Cheat Sheet for PySpark

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```
Spark Configuration
from pyspark.sql import SparkSession
spark = SparkSession.builder
         .appName("Python Spark regression example")
         .config("config.option", "value").getOrCreate()
```

Loading Data

```
From RDDs
# Using parallelize( )
df = spark.sparkContext.parallelize([('1', 'Joe', '70000', '1'),
         ('2', 'Henry', '80000', None)])
.toDF(['Id', 'Name', 'Sallary', 'DepartmentId'])
# Using createDataFrame( )
['Id','Name','Sallary','DepartmentId'])
  | Id| Name|Sallary|DepartmentId|
  | 1| Joe| 70000|
  | 2|Henry| 80000|
                    nulll
```

From Data Sources

```
▶ From .csv
```

```
header=True,inferSchema=True)
| TV|Radio|Newspaper|Sales|
```

From .ison

```
df = spark.read.json('/home/feng/Desktop/data.json')
|2957256202|[72.1,DE,8086,52....|2019-02-23 22:36:52|
129572562031 [598.5.BG.3963.42... | 2019-02-23 22:36:521
```

From Database

```
user = 'username'; pw ='password'
table_name = 'table_name'
url='jdbc:postgresql://##.###.##:5432/dataset?user='
+user+'&password='+pw
p='driver':'org.postgresql.Driver','password':pw,'user':user
df = spark.read.jdbc(url=url,table=table_name,properties=p)
| TV|Radio|Newspaper|Sales
[230.1] 37.8] 69.2[ 22.1]
| 44.5| 39.3| 45.1| 10.4|
From HDFS
```

```
from pyspark.conf import SparkConf
from pyspark.context import SparkContext
from pyspark.sql import HiveContext
sc= SparkContext('local', 'example')
hc = HiveContext(sc)
tf1 = sc.textFile("hdfs://###/user/data/file_name")
| TV|Radio|Newspaper|Sales|
|230.1| 37.8| 69.2| 22.1|
|44.5| 39.3| 45.1| 10.4|
```

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Auditing Data

Checking schema df.printSchema() _c0: integer (nullable = true) |-- TV: double (nullable = true) |-- Radio: double (nullable = true) |-- Newspaper: double (nullable = true)

Checking missing value

|-- Sales: double (nullable = true)

```
from pyspark.sql.functions import count
def my_count(df):
    df.agg(*[count(c).alias(c) for c in df_in.columns]).show()
mv count(df raw)
 InvoiceNo|StockCode|Quantity|InvoiceDate|UnitPrice|CustomerID|Country|
   541909| 541909| 541909| 541909| 541909| 406829| 541909|
```

Checking statistical results

```
df raw.describe().show()
summarv
                             Radiol
                                          Newspaper
                2001
                             2001
 count
             147.0425|23.264000000000024|30.55399999999995|
  mean
 stddev|85.85423631490805|14.846809176168728| 21.77862083852283|
  min
                 0.7|
   max
```

Manipulating Data (More details on next page)

Fixing missing value Function Description df.na.fill() #Replace null values df.na.drop() #Dropping any rows with null values.

Joining data

Description	Function
#Data join	<pre>left.join(right,key, how='*') * = left,right,inner,full</pre>

Wrangling with UDF

```
from pyspark.sql import functions as F
from pyspark.sql.types import DoubleType
# user defined function
def complexFun(x):
   return results
Fn = F.udf(lambda x: complexFun(x), DoubleType())
df.withColumn('2col', Fn(df.col))
```

Reducing features

```
df.select(featureNameList)
```

Modeling Pipeline

Deal with categorical feature and label data # Deal with categorical feature data from pyspark.ml.feature import VectorIndexer featureIndexer = VectorIndexer(inputCol="features", outputCol="indexedFeatures", maxCategories=4).fit(data) featureIndexer.transform(data).show(2. True) features|label| indexedFeatures| |(29,[1,11,14,16,1...| no|(29,[1,11,14,16,1...| # Deal with categorical label data labelIndexer=StringIndexer(inputCol='label', outputCol='indexedLabel').fit(data) labelIndexer.transform(data).show(2, True) features[label[indexedLabel] (29. [1.11.14.16.1...] nol 0.01

Spliting the data to training and test data sets

```
(trainingData, testData) = data.randomSplit([0.6, 0.4])
```

Importing the model

```
from pyspark.ml.classification import LogisticRegression
lr = LogisticRegression(featuresCol='indexedFeatures',
                       labelCol='indexedLabel')
```

Converting indexed labels back to original labels

```
from pyspark.ml.feature import IndexToString
labels=labelIndexer.labels)
```

Wrapping Pipeline

```
pipeline = Pipeline(stages=[labelIndexer, featureIndexer,
                            lr,labelConverter])
```

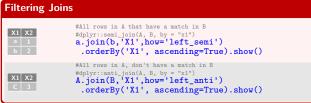
Training model and making predictions

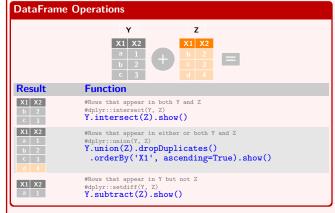
```
model = pipeline.fit(trainingData)
predictions = model.transform(testData)
predictions.select("features","label","predictedLabel").show(2)
         features|label|predictedLabel|
|(29,[0,11,13,16,1...| no|
```

Evaluating

```
from pyspark.ml.evaluation import *
evaluator = MulticlassClassificationEvaluator(
              labelCol="indexedLabel",
              predictionCol="prediction", metricName="accuracy")
accu = evaluator.evaluate(predictions)
print("Test Error: %g, AUC: %g"%(1-accu,Summary.areaUnderROC))
Test Error: 0.0986395, AUC: 0.886664269877
```





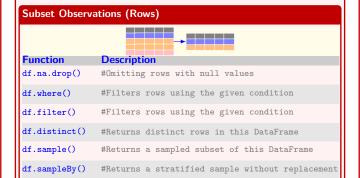




Data Wrangling: Reshaping Data **Spliting** Change **Function** #ArrayType() tidyr::separate ::separate one column into several df.select("key", df.value[0], df.value[1], df.value[2]).show() #StructType() df2.select('key', 'value.*').show() #Splitting one column into rows df.select("key",F.split("values", ",").alias("values"), key value F.posexplode(F.split("values",",")).alias("pos", "val")),drop("val") .select("key",F.expr("values[pos]").alias("val")).show() #Gather columns into rows def to long(df. by): key value cols, dtypes = zip(*((c,t) for (c, t) in df.dtypes if c not in by)) # Spark SQL supports only homogeneous columns assert len(set(dtypes))==1,"All columns have to be of the same type a 3 # Create and explode an array of (column_name, column_value) struct: kvs = explode(array([struct(lit(c).alias("key"), col(c).alias("val")) for c in cols 6])).alias("kvs") return df.select(by + [kvs]).select(by + ["kvs.key", "kvs.val"])

Pivot

kay coll



#Spread rows into columns

.pivot('col1').sum('col1').show()

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