

# Module 4: Big Data Analytics in Apache Spark

# Big Data Analytics with Spark

- Spark Dataframes to work with tabular data
- Data cleaning, summary, statistics
- Spark Dataframes with SQL and Hive

# Open PySpark

```
PYSPARK_DRIVER_PYTHON=ipython pyspark
```

# Introduction to Spark Dataframes

# Types of RDD: text

from local filesystem:

```
text_RDD =
```

```
sc.textFile("file:///home/cloudera/testfile1")
```

```
text_RDD.collect()
```

```
Out[]: [u'A long time ago in a galaxy far far away']
```

# Types of RDD: key-value pairs

```
def split_words(line):  
    return line.split()
```

```
def create_pair(word):  
    return (word, 1)
```

```
pairs_RDD=text_RDD.flatMap(split_words  
) .map(create_pair)
```

```
pairs_RDD.collect()
```

```
Out[]: [(u'A', 1),  
        (u'long', 1),  
        (u'time', 1),  
        (u'ago', 1),  
        (u'in', 1),  
        (u'a', 1),  
        (u'galaxy', 1),  
        (u'far', 1),  
        (u'far', 1),  
        (u'away', 1)]
```

# Tabular dataset

Most real-world datasets have  
records (rows)

each with  
multiple values (columns)



# Tweets

user	text	datetime	favorites	retweets
andreazonca	"spark is cool"	"2015-10-1 9:04"	5	3

# Reviews

business	text	datetime	starts	user
Pan Bon	"great pizza!"	"2015-10-1 9:04"	5	andreasonca

# Logs

http_code	ip	datetime	user_agent
200	127.0.0.1	"2015-10-1 9:04"	Firefox

# Tabular datasets

```
students = sc.parallelize([  
    [100, "Alice", 8.5, "Computer Science"],  
    [101, "Bob", 7.1, "Engineering"],  
    [102, "Carl", 6.2, "Engineering"]  
])
```

# Mean of a column

```
def extract_grade(row):  
    return row[2]
```

```
students.map(extract_grade).mean()
```

```
Out[]: 17.26666
```

# Group by column

```
def extract_degree_grade(row):
```

```
    return (row[3], row[2])
```

```
degree_grade_RDD =
```

```
students.map(extract_degree_grade)
```

```
degree_grade_RDD.collect()
```

# Group by column

Intermediate RDD:

```
degree_grade_RDD.collect()
```

Out[ ]:

```
[('Computer Science', 8.5),
```

```
('Engineering', 7.0999999999999996),
```

```
('Engineering', 6.2000000000000002)]
```

# Group by column

## Reduce by key to get the final result:

```
degree_grade_RDD.reduceByKey(max).collect()
```

# Out[]:

```
[('Engineering', 7.0999999999999996),
```

```
('Computer Science', 8.5)]
```



# Introducing Spark Dataframes

User friendly interface

Under-the-hood optimization  
for table-like datasets

```
students_df = sqlCtx.createDataFrame(students,  
    ["id", "name", "grade", "degree"])
```

```
students_df.printSchema()
```

root

```
|-- id: long (nullable = true)
```

```
|-- name: string (nullable = true)
```

```
|-- grade: double (nullable = true)
```

```
|-- degree: string (nullable = true)
```

# sqlCtx.createDataFrame?

Create a DataFrame from an RDD of tuple/list, list or pandas.DataFrame.

`schema` could be :class:`StructType` or a list of column names.

When `schema` is a list of column names, the type of each column will be inferred from `rdd`.

When `schema` is None, it will try to infer the column name and type from `rdd`, which should be an RDD of :class:`Row`, or namedtuple, or dict.

If referring needed, `samplingRatio` is used to determined how many rows will be used to do referring. The first row will be used if `samplingRatio` is None.

:param data: an RDD of Row/tuple/list/dict, list, or pandas.DataFrame

:param schema: a StructType or list of names of columns

:param samplingRatio: the sample ratio of rows used for inferring

:return: a DataFrame

```
>>> l = [('Alice', 1)]
```

```
>>> sqlCtx.createDataFrame(l).collect()
```

```
[Row(_1=u'Alice', _2=1)]
```

```
>>> sqlCtx.createDataFrame(l, ['name', 'age']).collect()
```

```
[Row(name=u'Alice', age=1)]
```

# Mean of a column

```
students_df.agg({"grade": "mean"}).collect()
```

```
Out[]: [Row(AVG(grade#30)=7.2666666666666666)]
```

Find all available operations:

# Group by column

```
students_df.groupBy("degree").max("grade").collect()
```

Out[]:

```
[Row(degree=u'Computer Science',  
MAX(grade#30)=8.5),  
Row(degree=u'Engineering',  
MAX(grade#30)=7.0999999999999)]
```

# Pretty print with show

```
students_df.groupby("degree").max("grade").show()
```

degree	MAX(grade#30)
--------	---------------

Computer Science	8.5
------------------	-----

Engineering	7.1
-------------	-----

# Final remarks on Dataframes

- special kind of RDD
- transformations/actions/DAG work the same way
- automatic optimization to Java bytecode
- Python as fast as Scala/Java

# Create Spark Dataframes



# Specify a Schema

In the last video:

```
students_df = sqlCtx.createDataFrame(students,  
    ["id", "name", "grade", "degree"]
```

```
from pyspark.sql.types import *
```

```
schema = StructType([
```

```
    StructField("id", LongType(), True),
```

```
    StructField("name", StringType(), True),
```

```
    StructField("grade", DoubleType(), True),
```

```
    StructField("degree", StringType(), True) ])
```

```
students_df = sqlCtx.createDataFrame(students, schema)
```

```
students_df.printSchema()
```

```
root
```

```
|-- id: long (nullable = true)
```

```
|-- name: string (nullable = true)
```

```
|-- grade: double (nullable = true)
```

```
|-- degree: string (nullable = true)
```

# Load a JSON file

```
students_json = [  
    '{"id":100, "name":"Alice", "grade":8.5,  
    "degree":"Computer Science"}',  
    '{"id":101, "name":"Bob", "grade":7.1,  
    "degree":"Engineering"}']  
  
with open("students.json", "w") as f:  
    f.write("\n".join(students_json))
```

# Dump JSON file content

```
!cat students.json
```

```
{ "id":100, "name":"Alice", "grade":8.5, "degree":"Computer  
Science" }
```

```
{ "id":101, "name":"Bob", "grade":7.1,  
"degree":"Engineering" }
```

# Create Dataframe with jsonFile

```
sqlCtx.jsonFile("file:///home/cloudera/students.json").show()
```

degree	grade	id	name
--------	-------	----	------

Computer Science	8.5	100	Alice
------------------	-----	-----	-------

Engineering	7.1	101	Bob
-------------	-----	-----	-----

# Load Dataframe from CSV

- Not included in Spark
- Load from [spark-packages.org](http://spark-packages.org)



A community index of packages for **Apache Spark**.

140 packages

[All \(140\)](#) [Core \(5\)](#) [Data Sources \(22\)](#) [Machine Learning \(35\)](#) [Streaming \(21\)](#) [Graph \(5\)](#) [PySpark \(2\)](#) [Applications \(5\)](#) [Deployment \(8\)](#) [Examples \(8\)](#) [Tools \(13\)](#)

## spark-avro

Integration utilities for using Spark with Apache Avro data

from: [@databricks](#) / owner: [@pwendell](#) / Latest release: 2.0.1-s\_2.10 (2015-09-08) / Apache-2.0 / ★★★★★ (18)

4 sql 3 input 3 avro

## spark-redshift

Spark and Redshift integration

from: [@databricks](#) / owner: [@pwendell](#) / Latest release: 0.5.2 (2015-10-23) / Apache-2.0 / ★★★★★ (13)

1 input 1 sql 1 redshift

## kafka-spark-consumer

Receiver Based Low Level Kafka-Spark Consumer with builtin Back-Pressure Controller

[@dibbhatt](#) / Latest release: 1.0.5 (2015-10-08) / Apache-2.0 / ★★★★★ (15)

3 streaming 2 kafka

## thunder

Large-scale neural data analysis with Spark

[@freeman-lab](#) / Latest release: 0.4.1 (2014-11-27) / Apache-2.0 / ★★★★★ (14)





## spark-csv [\(homepage\)](#)

Spark SQL CSV data source

from: [@databricks](#) / owner: [@falaki](#) / ★★★★★ (17)

This packages adds a new CSV data source to Spark SQL. CSV files can be read as Schema RDD and registered as table. Supports quotes and headers.

### Tags

1 sql

1 SparkSQL

1 DataSource

1 csv

### How to [ + ]

Include this package in your Spark Applications using:

spark-shell, pyspark, or spark-submit

```
> $SPARK_HOME/bin/spark-shell --packages com.databricks:spark-csv_2.11:1.2.0
```

### Releases

**Version: 1.2.0-s\_2.11** ( [82344b](#) | [zip](#) | [jar](#) ) / Date: 2015-08-07 / License: [Apache-2.0](#) / Scala version: 2.11

Spark Scala/Java API compatibility: 1.2.0 - [43%](#), 1.3.0 - [77%](#), 1.4.0 - [100%](#)

**Version: 1.2.0-s\_2.10** ( [82344b](#) | [zip](#) | [jar](#) ) / Date: 2015-08-07 / License: [Apache-2.0](#) / Scala version: 2.10

Spark Scala/Java API compatibility: 1.0.0 - [11%](#), 1.1.0 - [38%](#), 1.2.0 - [43%](#), 1.3.0 - [77%](#), 1.4.0 - [100%](#)

**Version: 1.0.3** ( [464a3e](#) | [zip](#) | [jar](#) ) / Date: 2015-04-04 / License: [Apache-2.0](#) / Scala version: 2.11

Spark Scala/Java API compatibility: 1.2.0 - [43%](#), 1.3.0 - [100%](#)

# Restart PySpark

```
PYSPARK_DRIVER_PYTHON=ipython pyspark --  
packages com.databricks:spark-csv_2.10:1.2.0
```

Automatically download and include new  
packages and dependencies

# Load sample yelp csv

```
yelp_df = sqlCtx.load(  
source="com.databricks.spark.csv",  
header = 'true',  
inferSchema = 'true',  
path =  
'file:///usr/lib/hue/apps/search/examples/collections/solr_co  
nfigs_yelp_demo/index_data.csv')
```

```
yelp_df.printSchema()
```

```
root
```

```
|-- business_id: string (nullable = true)
```

```
|-- cool: integer (nullable = true)
```

```
|-- date: string (nullable = true)
```

```
|-- funny: integer (nullable = true)
```

```
|-- id: string (nullable = true)
```

```
|-- stars: integer (nullable = true)
```

```
|-- text: string (nullable = true)
```

```
|-- type: string (nullable = true)
```

```
|-- useful: integer (nullable = true)
```

```
|-- user_id: string (nullable = true)
```

```
|-- name: string (nullable = true)
```

```
|-- full_address: string (nullable = true)
```

```
|-- latitude: double (nullable = true)
```

```
|-- longitude: double (nullable = true)
```

```
|-- neighborhoods: string (nullable = true)
```

```
|-- open: string (nullable = true)
```

```
|-- review_count: integer (nullable = true)
```

```
|-- state: string (nullable = true)
```

```
yelp_df.count()
```

```
Out[]: 1000L
```

# Analytics with Dataframes on a Yelp reviews dataset

# Explore the Yelp dataset

```
yelp_df = sqlCtx.load(  
    source='com.databricks.spark.csv',  
    header = 'true',  
    inferSchema = 'true',  
    path =  
    'file:///usr/lib/hue/apps/search/examples/collections/solr_co  
nfigs_yelp_demo/index_data.csv')
```

# Reference a column

As attribute:

```
yelp_df.useful
```

```
Out[]: Column<useful>
```

As key:

```
yelp_df["useful"]
```

```
Out[]: Column<useful>
```

# Filtering

```
yelp_df.filter(yelp_df.useful >= 1).count()
```

```
yelp_df.filter(yelp_df["useful"] >= 1).count()
```

```
yelp_df.filter("useful >= 1").count()
```

```
Out[]: 601L
```



# select

```
yelp_df["useful"].agg({"useful": "max"}).collect()
```

```
Out[]: AttributeError: 'Column' object has no attribute 'agg'
```

```
yelp_df.select("useful")
```

```
Out[]: DataFrame[useful: int]
```

```
yelp_df.select("useful").agg({"useful": "max"}).collect()
```

```
Out[]: [Row(MAX(useful#267)=28)]
```

# Create a modified DataFrame

Rescale the useful column from 0-28 to 0-100.

# Create a 2 columns DataFrame

```
yelp_df.select("id", "useful").take(5)
```

```
[Row(id=u'fWKvX83p0-ka4JS3dc6E5A', useful=5),  
 Row(id=u'ljZ33sJrzXqU-0X6U8NwyA', useful=0),  
 Row(id=u'IESLBzqUCLdSzSqm0eCSxQ', useful=1),  
 Row(id=u'G-WvGalSbqqaMHINnByodA', useful=2),  
 Row(id=u'1uJFq2r5QfJG_6ExMRCaGw', useful=0)]
```

# Modify column

```
yelp_df.select("id", yelp_df.useful/28*100).show(5)
```

id	((useful / 28) * 100)
fWKvX83p0-ka4JS3d...	17.857142857142858
IjZ33sJrzXqU-0X6U...	0.0
IESLBzqUCLdSzSqm0...	3.571428571428571
G-WvGaISbqqaMHINn...	7.142857142857142
1uJFq2r5QfJG_6ExM...	0.0

# Cast (truncate) to integer

```
yelp_df.select("id",  
(yelp_df.useful/28*100).cast("int")).show(5)
```

```
id                CAST(((useful / 28) * 100)), IntegerType)
```

```
fWKvX83p0-ka4JS3d... 17
```

```
IjZ33sJrzXqU-0X6U... 0
```

```
IESLBzqUCLdSzSqm0... 3
```

```
G-WvGaISbqqaMHINn... 7
```

```
1uJFq2r5QfJG_6ExM... 0
```

# Save as new dataframe

```
useful_perc_data = yelp_df.select(  
    "id",  
    (yelp_df.useful/28*100).cast("int")  
)
```

```
useful_perc_data.columns
```

```
Out[]: [u'id', u'CAST(((useful / 28) * 100), IntegerType)']
```

# alias - rename a column

```
useful_perc_data = yelp_df.select(  
    "id",  
    (yelp_df.useful/28*100).cast("int").alias("useful_perc")  
)
```

```
useful_perc_data.columns
```

```
Out[: [u'id', u'useful_perc']
```

# alias - rename a column

```
useful_perc_data = yelp_df.select(  
    "id",  
    (yelp_df.useful/28*100).cast("int").alias("useful_perc")  
)
```

```
useful_perc_data.columns
```

```
Out[: [u'id', u'useful_perc']
```



# alias - rename also id

```
useful_perc_data = yelp_df.select(  
    yelp_df["id"].alias("uid"),  
    (yelp_df.useful/28*100).cast("int").alias("useful_perc")  
)
```

```
useful_perc_data.columns
```

```
Out[: [u'uid', u'useful_perc']
```

# Ordering by column

Import functions for ascending/descending order:

```
from pyspark.sql.functions import asc, desc
```

# order by usefulness

```
useful_perc_data = yelp_df.select(  
    yelp_df["id"].alias("uid"),  
    (yelp_df.useful/28*100).cast("int").alias("useful_perc")  
).orderBy(desc("useful_perc"))
```

```
useful_perc_data.show(2)
```

uid	useful_perc
-----	-------------

RqwFPp_qPu-1h87pG...	100
----------------------	-----

YAXPKM-Hck6-mjF74...	82
----------------------	----

# Join inputs

id	useful_perc
9yKzy9PApe	17

id	review_count	state
9yKzy9PApe	6	"CA"

# Join results

id	useful_perc	review_count
9yKzy9PApe	17	6

# Join

```
useful_perc_data.join(
```

```
yelp_df,
```

```
yelp_df.id == useful_perc_data.uid,
```

```
"inner"
```

```
)
```

# Join - select

```
useful_perc_data.join(  
    yelp_df,  
    yelp_df.id == useful_perc_data.uid,  
    "inner"  
).select(useful_perc_data.uid, "useful_perc", "review_count")
```

# Join - select - show


```
useful_perc_data.join(  
    yelp_df,  
    yelp_df.id == useful_perc_data.uid,  
    "inner"  
).select(useful_perc_data.uid, "useful_perc",  
"review_count").show(5)
```



# Output dataset

uid	useful_perc	review_count
WRBYytJAaJI1BTQG5...	71	362
GXj4PNAi095-q9ynP...	3	76
1sn0-eY_d1Dhr6Q2u...	0	9
MtFe-FuiOmo0vlo16...	0	7
EMYmuTlyeNBy5QB9P...	7	19

# Cache in memory

```
useful_perc_data.join(  
    yelp_df,  
    yelp_df.id == useful_perc_data.uid,  
    "inner"  
).cache().select(useful_perc_data.uid, "useful_perc",  
    "review_count").show(5)
```

Run it again!

# Analytics with Dataframes on HTTP server logs

# Log analytics

Available in the Cloudera VM at:

```
/usr/lib/hue/apps/search/examples/collections/solr_configs_log_analytics_demo/index_data.csv
```

```
|
```

# Log analytics

Check file contents on the terminal:

```
head
```

```
/usr/lib/hue/apps/search/examples/collecti  
ons/solr_configs_log_analytics_demo/index  
_data.csv
```

```
|
```

# Columns

code,protocol,request,app,user\_agent\_major,region\_code,country\_code,id,city,subapp,latitude,method,client\_ip,user\_agent\_family,bytes,referrer,country\_name,extension,url,os\_major,longitude,device\_family,record,user\_agent,time,os\_family,country\_code3

# Start PySpark

Need to load spark-csv for CSV support:

```
PYSPARK_DRIVER_PYTHON=ipython pyspark --  
packages com.databricks:spark-csv_2.10:1.X.X
```

# (Try to) read logs CSV

```
logs_df = sqlCtx.load(  
    source="com.databricks.spark.csv",  
    header = 'true',  
    inferSchema = 'true',  
    path =  
    'file:///usr/lib/hue/apps/search/examples/collections/solr_co  
nfigs_log_analytics_demo/index_data.csv')  
  
logs_df.count()
```



# Parsing error

ERROR csv.CsvRelation\$: Exception while parsing  
line: ",Mozilla/4.0 (compatible; MSIE 7.0;  
Windows NT 5.1; Trident/4.0; ....

# Inspect the file with VIM

```
1 code,protocol,request,app,user_agent_major,region_code,country_code,id,city,subapp,latitude,method,client_ip,user_a
  record,user_agent,time,os_family,country_code3^M
2 200,HTTP/1.1,GET /metastore/table/default/sample_07 HTTP/1.1,metastore,,00,SG,8836e6ce-9a21-449f-a372-9e57641389b3,
  ore/table/default/sample_07,,103.85579999999999,0ther,"demo.gethue.com:80 128.199.234.236 - - [04/May/2014:06:35:49
  .0 (compatible; phpservermon/3.0.1; +http://www.phpservermonitor.org)""
3 ",Mozilla/5.0 (compatible; phpservermon/3.0.1; +http://www.phpservermonitor.org),2014-05-04T06:35:49Z,0ther,SGP^M
4 200,HTTP/1.1,GET /metastore/table/default/sample_07 HTTP/1.1,metastore,,00,SG,6ddf6e38-7b83-423c-8873-39842dca2dbb,
  ore/table/default/sample_07,,103.85579999999999,0ther,"demo.gethue.com:80 128.199.234.236 - - [04/May/2014:06:35:50
  .0 (compatible; phpservermon/3.0.1; +http://www.phpservermonitor.org)""
5 Mozilla/5.0 (compatible; phpservermon/3.0.1; +http://www.phpservermonitor.org),2014-05-04T06:35:50Z,0ther,SGP^M
6 200,HTTP/1.1,GET /search/?collection=10000001 HTTP/1.1,search,,00,SG,313bb28e-dd7c-4364-a11e-9ffb0db7b303,Singapore
```

# Access Hadoop configuration

Spark relies on Hadoop functionality for reading data.

```
sc._jsc.hadoopConfiguration()
```

# Set input file delimiter

Spark relies on Hadoop functionality for reading data.

```
sc._jsc.hadoopConfiguration().set('textinputforma  
t.record.delimiter', '\r\n')
```

```
|
```

# Read logs CSV

```
logs_df = sqlCtx.load(  
    source="com.databricks.spark.csv",  
    header = 'true', inferSchema = 'true',  
    path =  
    'file:///usr/lib/hue/apps/search/examples/collections/solr_co  
nfigs_log_analytics_demo/index_data.csv')
```

```
logs_df.count()
```

```
Out[]: 9410L
```

# Display of logs DataFrame

```
code protocol request          app      user_agent_major region_code country_code id
e extension url              os_major longitude      device_family record      user_agent
200  HTTP/1.1 GET /metastore/ta... metastore null          00          SG          8836e6ce-9a21-449...
      /metastore/table/... null          103.85579999999999 Other          demo.gethue.com:8... Mozilla/5.0
200  HTTP/1.1 GET /metastore/ta... metastore null          00          SG          6ddf6e38-7b83-423...
      /metastore/table/... null          103.85579999999999 Other          demo.gethue.com:8... Mozilla/5.0
200  HTTP/1.1 GET /search/?coll... search      null          00          SG          313bb28e-dd7c-436...
      /search/?collecti... null          103.85579999999999 Other          demo.gethue.com:8... Mozilla/5.0
200  HTTP/1.1 GET /search/?coll... search      null          00          SG          ecb47c61-a9e4-4b5...
      /search/?collecti... null          103.85579999999999 Other          demo.gethue.com:8... Mozilla/5.0
200  HTTP/1.1 HEAD / HTTP/1.1      null          00          SG          affdb6b9-3657-4d1...
      /              null          103.85579999999999 Other          demo.gethue.com:8... Mozilla/5.0
```

root

```
|-- code: integer (nullable = true)
|-- protocol: string (nullable = true)
|-- request: string (nullable = true)
|-- app: string (nullable = true)
|-- user_agent_major: integer (nullable = true)
|-- region_code: string (nullable = true)
|-- country_code: string (nullable = true)
|-- id: string (nullable = true)
|-- city: string (nullable = true)
|-- subapp: string (nullable = true)
|-- latitude: double (nullable = true)
|-- method: string (nullable = true)
|-- client_ip: string (nullable = true)
|-- user_agent_family: string (nullable = true)
|-- bytes: integer (nullable = true)
|-- referer: string (nullable = true)
|-- country_name: string (nullable = true)
|-- extension: string (nullable = true)
```

# Count by HTTP code

Count the log events by HTTP code (i.e. how many 200 OK, 404 Not found...)



```
logs_df.groupby("code").count().show()
```

```
code count
```

```
500 2
```

```
301 71
```

```
302 1943
```

```
502 6
```

```
304 117
```

```
400 1
```

```
200 7235
```

```
401 10
```

```
404 11
```

```
from pyspark.sql.functions import asc, desc
```

```
logs_df.groupBy("code").count().orderBy(desc("count")).show()
```

```
|
```

```
code count
```

```
200 7235
```

```
302 1943
```

```
304 117
```

```
301 71
```

```
408 14
```

```
404 11
```

# Compute average

```
logs_df.groupby("code").avg("bytes").show()
```

```
|
```

```
code AVG(bytes#47)
```

```
500 4684.5
```

```
301 424.61971830985914
```

```
302 415.6510550694802
```

```
502 581.0
```

```
304 185.26495726495727
```

```
400 0.0
```

# Mean, Min, Max by code

Compute in a single operation Mean, Min and Max by HTTP code

```
|  
import pyspark.sql.functions as F  
|
```

```
logs_df.groupBy("code").agg(  
    logs_df.code,  
    F.avg(logs_df.bytes),  
    F.min(logs_df.bytes),  
    F.max(logs_df.bytes)  
).show()
```

# Mean, Min, Max by code

code	AVG(bytes#47)	MIN(bytes#47)	MAX(bytes#47)
500	4684.5	422	8947
301	424.61971830985914	331	499
302	415.6510550694802	304	1034
502	581.0	581	581
304	185.26495726495727	157	204
400	0.0	0	0
200	41750.03759502419	0	9045352
401	12472.8	8318	28895
404	17872.454545454544	7197	23822
408	440.57142857142856	0	514

# Completed DataFrames

- Completed analytics with DataFrames
- Next we'll focus on interoperability with SQL query language and Hive