e., mean of cluster)

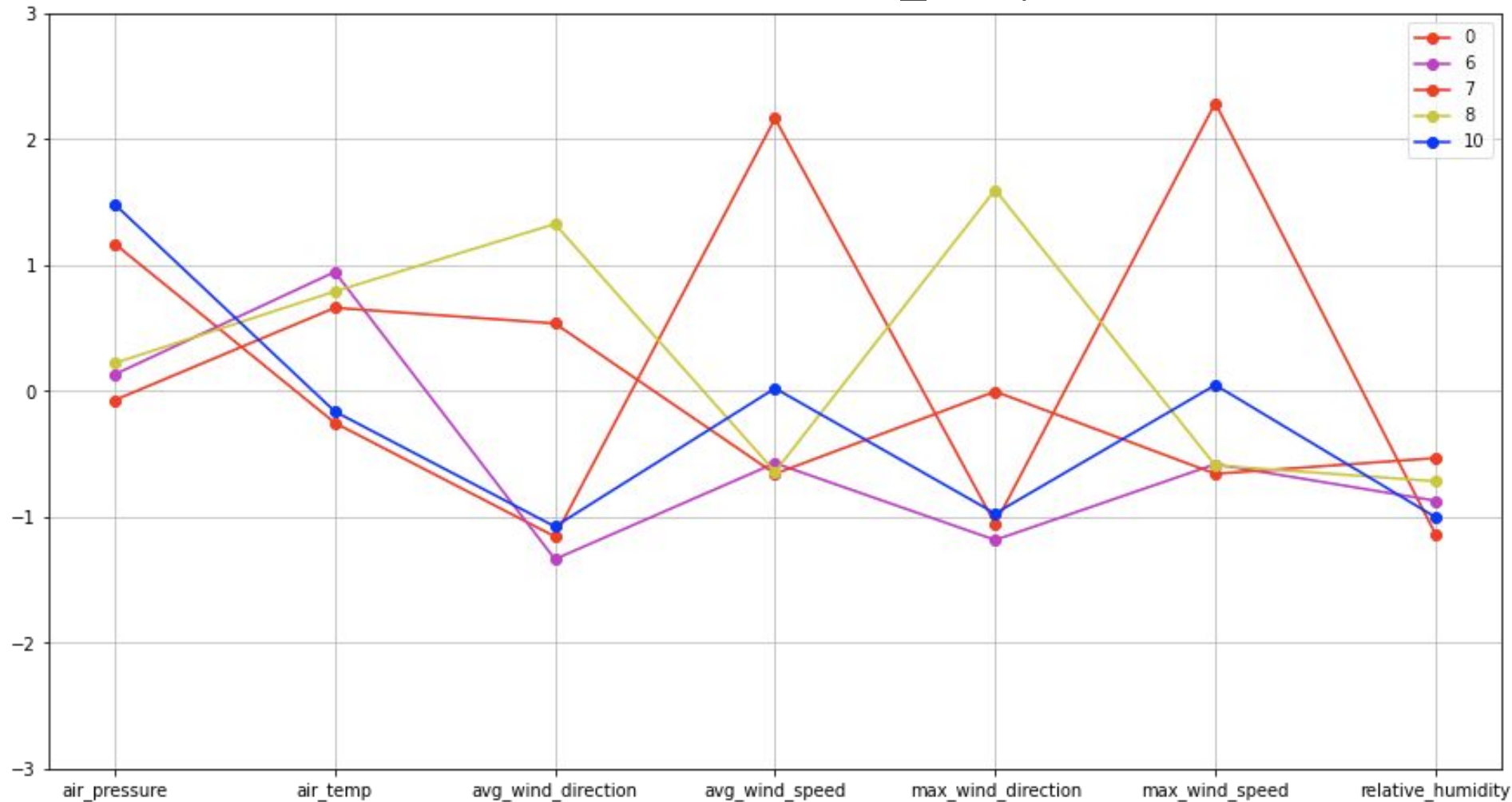
$\|x - m_i\|^2$: Euclidean distance between m_i and x

Clustering Hands-On Overview

- **Setup**
 - Start Spark
 - Load modules
- **Load data**
 - Specify schema
 - Read in data from “minute_weather.csv”
- **Explore data**
 - Look at schema, number of rows, summary statistics
- **Prepare data**
 - Drop nulls
 - Create feature vector
- **Perform k-means cluster analysis**
 - Use elbow plot to determine k
 - Build k-means model
- **Evaluate clusters**
 - Plot cluster profiles
- **Stop Spark session**

Cluster Profile: Parallel Plots

```
utils.parallel_plot(centersNamed[centersNamed['relative_humidity'] < -0.5],  
numClusters, colors=colors_used);
```



PySpark Cluster Analysis Hands-On

- **Code**

- **pyspark-clustering.ipynb**
 - Notebook for hands-on
 - Add your code where indicated
 - # ==> YOUR CODE HERE
- **pyspark-clustering-wOutput.ipynb**
 - Has cell outputs
- **utils.py**
 - Has utility functions

- **Resources**

- [Apache Spark](#)
- [PySpark Documentation — PySpark Documentation](#)
- [Spark SQL and DataFrames Guide](#)
- Python for Data Science Cheat Sheet (pdf)

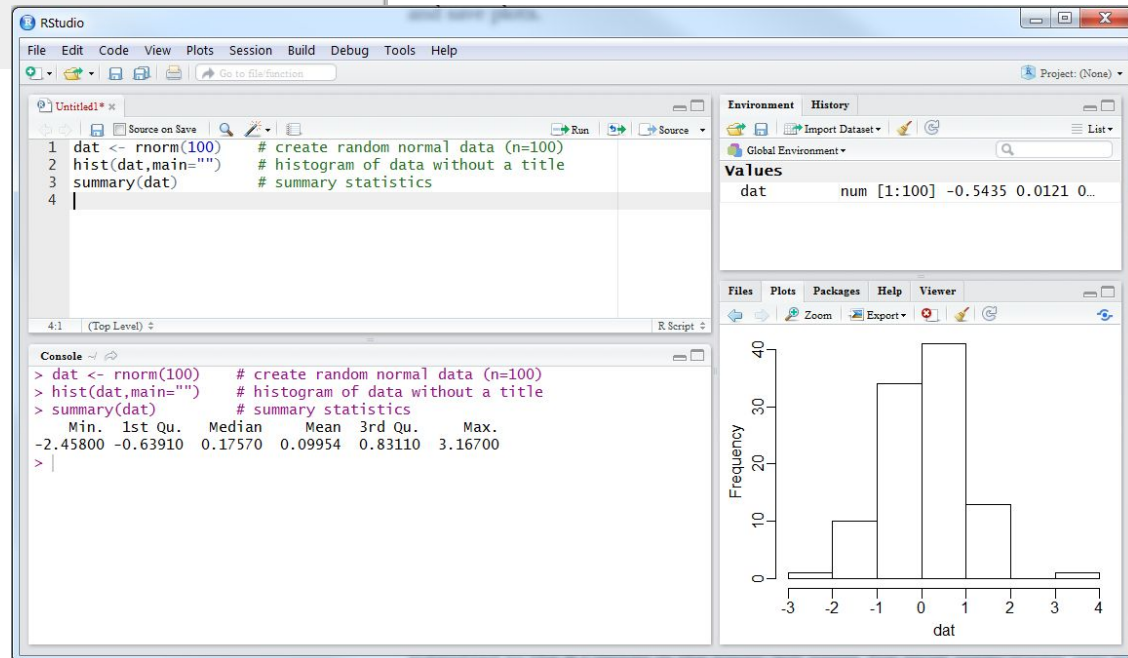
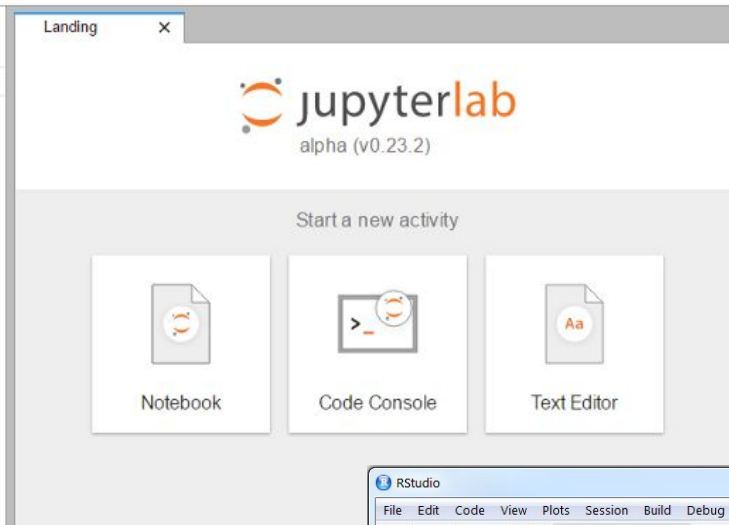
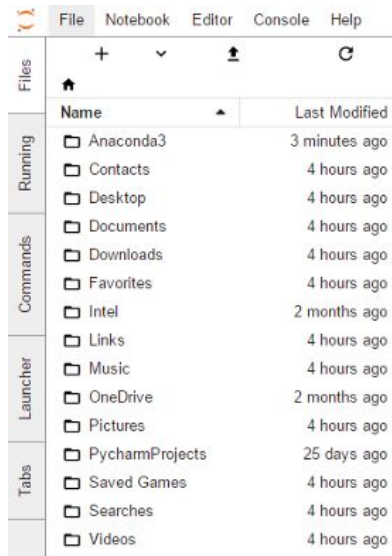
SparkR

- **R package that provides frontend to use Spark from R**
- **Supports distributed machine learning using R API**
- **Allows R script to connect to Spark cluster**
- **Can use with R shell or RStudio or other R IDEs.**

SparkR

- **Uses MLlib for machine learning functionality**
- **Familiar R syntax:**
 - Read contents of file into a Spark dataframe
 - `newdata <- read.df ("data.txt", source="csv")`
 - R formula operators for model fitting:
 - `model <- spark.randomForest(training, label ~ features, "classification", numTrees = 10)`
 - Get summary of fitted model
 - `summary(model)`
 - Apply model to make predictions
 - `predictions <- predict(model, testDF)`
 - Save model
 - `write.ml (model, "mymodel")`

Running SparkR Code





sparklyr



- **R interface for Spark**
- **R Legacy**
 - From Rstudio
 - Available on CRAN
- **Spark backend to dplyr and SQL**
 - Interactively manipulate Spark data using dplyr and SQL
- **Access to Spark functionality**
 - Interface to Spark MLlib algorithms
- **Connect to Spark clusters**

sparklyr

- **Setup**

- `install.packages("sparklyr")`
- `library(sparklyr)`
- `spark_install ()`

- **Connect to Spark**

- `sc <- spark_connect (master="local")`

- **Using dplyr**

- `flights_sdf %>% group_by(tailnum)`
`%>% filter(count > 20)`

- **Machine Learning**

- `model <- ml_random_forest (`
`train_sdf, quality ~ ., type="classification")`

sparklyr Functions

- **Spark Operations**
 - Manage Spark connections (e.g., `spark_config()`)
- **Spark Data Manipulation**
 - Read data in Spark DataFrame and perform operations (e.g., `spark_read_csv()`, `tbl_cache()`)
- **Feature Transformers**
 - Transform data (e.g., `ml_pca()`, `ft_bucketizer()`)
- **Distributed Machine Learning**
 - Access Spark Mllib algorithms (e.g., `ml_kmeans()`)
- **Streaming**
 - Support streaming data operations (e.g., `stream_read_json()`)
- **Extensions**
 - Interface to platforms for big data analysis, graph analytics, production

Machine Learning Algorithms in SparkR

Machine Learning

Algorithms

SparkR supports the following machine learning algorithms currently:

Classification

- `spark.logit`: Logistic Regression
- `spark.mlp`: Multilayer Perceptron (MLP)
- `spark.naiveBayes`: Naive Bayes
- `spark.svmLinear`: Linear Support Vector Machine

Regression

- `spark.survreg`: Accelerated Failure Time (AFT) Survival Model
- `spark.glm` or `glm`: Generalized Linear Model (GLM)
- `spark.isoreg`: Isotonic Regression

Tree

- `spark.gbt`: Gradient Boosted Trees for Regression and Classification
- `spark.randomForest`: Random Forest for Regression and Classification

Clustering

- `spark.bisectingKmeans`: Bisecting k-means
- `spark.gaussianMixture`: Gaussian Mixture Model (GMM)
- `spark.kmeans`: K-Means
- `spark.lda`: Latent Dirichlet Allocation (LDA)

Collaborative Filtering

- `spark.als`: Alternating Least Squares (ALS)

Frequent Pattern Mining

- `spark.fpGrowth`: FP-growth

Statistics

- `spark.kstest`: Kolmogorov-Smirnov Test

Resources

- **Spark**
 - <https://spark.apache.org/>
- **PySpark API**
 - <https://spark.apache.org/docs/latest/api/python/index.html>
- **Spark DataFrame**
 - <https://spark.apache.org/docs/latest/sql-programming-guide.html>
- **MLlib**
 - <https://spark.apache.org/mllib/>
- **SparkR**
 - <https://spark.apache.org/docs/latest/sparkr.html>
- **SparkR API**
 - <https://spark.apache.org/docs/latest/api/R/>
- **sparklyr**
 - <https://spark.rstudio.com/>

Questions?

