

Apache Spark

Current Topics in Distributed Systems: Internet of Things and Cloud Computing

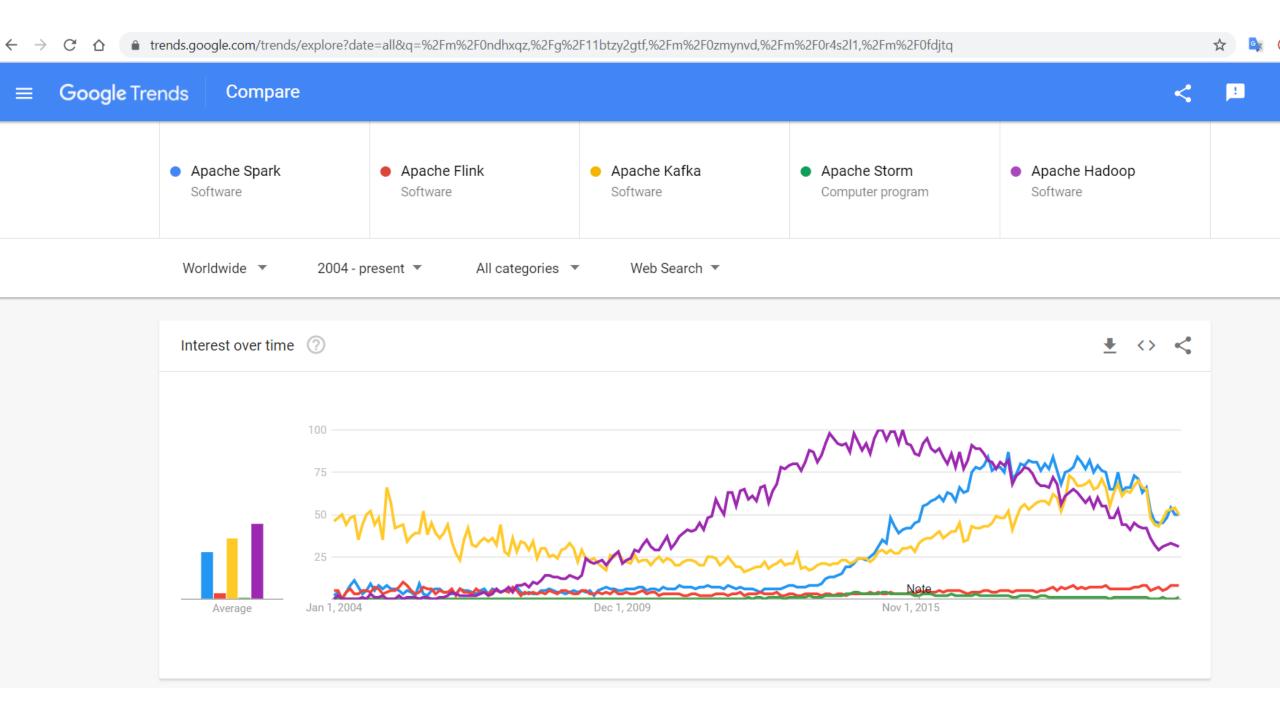
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Programming models in a Cloud environment

The growth of data volumes in industry and research poses tremendous opportunities, as well as tremendous computational challenges.

There are several programming models which focus on supporting batch or real-time data processing efficiently in a Cloud environment. For instance,

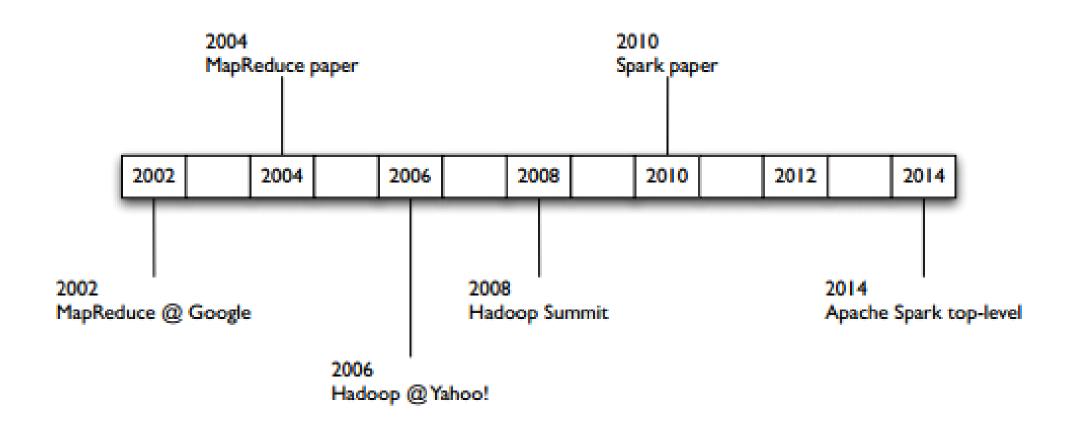
- MapReduce has become the de facto standard for batch data processing in the Apache Hadoop framework.
- Apache Spark is a distributed batch processing framework, but it also supports stream processing based on micro-batching.
- Apache Storm supports event-based stream processing.
- Apache Flink enables batch and stream processing with its unified APIs.



History

- Google invented a new methodology of processing data known as MapReduce.
- Doug Cutting and Mike Cafarella co-founded Hadoop to apply MapReduce concepts to an open-source software framework that supported the Nutch search engine project.
- Spark can process real-time data, such as real-time event streaming with a rate of millions of events/second, e.g., the live data streaming of Twitter, Facebook, Instagram.
 - But MapReduce fails as it cannot handle real-time data processing.
 - It can just perform batch processing on huge volumes of data.

Apache Platforms History



Introduction

- Apache Spark developed in 2009 at UC Berkeley AMPLab, then opensourced in 2010, to be a general-purpose distributed computing system for big data analytics.
- Spark can complete jobs substantially faster than previous big data tools (i.e., Apache Hadoop) because of its in-memory caching, and optimized query execution.
- Spark provides development APIs in Python, Java, Scala, and R.
- On top of the main computing framework, Spark provides machine learning, SQL, graph analysis, and streaming libraries.
- Dealing with massive amounts of data often requires parallelization and cluster computing;
 - Apache Spark is an industry standard for doing this.

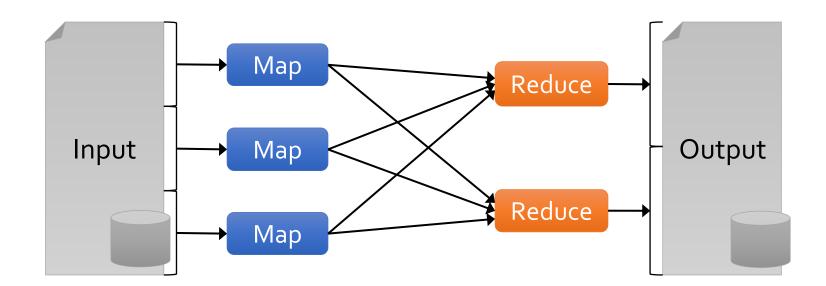
Goals

- Extend the MapReduce model to better support two common classes of analytics apps:
 - Iterative algorithms (machine learning, graphs)
 - Interactive data mining

- Enhance programmability:
 - Integrate into Scala programming language
 - Allow interactive use from Scala interpreter

Motivation (cont.)

• Most current cluster programming models are based on *acyclic data flow* from one stable storage to another stable storage



Motivation (cont.)

- Acyclic data flow is inefficient for applications that repeatedly reuse a working set of data:
 - Iterative algorithms (machine learning, graphs)
 - Interactive data mining tools (R, Excel, Python)
- With current frameworks, apps reload data from stable storage on each query

Solution: Resilient Distributed Datasets (RDDs)

Allow apps to keep working-sets in memory for efficient reuse

- Retain the attractive properties of MapReduce
 - Fault tolerance, data locality, scalability

Support a wide range of applications

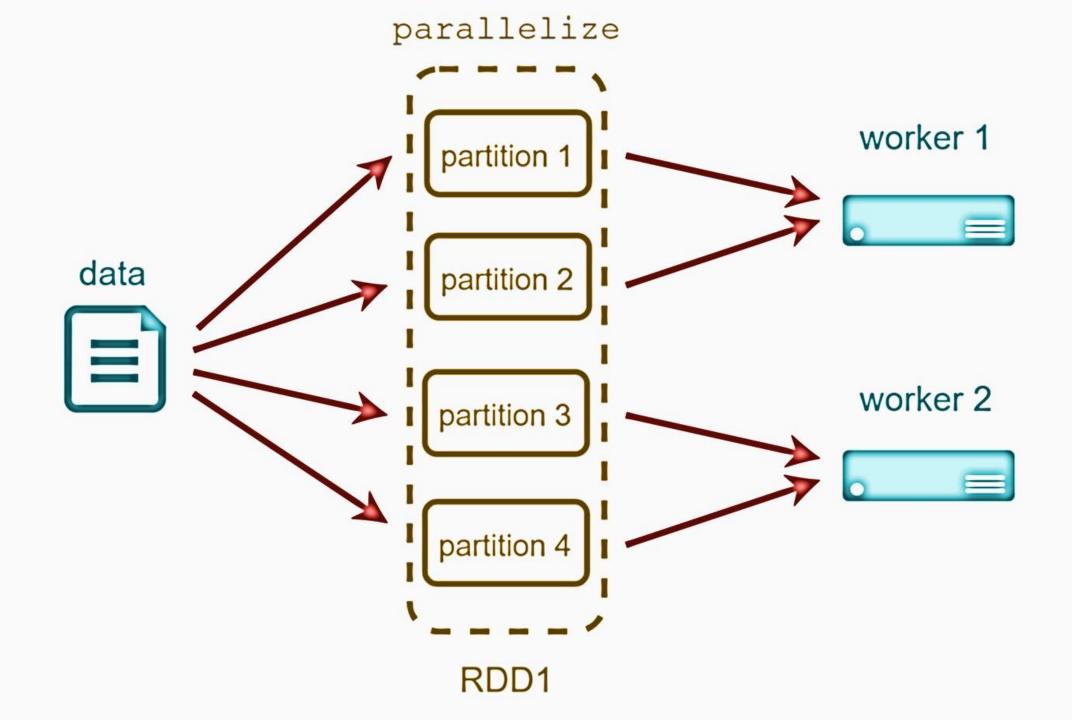
Resilient Distributed Dataset

An important building block of every Spark application is the "Resilient Distributed Dataset".

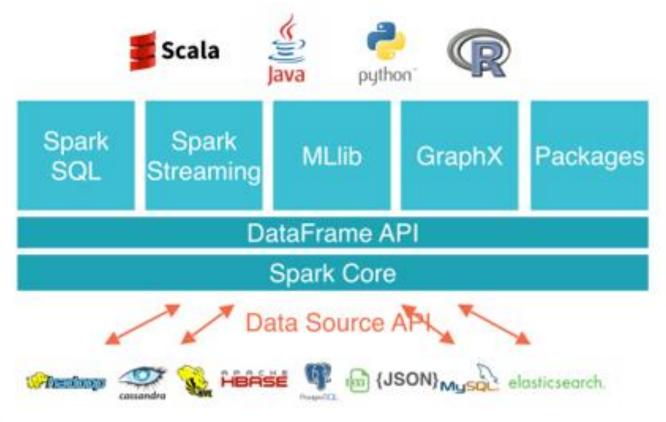
- "Resilient" means fault-tolerant and can rebuild data upon failure.
- Distributed with data residing on multiple nodes in a cluster. The RDD keeps track of data distribution across the workers.

Resilient Distributed Dataset (cont.)

- Dataset is a collection of partitioned data with primitive values or values of values, e.g., tuples or other objects (that represent records of the data with which you work).
- Users write applications that feed data into RDDs and each worker will work on a part of an RDD.



- You can code in Scala, Java, Python, R
- Interactive interpreters:Scala & Python





Computation Distribution

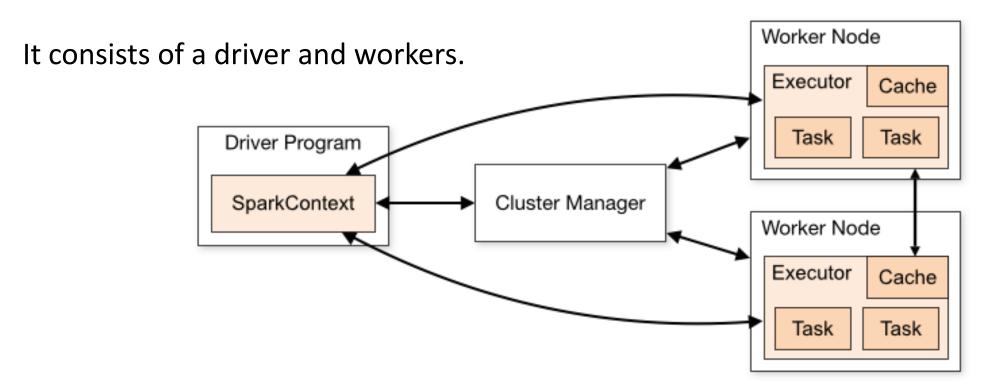
Spark Core contains the basic functionalities of Spark, such as the APIs that define RDDs, the operations and the actions (e.g., map, filter, reduce, etc.)

The rest of Spark's libraries are built on top of the RDD and Spark Core:

- Spark SQL for SQL and structured data processing. Every database table is represented as an RDD and Spark SQL queries are transformed into Spark operations.
- MLlib is a library of common machine learning algorithms implemented as Spark operations on RDDs. This library contains scalable learning algorithms like classifications, regressions, etc. that require iterative operations across large data sets. ML provides higher-level API built on top of DataFrames for constructing ML pipelines. The pipeline concept is inspired by the scikit-learn project.
- **GraphX** is a collection of algorithms and tools for manipulating graphs and performing parallel graph operations and computations.
- Spark Streaming for scalable, high-throughput, fault-tolerant stream processing of real-time data.

Spark Arch

The goal is to be able to process your data in parallel by the Spark runtime system.



There are other components such as schedulers, memory manager, but those are details.

Data storage

- Spark uses a variety of data storages:
 - ✓ HadoopHDFS
 - ✓ Apache Hbase
 - ✓ Cassandra
 - ✓ MapR-DB
 - ✓ MongoDB
 - ✓ Amazon S3

1st example

Download Spark

C ♠ spark.apache.org/downloads.html



Lightning-fast unified analytics engine

Download Libraries - Documentation - Examples Community - Developers -

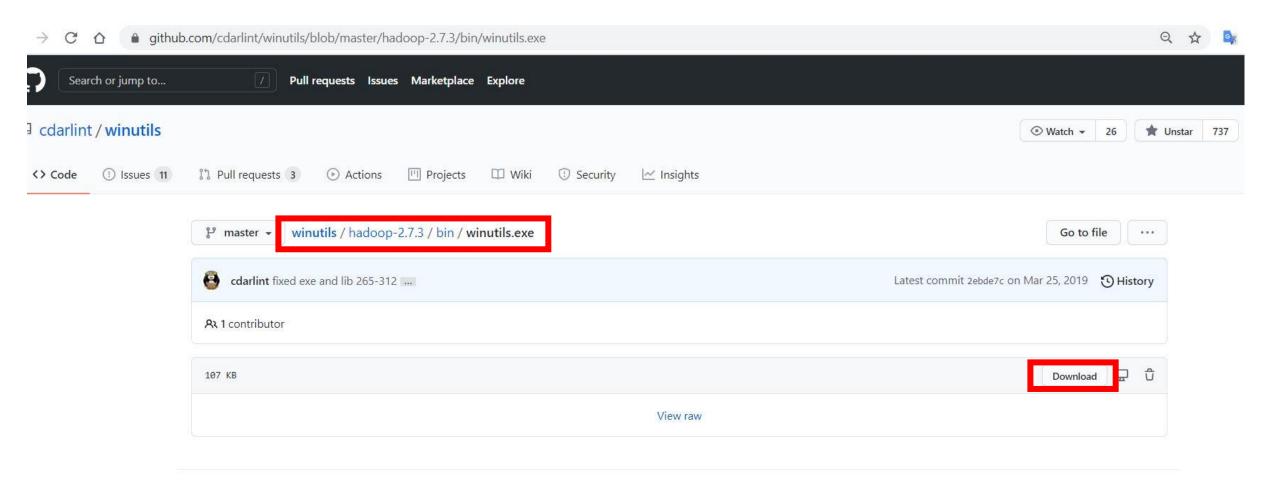
Download Apache Spark™

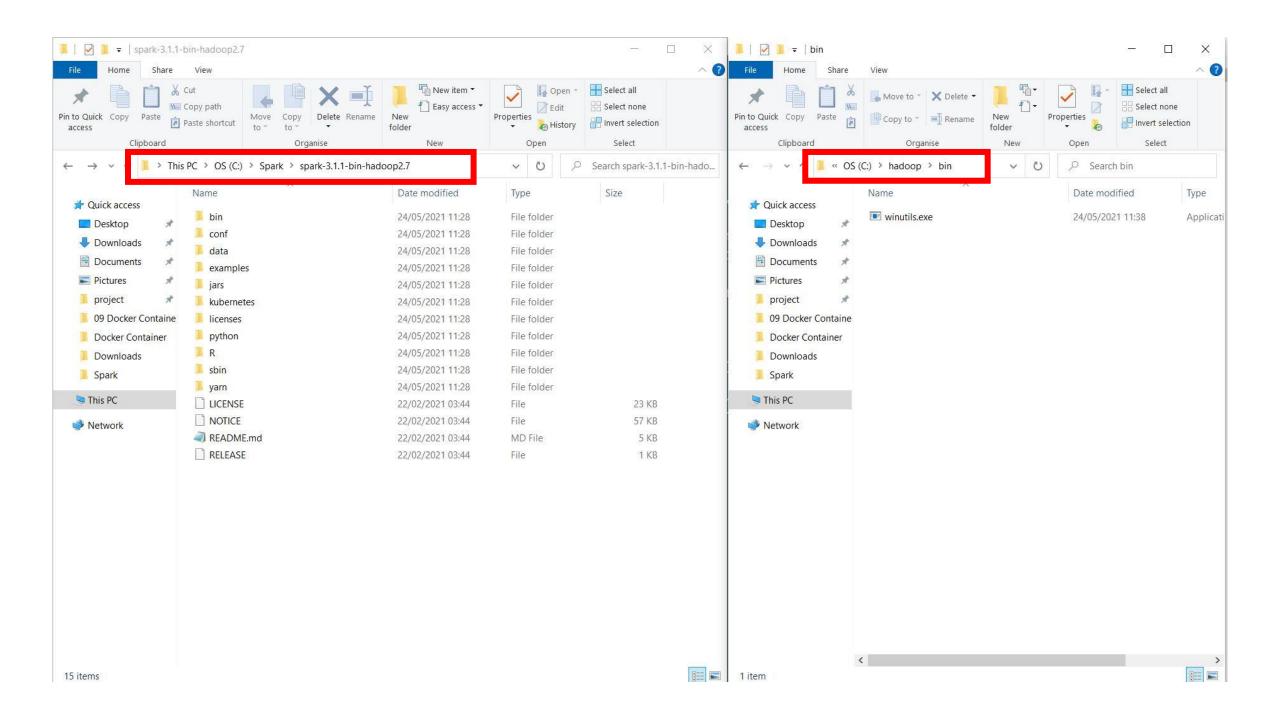
- 1. Choose a Spark release: 3.1.1 (Mar 02 2021) V
- 2. Choose a package type: Pre-built for Apache Hadoop 2.7
- 3 Download Spark: spark-3.1.1-bin-hadoop2.7.tgz
- 4. Verify this release using the 3.1.1 signatures, checksums and project release KEYS.

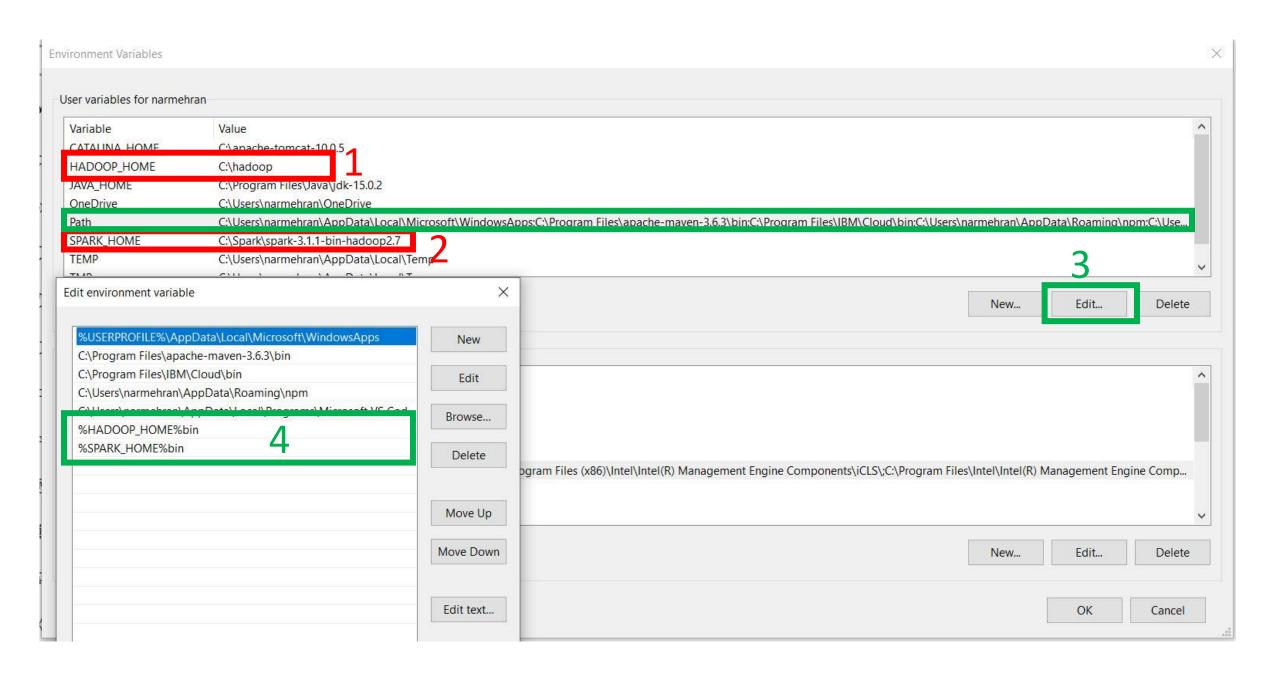
Note that, Spark 2.x is pre-built with Scala 2.11 except version 2.4.2, which is pre-built with Scala 2.12. Spark 3.0+ is pre-built with Scala 2.12.

V

Download winutils.exe file for the Hadoop version







To run Scala Spark interpreter

```
PS C:\WINDOWS\system32> spark-shell.cmd
WAKNING. An illegal reflective access operation has occurred
WARNING: Illegal reflective access by org.apache.spark.unsafe.Platform (file:/C:/Spark/spark-3.1.1-bin-hadoop2.7/jars/spark-unsafe_2.12
-3.1.1.jar) to constructor java.nio.DirectByteBuffer(long,int)
WARNING: Please consider reporting this to the maintainers of org.apache.spark.unsafe.Platform
WARNING: Use --illegal-access=warn to enable warnings of further illegal reflective access operations
WARNING: All illegal access operations will be denied in a future release
21/05/24 13:39:27 WARN NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where app
licable
Using Spark's default log4j profile: org/apache/spark/log4j-defaults.properties
Setting default log level to "WARN".
To adjust logging level use sc.setloglevel(newlevel). For SnarkR. use setlogLevel(newLevel).
Spark context Web UI available at http://DESKTOP-UHNUR97.mshome.net:4040
Spark context available as 'sc' (master - local["], app id - local 152185</del>373501).
Spark session available as 'spark'.
Welcome to
  Using Scala version 2.12.10 (Java HotSpot(TM) 64-Bit Server VM, Java 15.0.2)
Type in expressions to have them evaluated.
Type :help for more information.
scala>
```

To run Scala Spark interpreter

```
scala> spark.range(1000 * 1000 * 1000).count()
res1: Long = 1000000000
scala> val x =sc.textFile("C:\\Spark\\spark-3.1.1-bin-hadoop2.7\\README.md")
x: org.apache.spark.rdd.RDD[String] = C:\Spark\spark-3.1.1-bin-hadoop2.7\README.md MapPartitionsRDD[8] at textFile at <console>:24
scala> x.take(11).foreach(println)
# Apache Spark
Spark is a unified analytics engine for large-scale data processing. It provides
high-level APIs in Scala, Java, Python, and R, and an optimized engine that
supports general computation graphs for data analysis. It also supports a
rich set of higher-level tools including Spark SQL for SQL and DataFrames,
MLlib for machine learning, GraphX for graph processing,
and Structured Streaming for stream processing.
<https://spark.apache.org/>
```

To run Scala Spark interpreter

```
scala> val y = x.map( .reverse)
y: org.apache.spark.rdd.RDD[String] = MapPartitionsRDD[9] at map at <console>:25
scala> y.take(11).foreach(println)
krapS ehcapA #
sedivorp tI .gnissecorp atad elacs-egral rof enigne scitylana deifinu a si krapS
alacS ni sIPA level-hgih, avaJ alacS ni sIPA level-hgih
a stroppus osla tI .sisylana atad rof shparg noitatupmoc lareneg stroppus
,semarFataD dna LQS rof LQS krapS gnidulcni sloot level-rehgih fo tes hcir
gnissecorp hparg rof XhparG ,gninrael enihcam rof bilLM
.gnissecorp maerts rof gnimaertS derutcurtS dna
>/gro.ehcapa.kraps//:sptth
```

2nd example

Spark installation in Ubuntu

```
~$ pip3 install pyspark
~$ wget https://www.apache.org/dyn/closer.lua/spark/spark-3.0.1/spark-3.0.1.tgz
~$ tar xf spark-3.0.1.tgz
~$ cd spark-3.0.1/
~/spark-3.0.1$ export SPARK_HOME="/home/narges/spark-3.0.1"
~/spark-3.0.1$ export PATH="$SPARK_HOME/bin:$PATH"
~/spark-3.0.1$ export PYTHONPATH="$SPARK HOME/python:$PYTHONPATH"
~/spark-3.0.1$ export PYSPARK PYTHON=python3
~/spark-3.0.1$ spark-shell
~/spark-3.0.1$ pyspark
```

PySpark running?

```
narges@ThinkCentreM910s:~/spark-3.0.1$ pyspark
Python 3.6.9 (default, Oct 8 2020, 12:12:24)
[GCC 8.4.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
20/12/10 17:51:02 WARN Utils: Your hostname, ThinkCentreM910s resolves to a loopback address: 127.0.1.1; using 143.205.122.147 instead (on interface enp0s31f6)
20/12/10 17:51:02 WARN Utils: Set SPARK LOCAL IP if you need to bind to another address
WARNING: An illegal reflective access operation has occurred
WARNING: Illegal reflective access by org.apache.spark.unsafe.Platform (file:/home/narges/spark-3.0.1/assembly/target/scala-2.12/jars/spark-unsafe_2.12-3.0.1.jar) to constructor java.nio.DirectByteBuffer(lo
ng,int)
WARNING: Please consider reporting this to the maintainers of org.apache.spark.unsafe.Platform
WARNING: Use --illegal-access=warn to enable warnings of further illegal reflective access operations
WARNING: All illegal access operations will be denied in a future release
20/12/10 17:51:03 WARN NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicable
Using Spark's default log4j profile: org/apache/spark/log4j-defaults.properties
Setting default log level to "WARN".
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
/ _/_ ____/___
_\ \/ _ \/ _ \/ _ \/ _/ '_/
/_ / .__/\_,_/_/ /_\_\ version 3.0.1
Using Python version 3.6.9 (default, Oct 8 2020 12:12:24)
SparkSession available as 'spark'.
>>> nums = sc.parallelize([1,2,3,4])
>>> nums.map(lambda x: x*x).collect()
[1, 4, 9, 16]
```

PySpark wordcount example

narges@ThinkCentreM910s:~/spark-3.0.1/examples/src/main/python\$ python3 wordcount.py ~/spark-3.0.1/AUTHORS.txt

~/spark-3.0.1/examples/src/main/python\$ python3 wordcount.py ~/spark-3.0.1/AUTHORS.txt

Lambda function in Python

• The general structure of a lambda function is:

```
lambda <args>: <expr>
```

 For instance, "a python function to double the value of a scalar" is:

lambda x: x**2

Basic Transformations

```
# Create a simple RDD by simply passing some data such as a
file or a list
> nums = sc.parallelize([1, 2, 3])
# Pass each element through a function
> squares = nums.map(lambda x: x*x) // {1, 4, 9}
# Keep elements passing a predicate
> even = squares filter (lambda x: x % 2 == 0) // {4}
# Read a text file and count the number of lines containing
error
>lines = sc.textFile("file.log")
>lines.filter(lambda`s: "ERROR" in s).count()
```

Basic Actions

```
> nums = sc.parallelize([1, 2, 3])
# Retrieve RDD contents as a local collection
> nums.collect() # => [1, 2, 3]
# Return first K elements
> nums.take(2) # => [1, 2]
# Count number of elements
> nums.count() # => 3
# Merge elements with an associative function
> nums.reduce(lambda x, y: x + y) # => 6
# Write elements to a text file
> nums.saveAsTextFile("hdfs://file.txt")
```

Several Key-Value Operations

Example: Word Count

"to be or"
$$\longrightarrow$$
 "be" \longrightarrow (be, 1) \longrightarrow (be, 1) \longrightarrow (not, 1) \longrightarrow (not, 1) \longrightarrow "not to be" \longrightarrow "to" \longrightarrow (to, 1) \longrightarrow (or, 1) \longrightarrow (or, 1) \longrightarrow (or, 1) \longrightarrow (to, 1) \longrightarrow (to, 2) \longrightarrow "be" \longrightarrow "be"

Spark Operations

Transformations (define a new RDD)

https://spark.apache.org/docs/latest/rdd-programming-guide.html#transformations

map filter sample groupByKey reduceByKey

sortByKey

flatMap
union
join
cogroup
cross
mapValues

Actions

(return a result to driver program)

https://spark.apache.org/docs/latest/rddprogramming-guide.html#actions collect reduce count

lookupKey

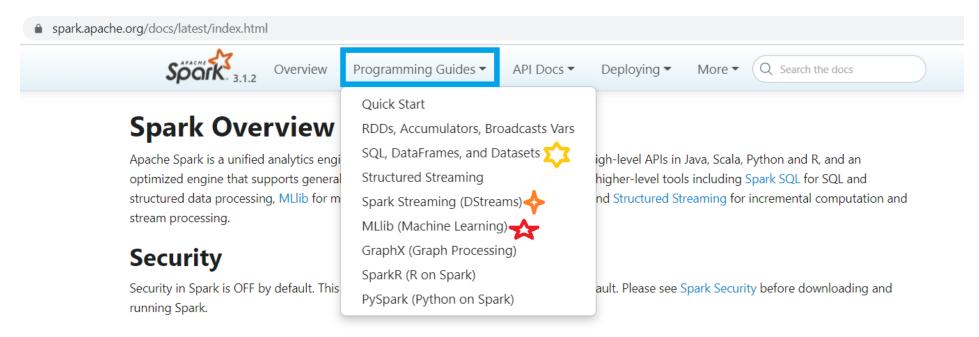
save

Running PySpark with Jupyter in Docker Containers

- Browse <u>DockerHub</u> and,
 - ✓ search for the image of <u>jupyter/pyspark-notebook</u> to start your trip with PySpark in Jupyter,
- or:
 - ✓ https://github.com/jupyter/docker-stacks/tree/master/pyspark-notebook
- https://realpython.com/pyspark-intro/

Assignment 11

 Each group can be responsible to consider one of the Spark features such as Spark Streaming, SQL or ML,



• Explain about this feature of Spark, prepare an example of it, execute it and get prepared to explain it line by line during the class.

Some links

- https://www.youtube.com/watch?v=3bWdJB96EF4
- https://www.youtube.com/watch?v=5dARTeE6OpU
- https://spark.incubator.apache.org/docs/latest/api/python/index.html
- https://spark.apache.org/docs/latest/api/python/getting_started/quickstart.html
- https://github.com/StephenHarrington/spark
- https://www.tutorialspoint.com/pyspark/pyspark mllib.htm
- https://www.datacamp.com/community/tutorials/apache-spark-tutorialmachine-learning

References

- https://spark.apache.org/downloads.html
- https://github.com/cdarlint/winutils
- https://phoenixnap.com/kb/install-spark-on-windows-10
- https://runawayhorse001.github.io/LearningApacheSpark/fnn.html
- Zaharia, M., et al. "Apache spark: a unified engine for big data processing." Communications of the ACM 59.11 (2016): 56-65.
- Zaharia, M. An Architecture for Fast and General Data Processing on Large Clusters. Ph.D. thesis, Electrical Engineering and Computer Sciences Department, University of California, Berkeley, 2014; https://www.eecs.berkeley.edu/Pubs/TechRpts/2014/EECS-2014-12.pdf
- https://github.com/himank/K-Means/blob/master/src/KMeans.java