

Spark

Cluster computing

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Outline i

Background

Introduction to Spark

What is Spark?

vs MapReduce

Key concepts

Working with RDDs

Using Spark

Full examples

Acknowledgements and useful links

Background

LISP and functional programming

- Higher level programming
- Avoid side effects
- Pattern matching

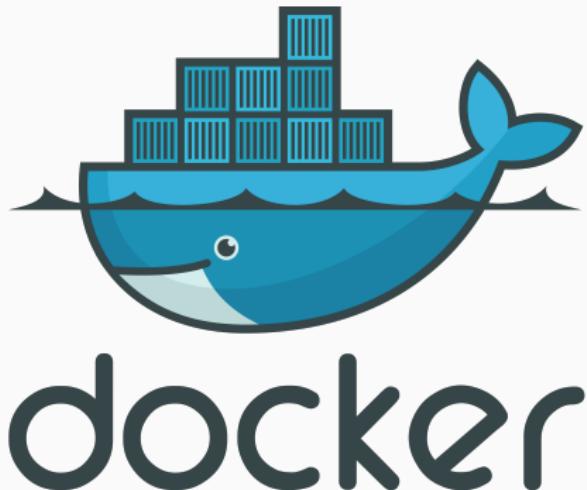


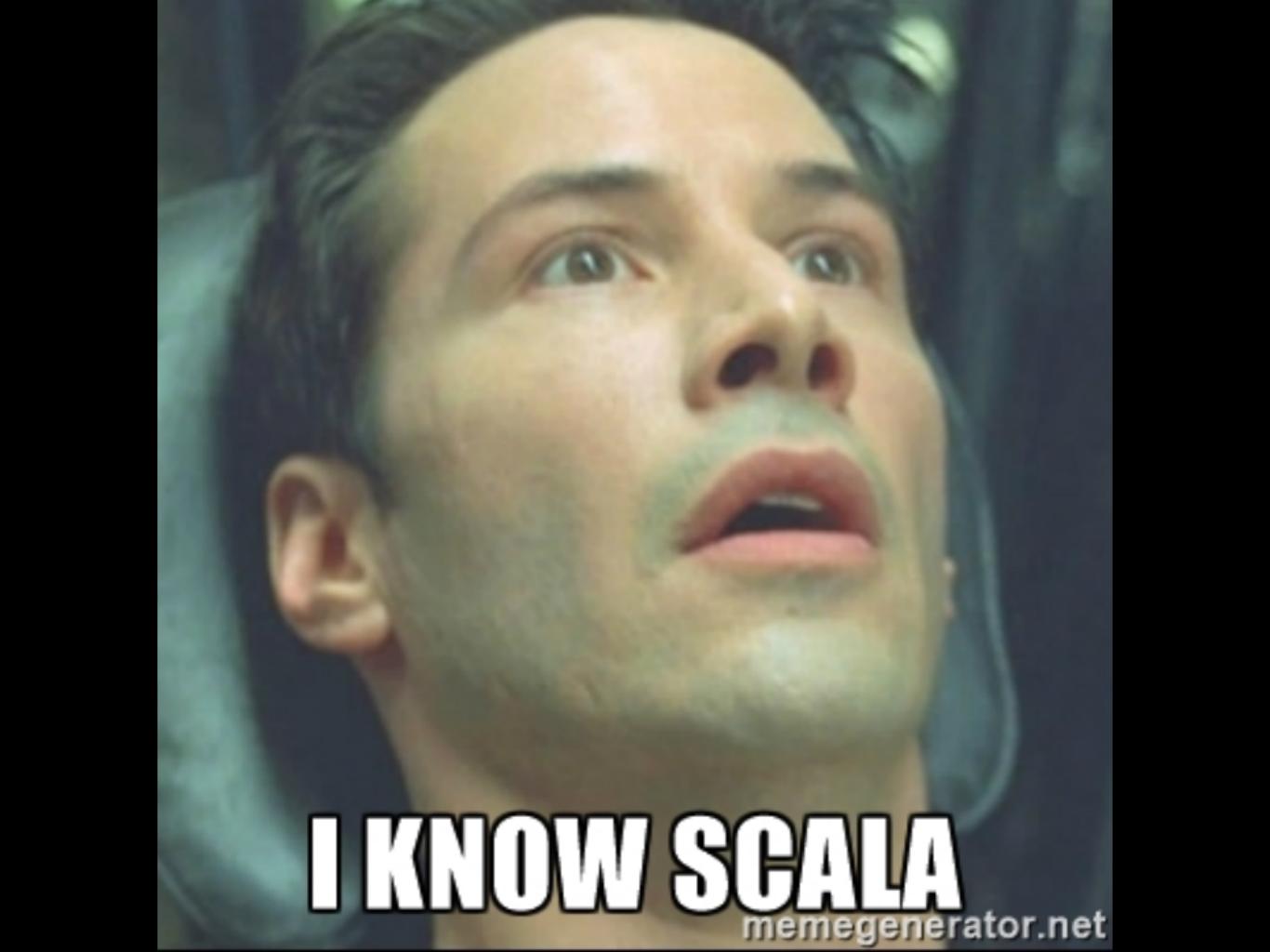


- A *better* Java
- Functional programming
(optional)
- Actors for (coarse)
concurrency

Docker

- Easy and repeatable deployment
- Lots of pre-built images @ hub.docker.com
- Building block for other tools (swarm, compose, machine...)

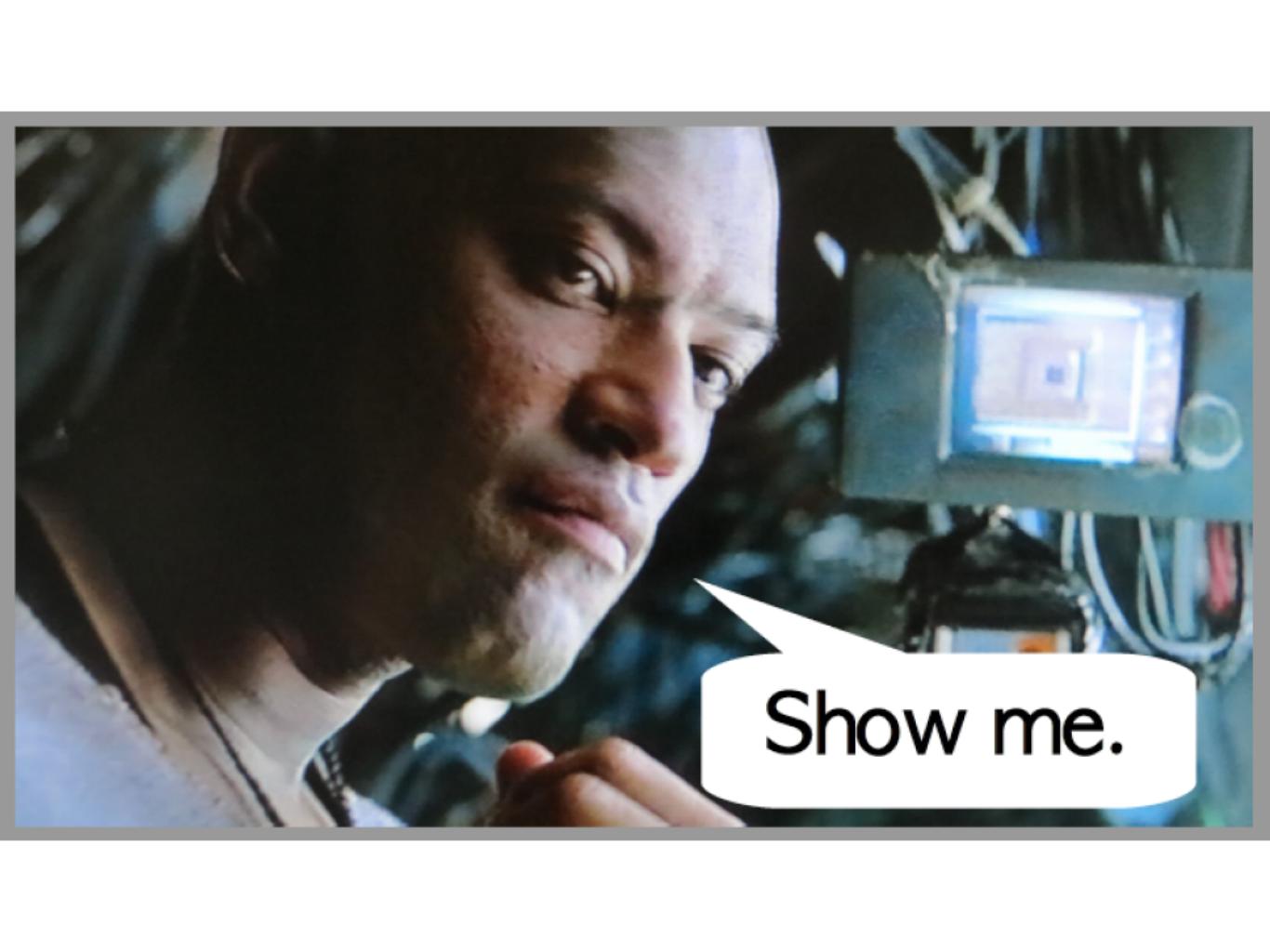


A close-up photograph of a young man with dark hair, looking upwards with a wide-eyed, surprised expression. His mouth is slightly open. The background is dark and out of focus.

I KNOW SCALA

A close-up photograph of a young man with dark hair, looking upwards with a wide-eyed, surprised expression. He has a slight smile on his lips. The background is dark and out of focus.

I KNOW BIG DATA



Show me.

Word count in Wikipedia

Problem: find the frequency each word is used in Wikipedia.

We have the text of all wikipedia in a text file¹. It begins like this:

```
hi world
```

```
hi
```

```
Scala (SKAH-lah) [9] is a general-purpose  
programming language. Scala has full support for  
functional programming and a strong static type  
system. Designed to be concise, [10] many of Scala  
's design decisions were inspired by criticism of  
Java's shortcomings. [8]
```

¹By happy coincidence, the first two lines are "hi world" and "hi"

Algorithm

- Read every line
- Chunk every line into words
- Count every occurrence
- For every word, sum its occurrences

Possible results of every step

```
(("hi world"), ("hi") ...)
```

```
List((hi, 1), (hi, 1), (world, 1) ...)
```

```
Map(hi -> ("hi", 1), (hi, 1)),  
    world -> ((world, 1)) ...)
```

```
Map(hi-> 2, world -> 1 ...)
```

Possible results of every step

```
(("hi world"), ("hi") ...)  
List(hi, 1), (hi, 1), (world, 1) ...  
Map(hi -> ("hi", 1), (hi, 1)),  
    world -> ((world, 1)) ...)  
Map(hi-> 2, world -> 1 ...)
```

Possible results of every step

(("hi world"), ("hi") ...)

List(hi, 1), (hi, 1), (world, 1) ...)

Map(hi -> ("hi", 1), (hi, 1)),
world -> ((world, 1)) ...)

Map(hi-> 2, world -> 1 ...)

Possible results of every step

```
(("hi world"), ("hi") ...)  
List((hi, 1), (hi, 1), (world, 1) ...)  
Map(hi ->(("hi", 1), (hi, 1)),  
    world -> ((world, 1)) ...)  
Map(hi-> 2, world -> 1 ...)
```

Running the scala shell

We will use docker.

```
docker run -it -v $PWD:Wikipedia.txt:Wiki \
--rm williamyeh/scala
```

Scala code

```
import scala.io.Source
val wiki = Source.fromFile("Wiki").getLines
wiki.flatMap(line=> line.split(" "))
  .map(x=>(x, 1)).toList
  .groupBy(x => x._1)
  .map({case (k, v) =>
    (k, v.foldLeft(0)((a, b) => a+b._2))})
```

Scala code

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import scala.io.Source
val wiki = Source.fromFile("Wiki").getLines
wiki.flatMap(line=> line.split(" "))
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        (k, v.foldLeft(0)((a, b) => a+b._2))})
```

Scala code

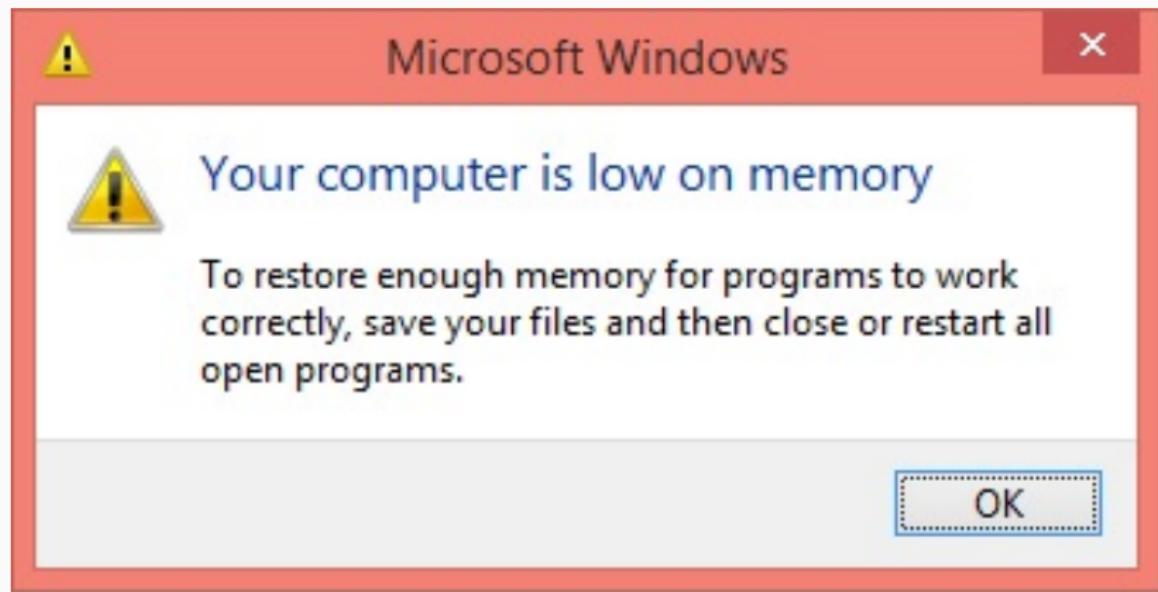
```
import scala.io.Source
val wiki = Source.fromFile("Wiki").getLines
wiki.flatMap(line=> line.split(" „"))
    map(x=>(x, 1)).toList
    groupBy(x => x._1)
    map({case (k, v) =>
        (k, v.foldLeft(0)((a, b) => a+b._2)))})
```

Scala code

```
import scala.io.Source
val wiki = Source.fromFile("Wiki").getLines
wiki.flatMap(line=> line.split(" „"))
    map(x=>(x, 1)).toList
    groupBy(x => x._1)
    map({case (k, v) =>
        (k, v.foldLeft(0)((a, b) => a+b._2))})
```

Let's run it in the shell.





What happened?

Limited resources

- CPU limits our speed
 - Multi-cores help...
 - ...but real parallelism is **hard**
- RAM limits how much data you can process at the same time
 - What if you need more than 128GB?
 - You could use more than one computer...
 - ... but cluster computing is even harder than “local” parallelism
- Functional programming helps a bit

But... this was supposed to be fun,
wasn't it?

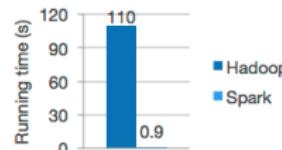
Introduction to Spark

Apache Spark™ is a fast and general engine for large-scale data processing.

Speed

Run programs up to 100x faster than Hadoop MapReduce in memory, or 10x faster on disk.

Apache Spark has an advanced DAG execution engine that supports cyclic data flow and in-memory computing.



Logistic regression in Hadoop and Spark

Ease of Use

Write applications quickly in Java, Scala, Python, R.

Spark offers over 80 high-level operators that make it easy to build parallel apps. And you can use it *interactively* from the Scala, Python and R shells.

```
text_file = spark.textFile("hdfs://...")
```

```
text_file.flatMap(lambda line: line.split())
.map(lambda word: (word, 1))
.reduceByKey(lambda a, b: a+b)
```

Word count in Spark's Python API

Latest News

Spark wins CloudSort Benchmark as the most efficient engine (Nov 15, 2016)

Spark 2.0.2 released (Nov 14, 2016)

Spark 1.6.3 released (Nov 07, 2016)

Spark 2.0.1 released (Oct 03, 2016)

[Archive](#)

[Download Spark](#)

Built-in Libraries:

[SQL and DataFrames](#)
[Spark Streaming](#)
[MLlib \(machine learning\)](#)
[GraphX \(graph\)](#)

[Third-Party Packages](#)

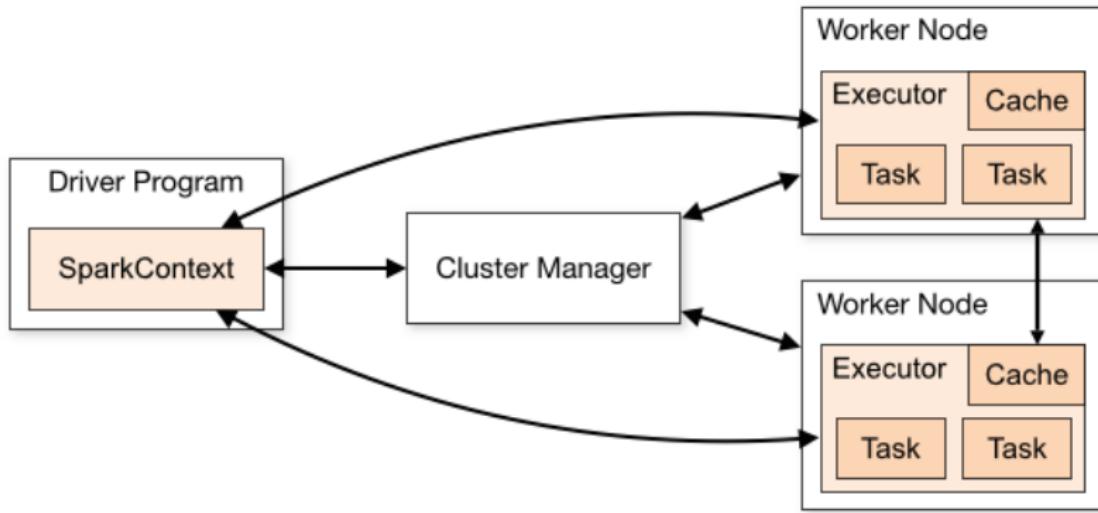
Apache Spark™ is a fast and general engine for large-scale data processing.²

On top of that:

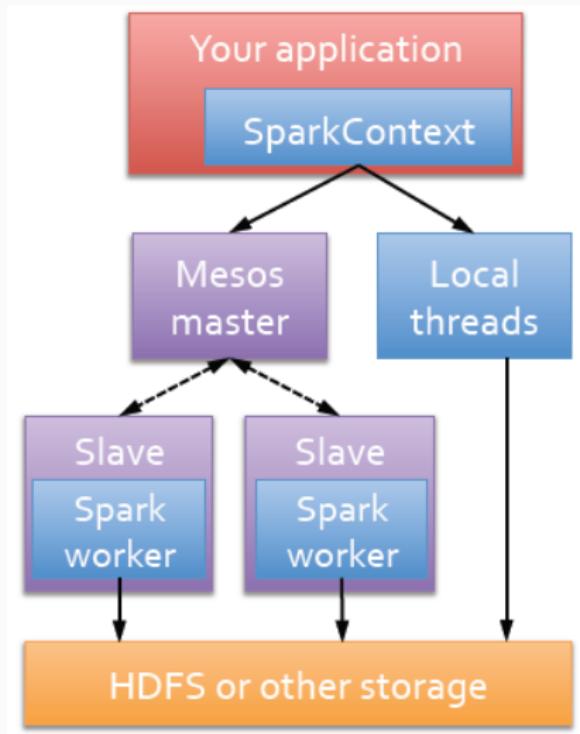
- Open source (Top-level Apache project)
- Plays well with other tools

²<http://spark.apache.org>

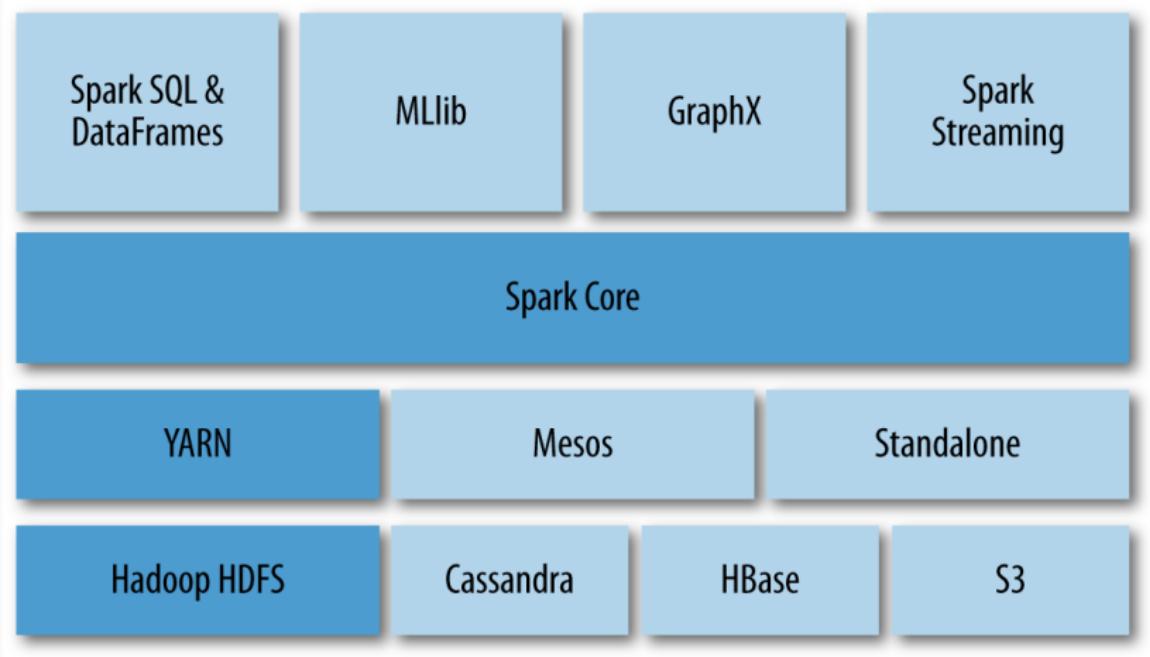
Architecture



Programs

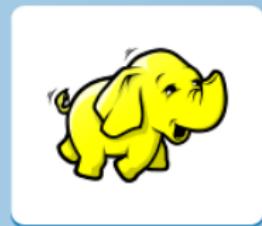


Ecosystem





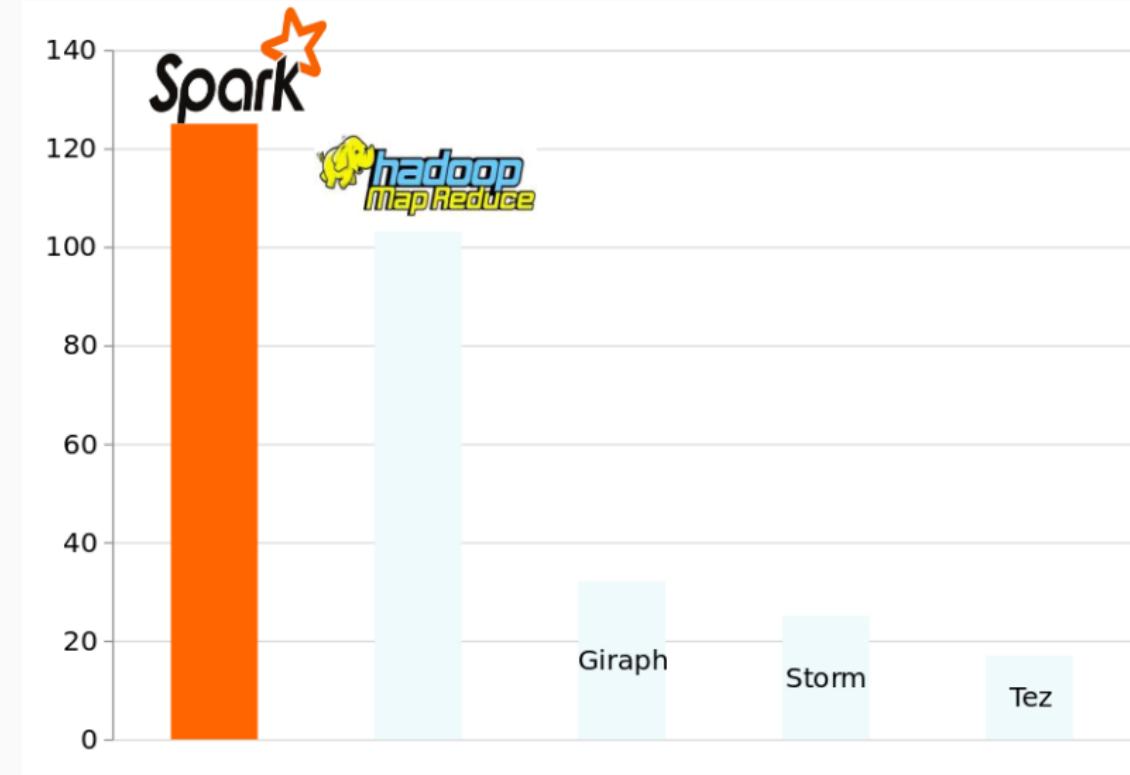
VS.



Comparison to MapReduce

- In-memory data
 - Less i/o overhead
 - Faster operations
 - Caching
- Better for recursive tasks (e.g. machine learning)
- Some libraries are dropping MapReduce support

Contributors to Spark/Hadoop 2014



Project status

apache / spark

View Repository Watch 2,081 Star 19,117 Fork 17,165

Code Pull requests 523 Projects Insights

Pulse

Contributors

Community

Commits

Code frequency

Dependency graph

Network

Mar 28, 2010 – Oct 17, 2018 Contributions: Commits

Contributions to master, excluding merge commits

YAHOO! amplab UC BERKELEY DATABRICKS MAPR Alibaba.com Hortonworks Amazon web services

wandisco intel AdMobius ClearStory IBM TAGGED

CLOUD PHYSICS guavus CONVIVA Adobe bizo webtrends

stratio HHMI Janelia farm

ADATAD sharethrough OYALA quantifind tuplejump cloudera

Overview

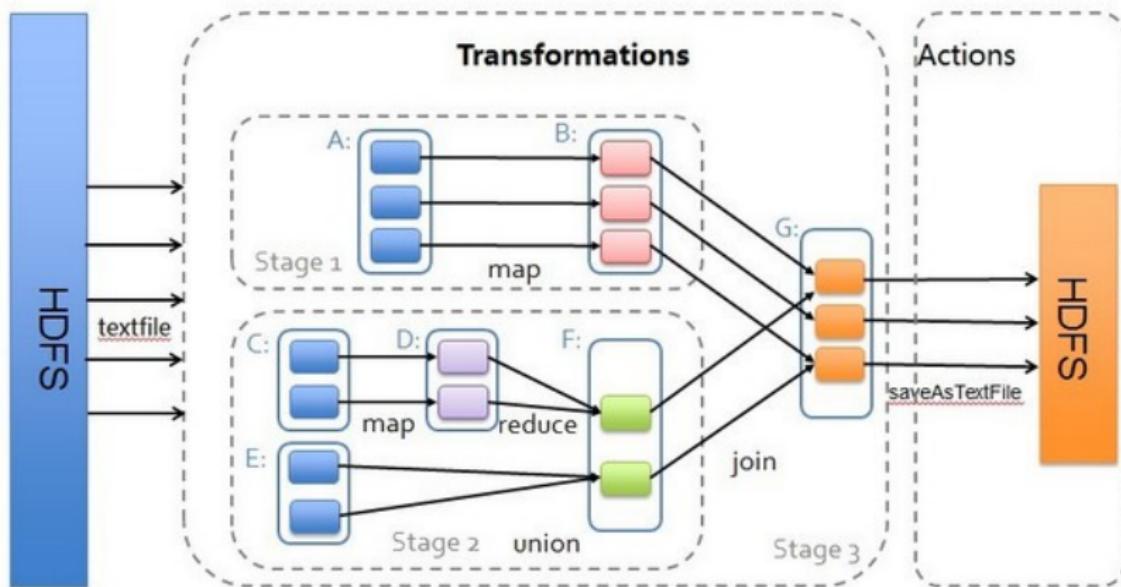
Data (RDDs/Datasets)

- RDD: Resilient Distributed Dataset
- Collections of objects spread across a cluster, stored in RAM or on Disk
- Built through parallel transformations
- Automatically rebuilt on failure

Operations

- Transformations (e.g. group, map, groupBy)
- Actions (e.g. count, collect, save)

Transformations and actions



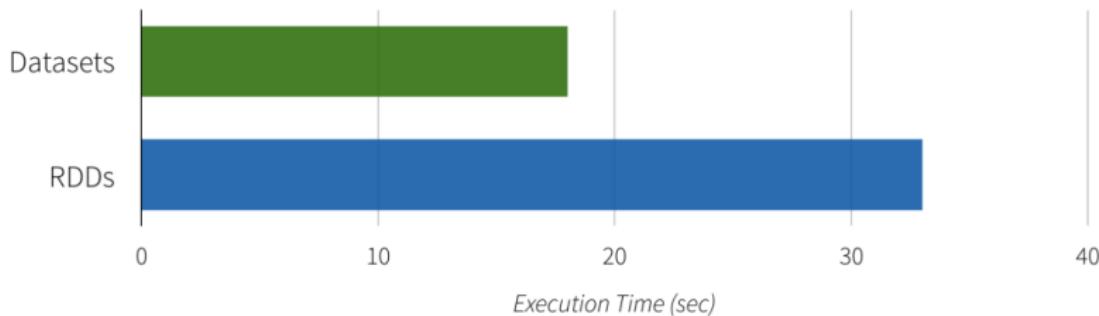
RDDs vs Datasets

Datasets are the future

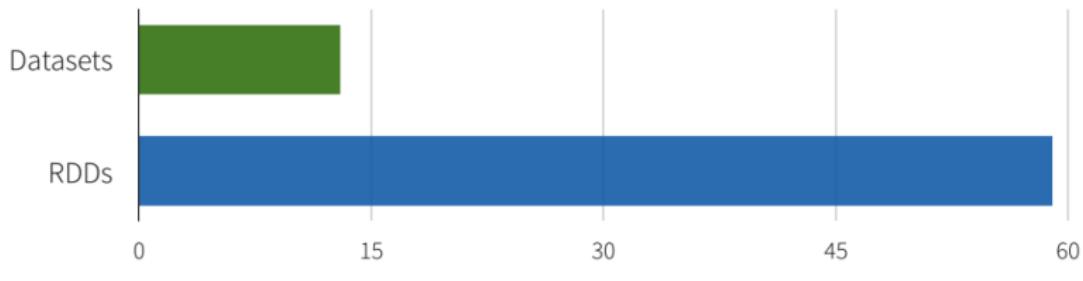
- More memory efficient
- Libraries dropping support for RDDs

RDDs vs Datasets

Distributed Wordcount



Memory Usage when Caching



Language support

Scala

```
val lines = sc.textFile(...)  
lines.filter(x => x.contains("ERROR")).count()
```

Python

```
lines = sc.textFile(...)  
lines.filter(lambda s: "ERROR" in s).count()
```

Java

Removed, to keep the slides clean :)

sc is the Spark Context (more on this later)

Different flavors

Language	App	REPL	Performance
Scala	Yes	Yes	
Java	Yes	No	
Python	Yes	Yes	

The Read-eval-print-loop (REPL) is the easiest way to get started and explore datasets. It is just a special Spark application that accepts user input (scala code).

Working with RDDs

Creation

From normal data structures

```
val nums = sc.parallelize(List(1, 2, 3))
val cont = sc.parallelize(List(("a", 1),
                               ("a", 1),
                               ("b", 3)))
```

From distributed/local sources

```
sc.textFile('myfile')
```

Note: sc is the spark context in the Spark interpreter

Operations: collect

collect()

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: collect

Runs any pending transformation and returns the real values

```
nums.collect()
```

```
> List(1, 2, 3)
```

```
cont.collect()
```

```
> List((a, 1), (a, 1), (b, 3))
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: take

`take(N)`

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: take

Returns the N first elements

```
nums.take(2)
```

```
> List(1, 2)
```

```
cont.take(1)
```

```
> List((a, 1))
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: count

count()

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: count

Returns the number of elements in a collection

```
nums.count()
```

```
> 3
```

```
cont.count()
```

```
> 3
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: filter

`filter(fn)`

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: filter

filter(fn)

This time, we need to define a function.

Filter applies that function to every element, and returns those where the function returns true.

For example:

```
val fn = (x:Int) => x > 1
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: filter

```
val fn = (x:Int) => x > 1
```

Return a list containing the values where the function returns true

```
nums.filter(fn)
```

```
> List(2, 3)
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Quick aside: anonymous functions and underscores

In Scala, we can define “anonymous functions”, also known as lambda functions.

Quick aside: anonymous functions and underscores

In Scala, we can define “anonymous functions”, also known as lambda functions.

```
val fn = (x:Int) => x > 1  
cont.filter(fn)
```

is equivalent to:

```
cont.filter((x:Int) => x>1)
```

Quick aside: anonymous functions and underscores

```
val fn = (x:Int) => x > 1  
cont.filter(fn)
```

```
cont.filter((x:Int) => x>1)
```

Additionally, the scala compiler is smart enough to infer types in this example. Hence, we could simply write:

```
cont.filter(x => x>1)
```

Quick aside: anonymous functions and underscores

```
val fn = (x:Int) => x > 1  
cont.filter(fn)
```

```
cont.filter((x:Int) => x>1)
```

```
cont.filter(x => x>1)
```

Furthermore, we could use underscores to replace arguments:

```
cont.filter(_>1)
```

Quick aside: anonymous functions and underscores

Every new argument in the lambda function represents a parameter

Hence, these two expressions are equivalent

$- + -$
 $(x, y) \Rightarrow x+y$

Operations: filter

Our last example could be written more concisely as:

```
nums.filter(_ > 1)
```

```
> List(2, 3)
```

Operations: filter

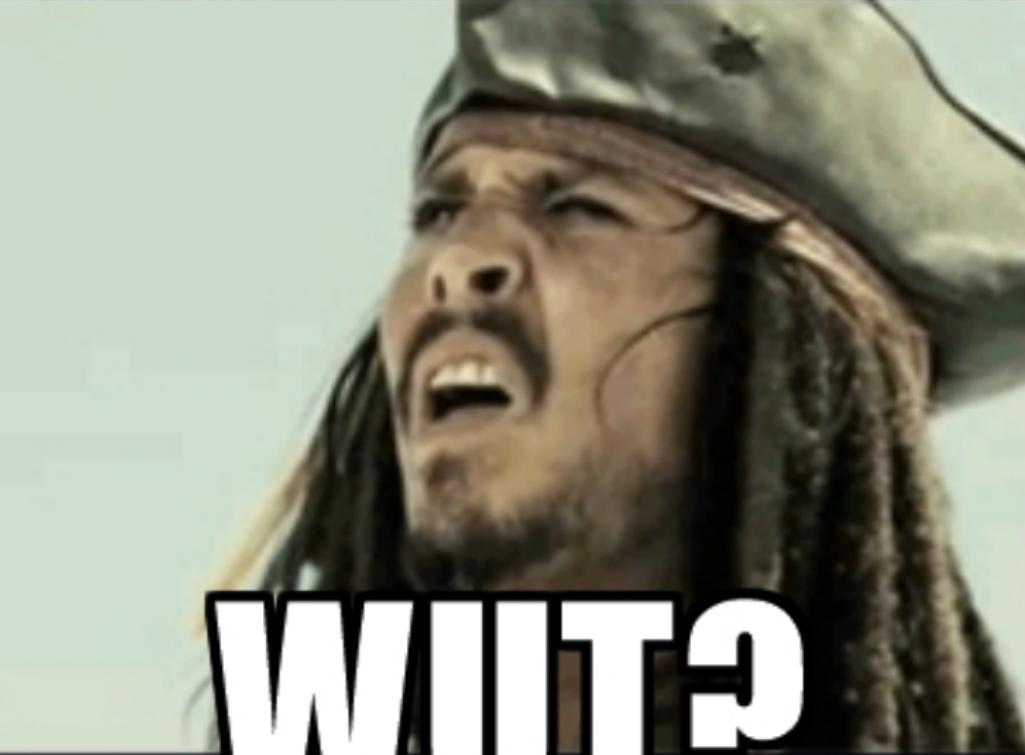
What would this filter do?

```
cont.filter(_.1 == "a" && _.1 == 1)
```

```
> ???
```

Operations: filter

```
<console>:9: error: missing parameter type
for expanded function ((x$1, x$2) =>
x$1._1.$eq$eq(a).ampamp(x$2._1.$eq$eq(1)))
Note: The expected type requires a
one-argument function accepting a 2-Tuple.
```



WUT?

Operations: filter

Remember, each new underscore represents a new argument. So that expression expands to:

```
cont.filter((x, y) => x._1 == "a" &&  
           y._2 == 1)
```

Operations: filter

The right expression is:

```
cont.filter(x => x._1 == "a" &&  
            x._2 == 1)  
> List((a, 1), (a, 1))
```

Operations: map

map(fn)

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: map

Apply a function to every item in the list

```
cont.map(x._2)
```

```
> List(1, 1, 3)
```

```
nums.map(_*3)
```

```
> List(3, 6, 9)
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: reduce

reduce(fn)

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations: reduce

Merge elements with an associative function (concisely)

```
nums.reduce(_+_)
```

```
> 6
```

```
cont.reduce((x, y) => (x._1+y._1,
                           x._2*y._2))
```

```
> (aab, 3)
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations

groupByKey()

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations

Group elements of a list by the first item in the tuple

```
cont.groupByKey()
```

```
> [(b, [3]), (a, [1,1])]
```

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

Operations

reduceByKey(fn)

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

³Databricks' post on avoiding groupByKey

Operations

Group by key and reduce each value

```
cont.reduceByKey((x,y)=>x+y)
```

```
> [(b,3), (a,2)]
```

reduceByKey is more efficient than applying group, map and reduce separately. The reduce function can be given to each worker, which avoids passing unnecessary data.³

Reminder:

```
nums: List(1, 2, 3)
```

```
cont: List(("a", 1), ("a", 1), ("b", 3))
```

³Databricks' post on avoiding groupByKey

Operations

And once you are done, save your results to a file.

```
nums.saveAsTextFile("hdfs://file.txt")
```

Example: Search in logs

```
val lines = sc.textFile("hdfs://...")  
val errors = lines.filter(s =>  
                          s.startsWith("ERROR"))  
val messages = errors.map(s => s.split("\t")._2)  
messages.cache()  
messages.filter(s => s.contains("mysql")).count()  
messages.filter(s => s.contains("php")).count()
```

Using Spark

Local deployment using docker-compose

Get the repo

```
git clone http://github.com/gettyimages/docker-spark  
cd docker-spark  
docker-compose up  
docker exec -it dockerspark_master_1 bin/spark-shell
```

Local deployment using docker-compose

Move to the repo

```
git clone http://github.com/gettyimages/docker-spark  
cd docker-spark  
docker-compose up  
docker exec -it dockerspark_master_1 bin/spark-shell
```

Local deployment using docker-compose

Run all the containers

```
git clone http://github.com/gettyimages/docker-spark  
cd docker-spark  
docker-compose up  
docker exec -it dockerspark_master_1 bin/spark-shell
```

Local deployment using docker-compose

Launch spark-shell inside the master container

```
git clone http://github.com/gettyimages/docker-spark  
cd docker-spark  
docker-compose up  
docker exec -it dockerspark_master_1 bin/spark-shell
```

Demo

```
worker_1 | 16/11/16 11:53:32 INFO worker.Worker: Connecting to master master:7077...
worker_1 | 16/11/16 11:53:32 INFO handler.ContextHandler: Started o.s.j.s.ServletContextH
ndler@77f93121{/metrics/json,null,AVAILABLE}
worker_1 | 16/11/16 11:53:32 INFO client.TransportClientFactory: Successfully created con
nection to master/172.17.0.2:7077 after 25 ms (0 ms spent in bootstraps)
master_1 | 16/11/16 11:53:32 INFO master.Master: Registering worker 172.17.0.3:8881 with 1
cores, 1024.0 MB RAM
worker_1 | 16/11/16 11:53:32 INFO worker.Worker: Successfully registered with master spar
://master:7077
master_1 | 16/11/16 11:53:47 INFO master.Master: Registering app Spark shell
worker_1 | 16/11/16 11:53:48 INFO spark.SecurityManager: Changing view acls to: root
worker_1 | 16/11/16 11:53:48 INFO spark.SecurityManager: Changing modify acls to: root
worker_1 | 16/11/16 11:53:48 INFO spark.SecurityManager: Changing view acls groups to:
worker_1 | 16/11/16 11:53:48 INFO spark.SecurityManager: Changing modify acls groups to:
worker_1 | 16/11/16 11:53:48 INFO worker.ExecutorRunner: Launchtcommand::"/usr/jdk1.8.0_9
/bin/java"-cp"a"/conf:/usr/spark-2.0.1/jars/*:/usr/hadoop-2.7.2/etc/hadoop:/usr/hadoop
```

 version 2.0.1

```
Using Scala version 2.11.8 (Java HotSpot(TM) 64-Bit Server VM, Java 1.8.0_92)
Type in expressions to have them evaluated.
Type :help for more information.
```

```
scala> val a = sc.parallelize(List(1,2,3,4))
a: org.apache.spark.rdd.RDD[Int] = ParallelCollectionRDD[0] at parallelize at <console>:24
```

Compose.yml Master

```
master:  
    image: gettyimages/spark  
    command: bin/spark-class org.apache.spark.deploy.  
             ↴ master.Master -h master  
    hostname: master  
    environment:  
        MASTER: spark://master:7077  
        SPARK_CONF_DIR: /conf  
        SPARK_PUBLIC_DNS: localhost  
        ... bunch of ports ...  
    volumes:  
        - ./conf/master:/conf  
        - ./data:/tmp/data
```

Compose.yml Master

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    SPARK_CONF_DIR: /conf  
    SPARK_PUBLIC_DNS: localhost  
    ... bunch of ports ...  
  volumes:  
    - ./conf/master:/conf  
    - ./data:/tmp/data
```

Compose.yml Worker

```
worker:
  image: gettyimages/spark
  command: bin/spark-class org.apache.spark.deploy.
    ↴ worker.Worker spark://master:7077
  hostname: worker
  environment:
    SPARK_CONF_DIR: /conf
    SPARK_WORKER_CORES: 2
    SPARK_WORKER_MEMORY: 1g
  links:
    - master
  volumes:
    - ./conf/worker:/conf
    - ./data:/tmp/data
```

Compose.yml Worker

```
worker:
  image: gettyimages/spark
  command: bin/spark-class org.apache.spark.deploy.
    ↴ worker.Worker spark://master:7077
  hostname: worker
  environment:
    SPARK_CONF_DIR: /conf
    SPARK_WORKER_CORES: 2
    SPARK_WORKER_MEMORY: 1g
  links:
    - master
  volumes:
    - ./conf/worker:/conf
    - ./data:/tmp/data
```

Compose.yml Worker

```
worker:
  image: gettyimages/spark
  command: bin/spark-class org.apache.spark.deploy.
            ↴ worker.Worker spark://master:7077
  hostname: worker
  environment:
    SPARK_CONF_DIR: /conf
    SPARK_WORKER_CORES: 2
    SPARK_WORKER_MEMORY: 1g
  links:
    - master
  volumes:
    - ./conf/worker:/conf
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volumes:
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  - ./data:/tmp/data
```

Useful info

- ./data folder is mounted as /tmp/data
 - Copy your datasets there
 - Load them in the shell: `sc.textFile("/tmp/data/...")`
- Master Web UI (localhost:8080)
- Worker Web UI (localhost:8081)
- REPL Web UI (localhost:4040 when launched)

A word of caution: as any other app, the shell reserves resources on startup, whether you are using them or not.

Applications

Steps:

- Write the code
- Compile the jar
- Make your data available to every node in the cluster
- Submit it to your cluster

Writing applications

Example application

```
import org.apache.spark.SparkContext
import org.apache.spark.SparkContext._
import org.apache.spark.SparkConf

object SparkWordCount {
    def main(args: Array[String]) {
        // create Spark context with Spark configuration
        val sc = new SparkContext(new SparkConf().
            ↴ setAppName("Spark\u201cExample"))
        ...
    }
}
```

Running an application

```
./bin/spark-submit --class <main-class> \
                  --master <master-url> \
                  --deploy-mode <deploy-mode> \
                  --conf <key>=<value> \
                  ... # other options
                  <application-jar> \
                  [application-arguments]
```

Full examples

Word frequency in wikipedia, revisited

Spark

```
val wiki = sc.textFile("Wikipedia.txt")
val counts = wiki.flatMap(line=> line.split(" "))
              map(word => (word, 1)))
              reduceByKey(_ + _)
```

Pure scala

```
val wiki = scala.io.Source.fromFile("Wiki").getLines
wiki.flatMap(line=> line.split(" "))
          map(x=>(x, 1)).toList
          groupBy(x => x._1)
```

Shall we try it in the shell?



Spark is not magic

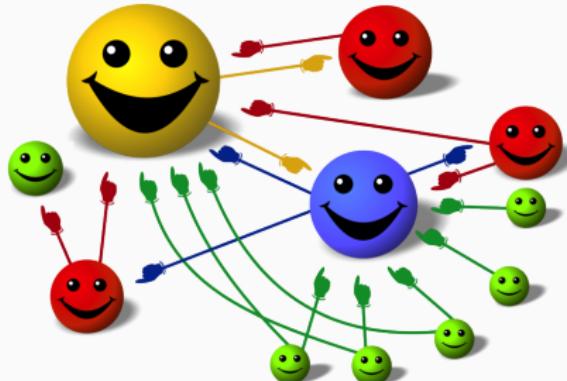
- We still have to add more resources
- Caching may cause the spark version to use **more memory**
(this can be configured)

Spark is not magic

- We still have to add more resources
- Caching may cause the spark version to use **more memory**
(this can be configured)

However, it allows us to scale our
application

Page rank



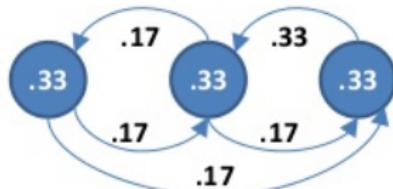
- Created by Google
- Rank given by links and their importance
- Iterative (Perfect for Spark!)

PageRank of site = $\sum \frac{\text{Page rank of inbound link}}{\text{Number of links on that page}}$

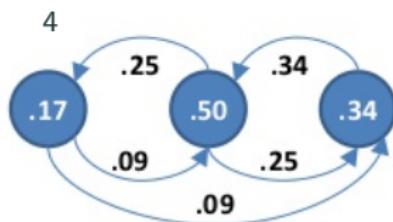
OR

$$PR(u) = (1 - d) + d \times \sum \frac{PR(v)}{N(V)}$$

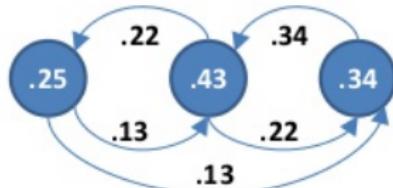
PageRank toy example



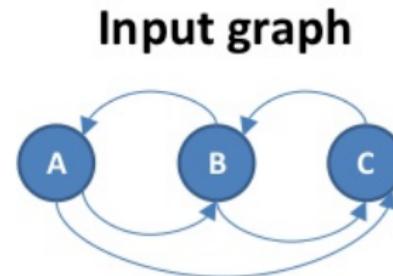
Superstep 0



Superstep 1



Superstep 2



⁴<http://www.slideshare.net/sscdotopen/large-scale>

Spark Program

```
val links = // RDD of (url, neighbors) pairs
var ranks = // RDD of (url, rank) pairs

for (i <- 1 to ITERATIONS) {
    val contribs = links.join(ranks).flatMap {
        case (url, (links, rank)) =>
            links.map(dest => (dest, rank/links.size))
    }
    ranks = contribs.reduceByKey(_ + _)
        .mapValues(0.15 + 0.85 * _)
}

ranks.saveAsTextFile(...)
```

Next week

Next week

- Advanced Spark configuration
- Multiple hosts
- Spark ecosystem
- More examples in IBM BlueMix

Acknowledgements and useful links

- Spark programming guide
- Databricks introducing apache spark datasets
- Data Analytics with Hadoop: In-Memory Computing with Spark