

Dive into Apache Spark

TANG Gen

Outline

- Detailed description of RDD
 - Formalized conception of RDD
 - RDD graph
- Spark architecture
 - Spark runtime architecture
 - Memory design
- Tips for Spark coding
- Testing Spark

/What is RDD?

Just for review:

```
val sc = new SparkContext(...)
val data = sc.textFile("hdfs://")
val data1 = rdd.filter(_.contains("..."))
```

- We know that data and data1 are RDDs
- But, what is the formalized conception of RDD?

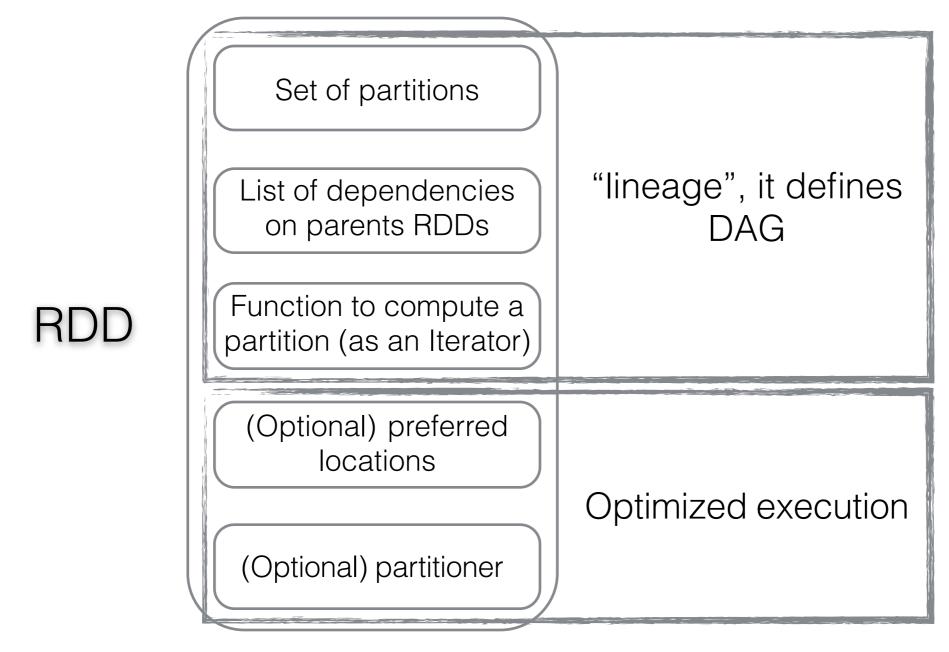
/Formalized conception of RDD

- RDD could be
 - A distributed collection of objects on disk
 - A distributed collection of objects on memory
 - A distributed collection of objects on HDFS
 - A distributed collection of objects on Cassandra

• ...

/Formalized conception of RDD

Scientifically, RDD is an interface with five members



/Examples

- For the RDD "data"
 - partitions = one per HDFS block
 - dependencies = None
 - compute = read corresponding block
 - preferredLocations = HDFS block location
 - partitioner = None

/Examples

- For the RDD "data1"
 - partitions = same as parent RDD
 - dependencies = "one-to-one" on parent
 - compute = compute parent and filter it
 - preferredLocations = None
 - partitioner = None

/RDD graph

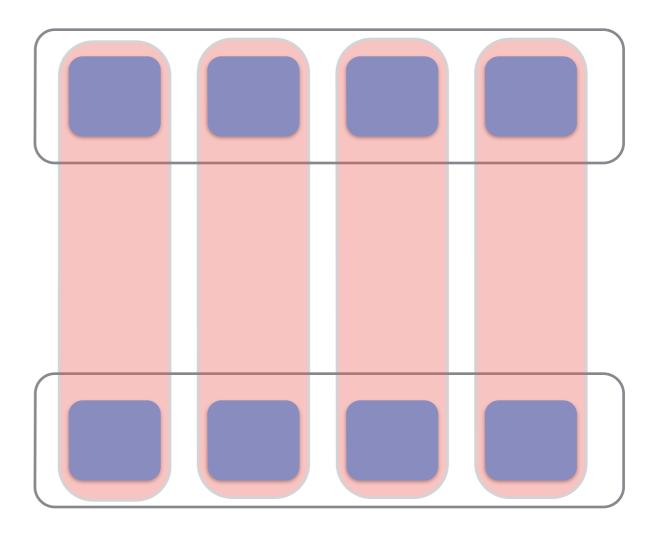
Data-level view

HadoopRDD path = hdfs://...



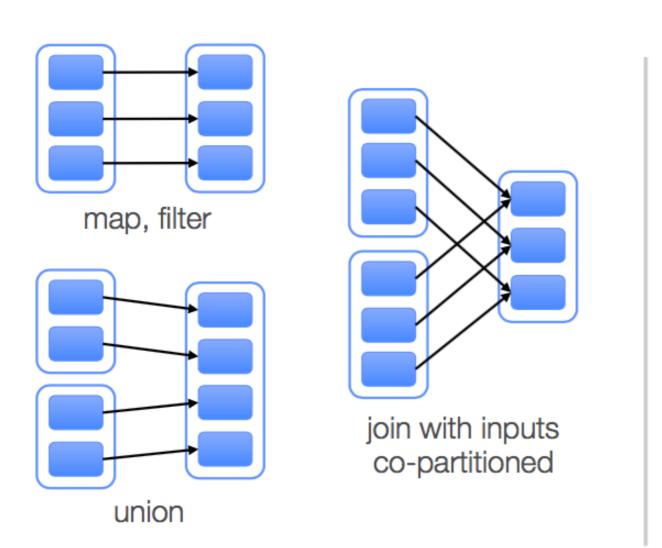
FilteredRDD func = _.contains

Task-level view

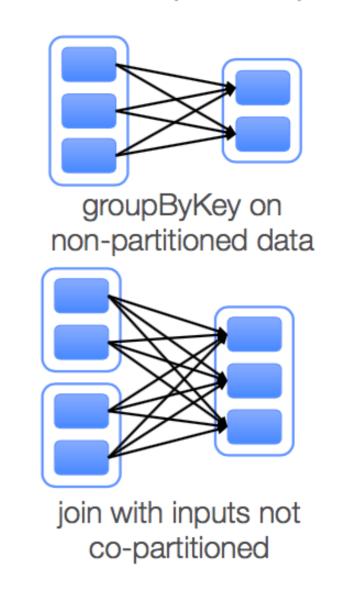


/RDD's dependencies and stage cutting

"Narrow" (pipeline-able)

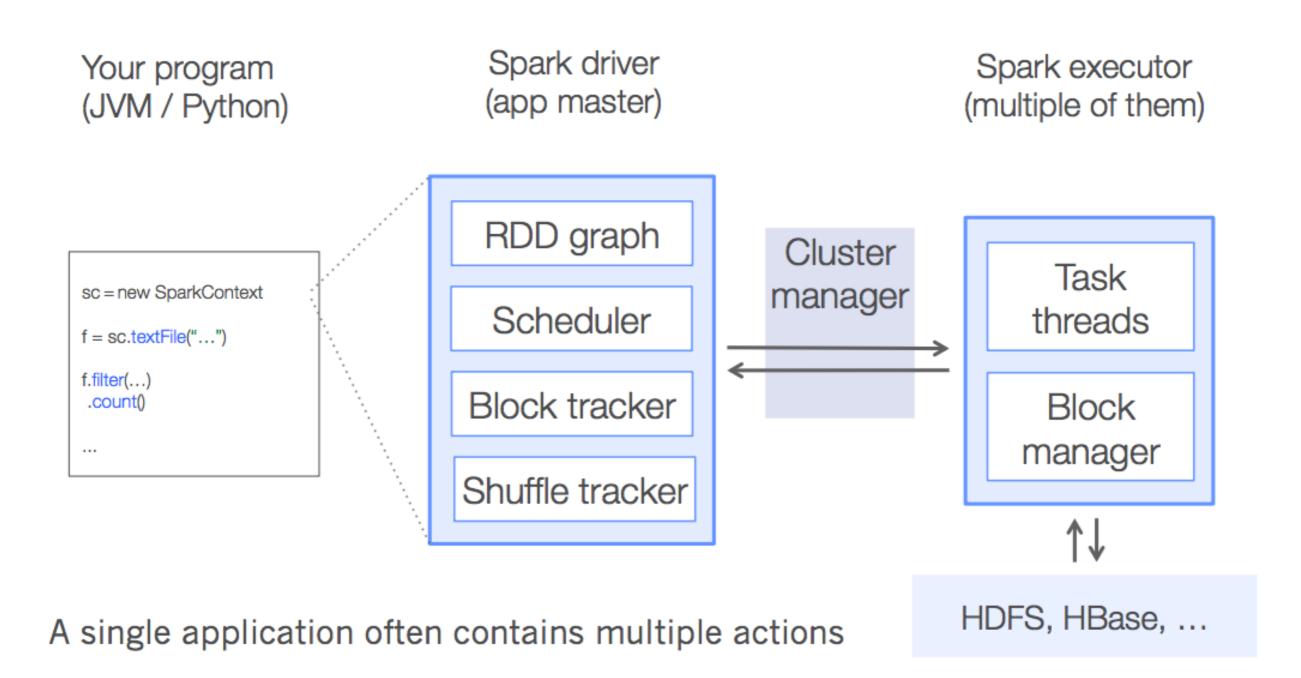


"Wide" (shuffle)



Spark architecture

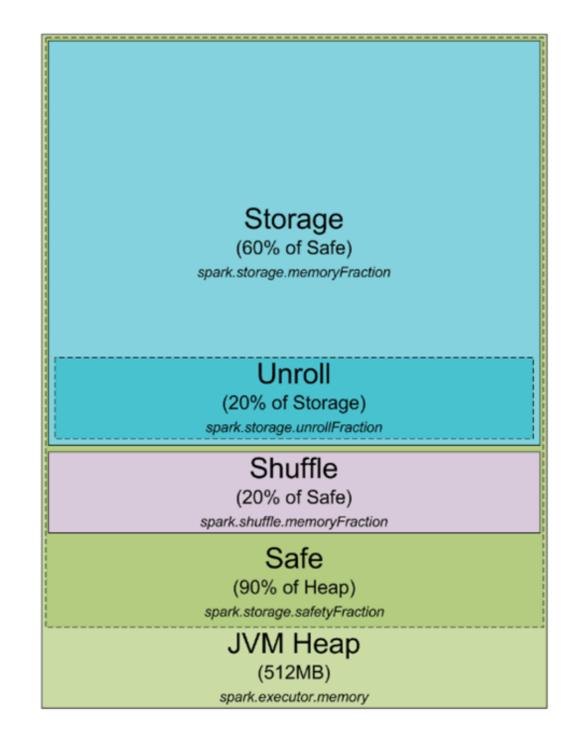
/Runtime structure



Spark architecture

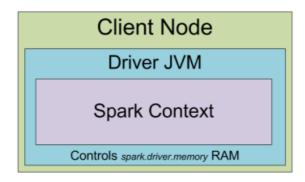
/Memory design

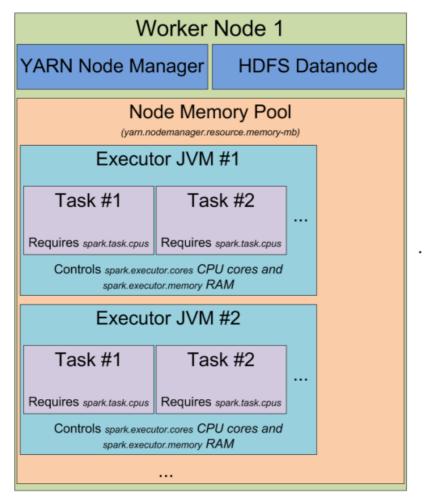
- Spark allows you to store the data both in serialized and deserialized form. The data in serialized form cannot be used directly, so you have to unroll it before using, so this is the RAM that is used for unrolling
- Spark is good in implementing the idea of efficient LRU cache. And Spark use shuffle part to store the sorted data during shuffling

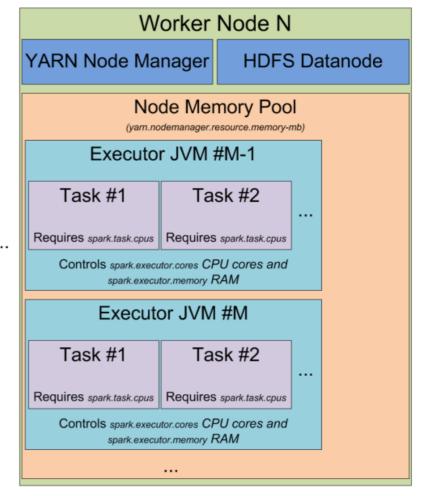


Spark architecture

/Spark on YARN







/ReduceByKey vs. groupByKey

Consider the wordcount example and data is RDD[word]

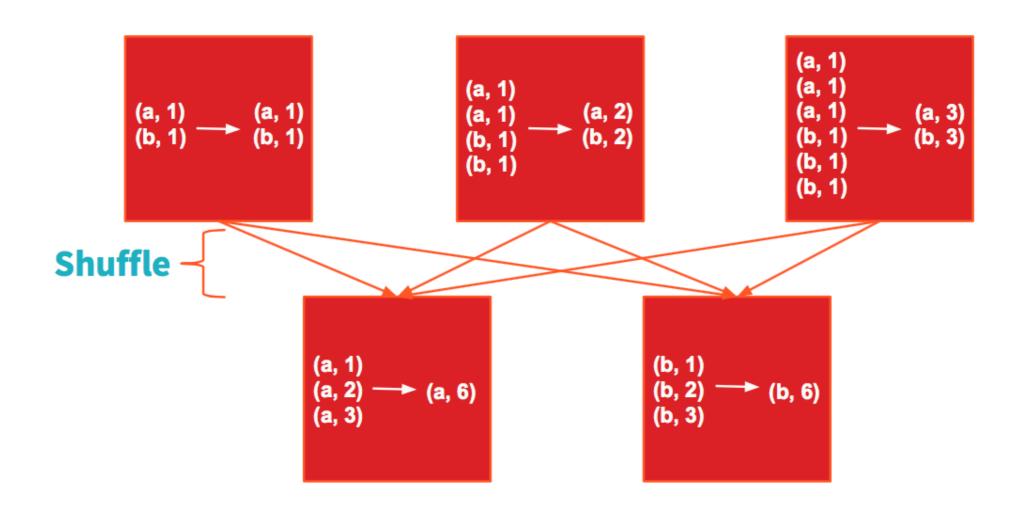
```
    Use reduceByKey // [(a, 1), (a, 2)] => [(a, 3)] data.map(x => (x, 1))
    .reduceByKey( {case(a, b) => a + b} )
```

```
    Use groupByKey // [(a, 1), (a, 2)] => [(a, [1, 2])] data.map(x => (x, 1))
        .groupByKey()
        .map( {case(a, b) => (a, b.reduce(_ + _))} )
```

 Both will give you the same answer, but reduceByKey is more efficient

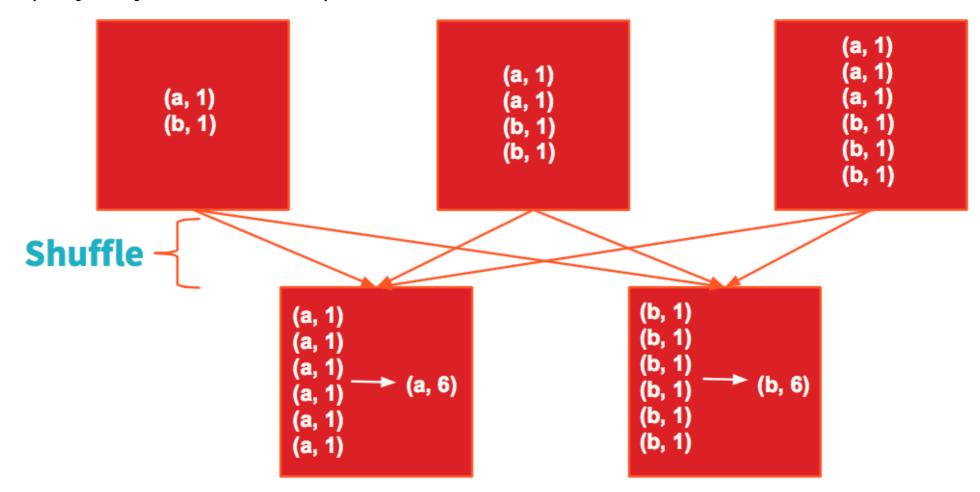
/ReduceByKey vs. groupByKey

- Both reduceByKey and groupByKey triggers shuffle of data.
- ReduceByKey: shuffle step



/ReduceByKey vs. groupByKey

GroupByKey: shuffle step



 Compared to reduceByKey, all the data is wastefully sent over the network and collected on the reduce workers

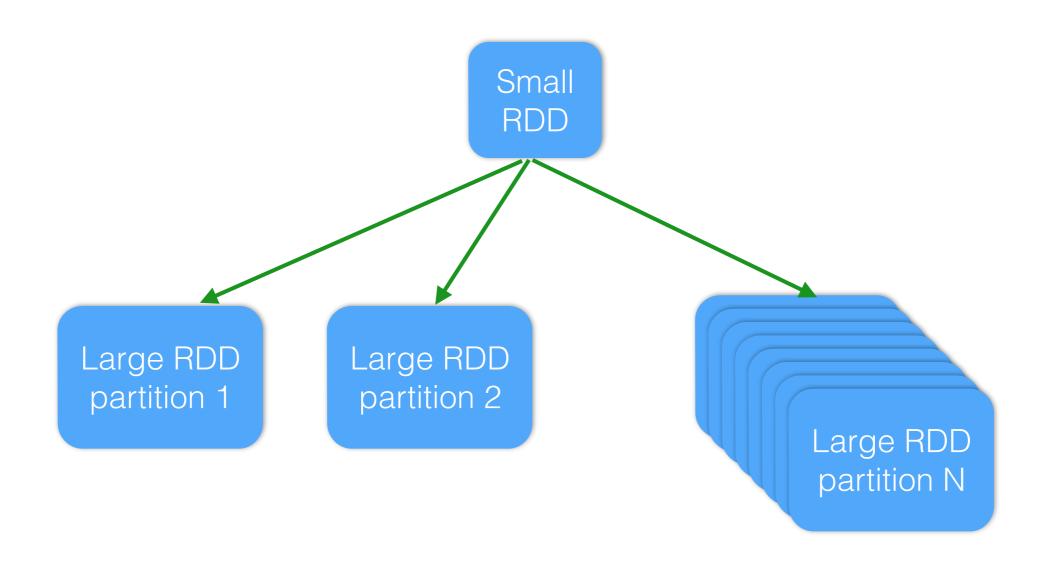
/Join

Consider the following join a large table with a small table join_rdd = sqlContext.sql("select
 FROM people_in_China
 JOIN ShangHai
 ON people_in_China.province = ShangHai.name")

- If we use ShuffledHashJoin, all the data for China will be shuffled into only XX keys (one key for each province)
 - Uneven sharding
 - Limited parallelism with XX partitions

/Join

 Solution: BroadcastHashJoin; Broadcast the small RDD to all worker nodes



Tips for Spark coding /Join

- How to configure BroadcastHashJoin
 - Set spark.sql.autoBroadcastJoinThreshold
 - sqlContext.sql("ANALYZE TABLE result_table COMPUTE STATISTICS noscan")
- Use RDD.toDebugString() or EXPLAIN to double check

Testing Spark

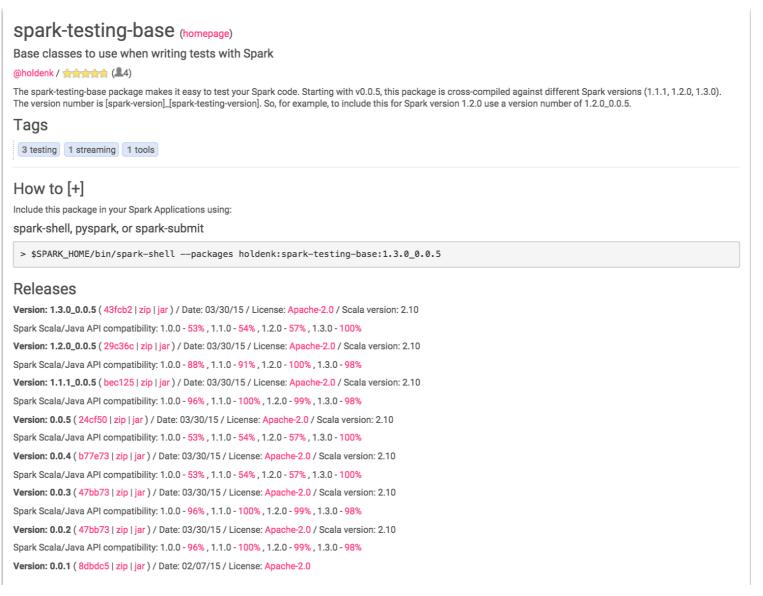
/Unit-tests of functions

```
instead of:
val splitLines = inFile.map(line => {
 val reader = new CSVReader(new StringReader(line))
 reader.readNext()
write:
def parseLine(line: String): Array[Double] = {
 val reader = new CSVReader(new StringReader(line))
 reader.readNext().map(_.toDouble)
then we can:
test("should parse a csv line with numbers") {
  MoreTestableLoadCsvExample.parseLine("1,2") should equal
(Array[Double](1.0, 2.0))
```

Testing Spark

/Testing with RDDs

Include http://spark-packages.org/package/holdenk/spark-testing-base



Case studies

/Spark at OOYALA

https://github.com/ooyala/spark-jobserver

- Company vision for Spark is as a multi-team big data service
- Shares Spark RDDs in one SparkContext among multiple jobs
- REST server for submitting, running, managing Spark jobs and contexts