# 03 - Core Spark

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### Spark Languages



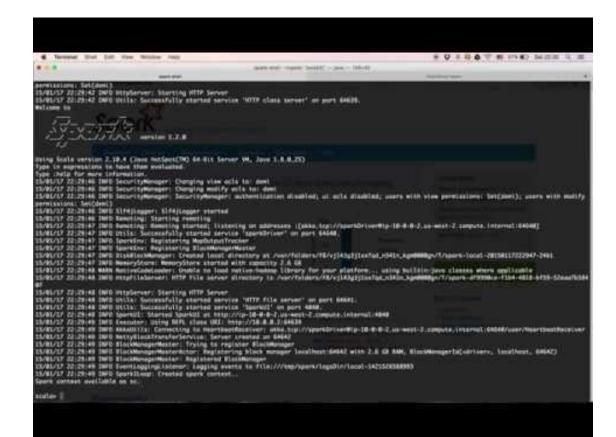








### Spark Word Count example - Spark Shell



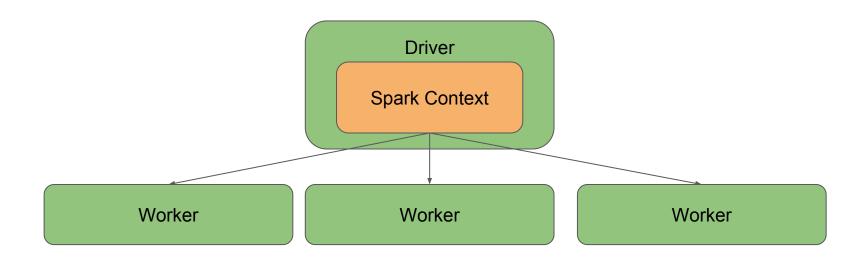


#### Module Overview

- RDD
- Transformations
- Actions



### **Spark Mechanics**





### Spark Mechanics

- Task Creator
  - Builds execution graph to be sent to each worker
- Scheduler
  - Scheduling all of the task across the nodes
- Data locality
  - Sending the work to the data to avoid moving data across the cluster
- Fault tolerance
  - Monitoring the tasks for any failures so it can trigger the task on a different node



### Spark Application Configuration

Priority and Hierarchy (From most final to the most general):

- 1) Code
- 2) spark-submit --Flags
- app.properties file ([app])
- 4) defaults (spark-default.sh)



### spark-submit example

```
$SPARK_HOME/bin/spark-submit --class org.apache.spark.examples.SparkPi \
    --master yarn-cluster \
    --num-executors 10 \
    --executor-cores 2 \
    spark-examples-1.6.0-hadoop2.6.0.jar \
    100
```



#### RDD - Resilient Distributed Dataset

- ... Collection of elements partitioned across the nodes of the cluster that can be operated on it in parallel...
  - http://spark.apache.org/docs/latest/programming-guide.html#overview
- RDD Resilient Distributed Dataset
  - Collection similar to a List / Array (Abstraction)
  - It's actually an Interface (Behind the scenes it's distributed over the cluster)
- DAG Directed Acyclic Graph
- Are Immutable!!!

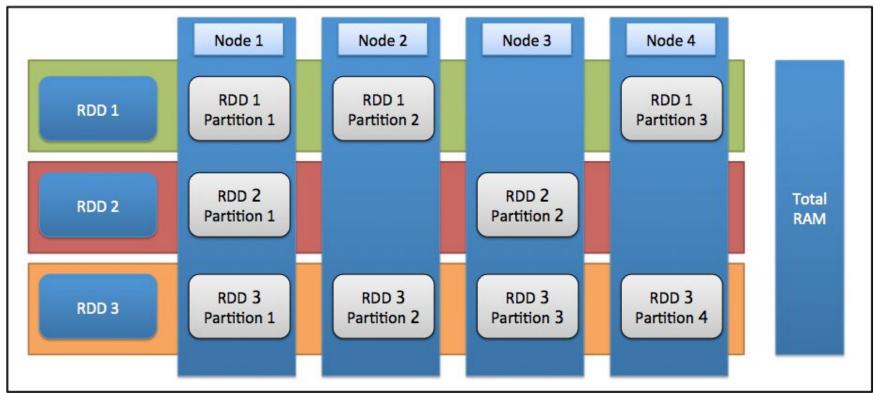


#### RDD - Resilient Distributed Dataset

- Transformations are Lazy evaluated
  - o map
  - filter
  - 0 ....
- Actions Triggers DAG computation
  - collect
  - o count
  - o reduce

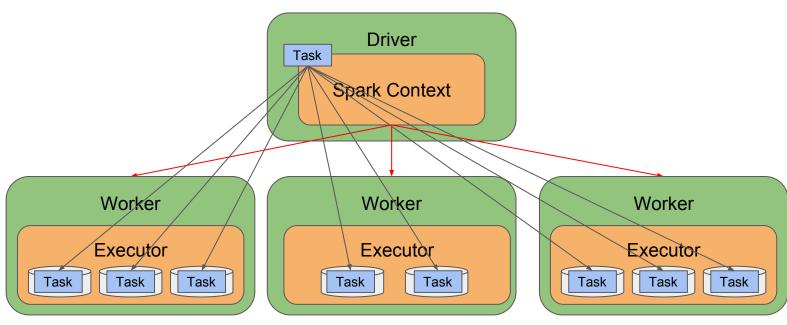


### What's really an RDD???





### Spark Mechanics





### Lambdas - Anonymous Functions

- How It's done in Java
- Why not to use it :)
  - Not testable
  - Results in a very verbose main class (inner classes)



# Input



### Input methods

- Local FileSystem
- HDFS
- Cassandra
- Avro
- Parquet



### Input methods

- sc.parallelize(<Collection>, <number of partitions>)
- sc.textFile(<path>)
- sc.sequenceFile()
  - Hadoop format
- sc.newAPIHadoopFile()
  - o instead of partitioning it accepts Hadoop configuration



### **Transformations**

Thanks to <u>Jeffrey Thompson</u> for many of the schemas <a href="https://github.com/jkthompson/pyspark-pictures">https://github.com/jkthompson/pyspark-pictures</a>

Official Documentation

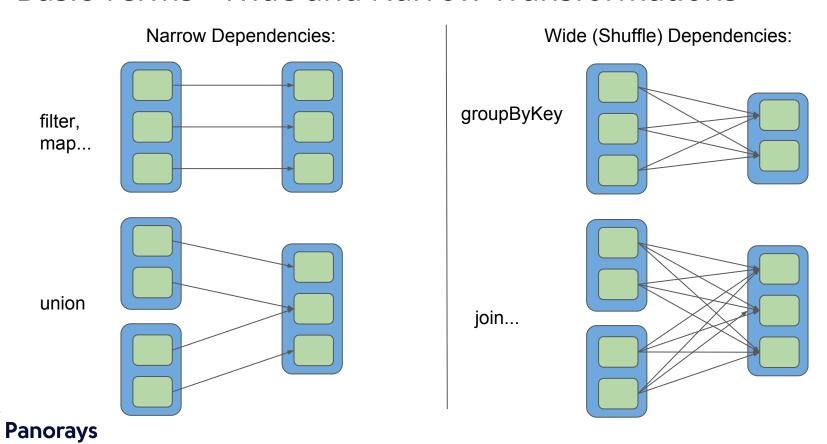


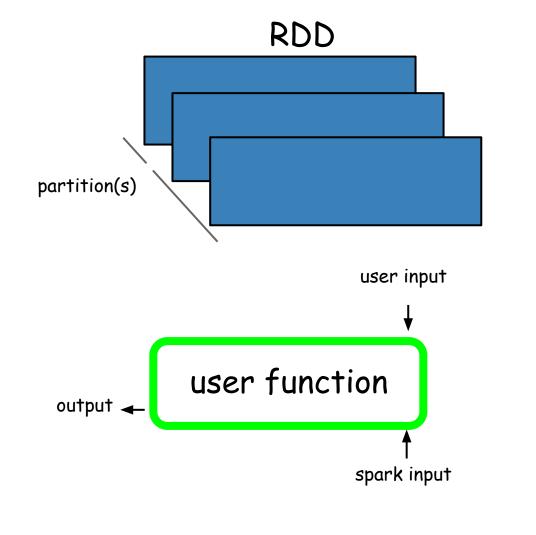
#### Wide and Narrow Transformations

- Narrow dependency: each partition of the parent RDD is used by at most one partition of the child RDD. This means the task can be executed locally and we don't have to shuffle. (Eg: map, flatMap, Filter, sample)
- **Wide dependency**: multiple child partitions may depend on one partition of the parent RDD. <u>This means we have to shuffle data</u> unless the parents are hash-partitioned (Eg: sortByKey, reduceByKey, groupByKey, cogroupByKey, join, cartesian)
- You can read a good blog post about it.



#### Basic Terms - Wide and Narrow Transformations





#### RDD Elements

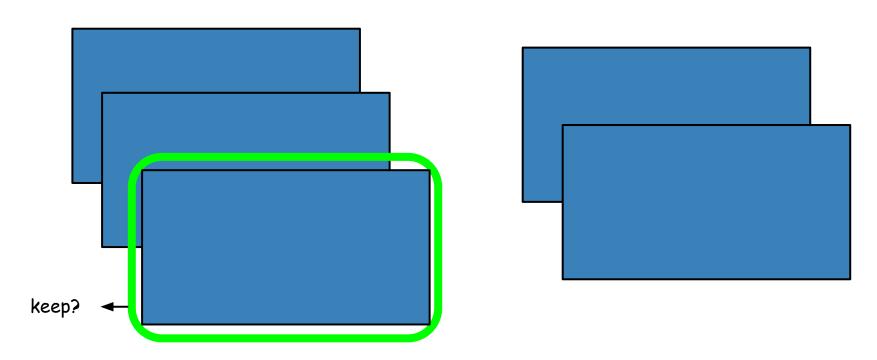
original

transformed value

transformed type

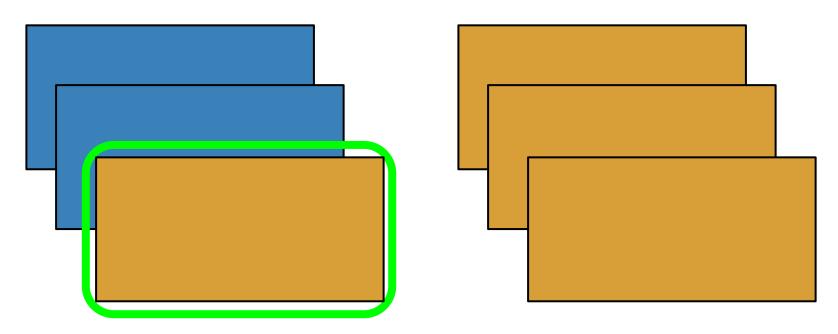
object on driver

### filter



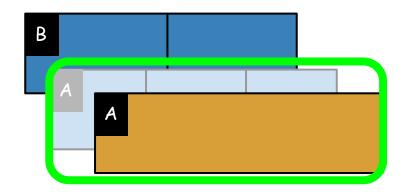


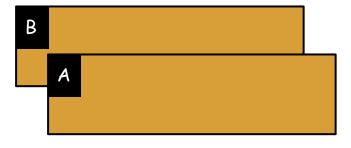
### map





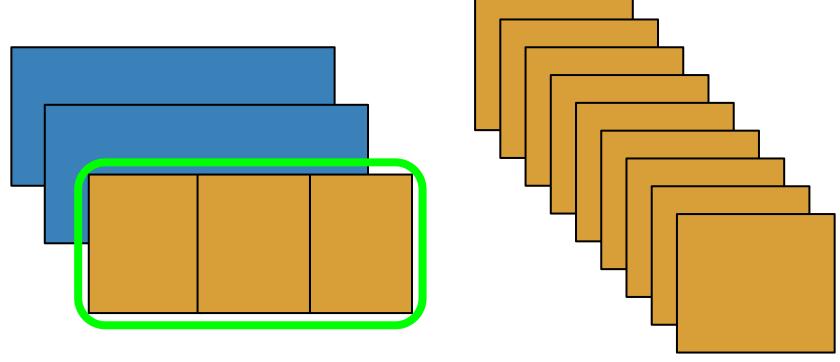
# mapValues





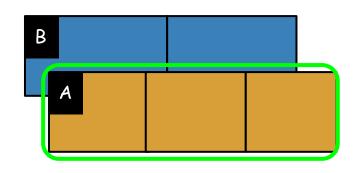


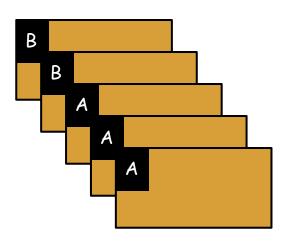
# flatMap





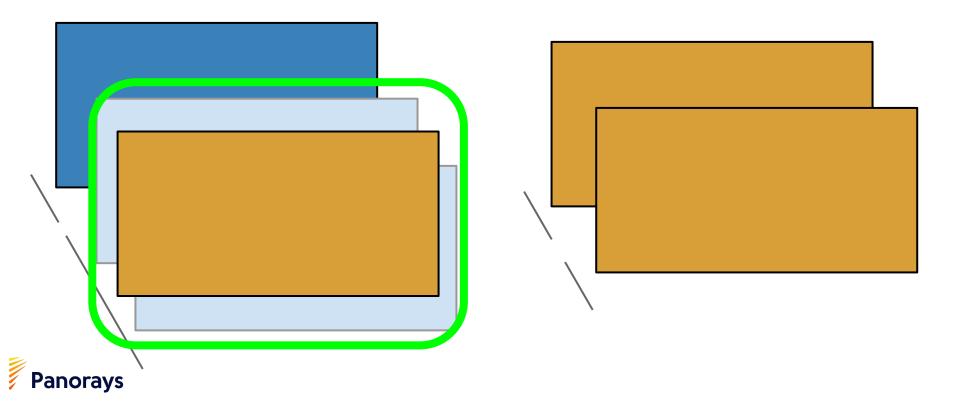
## flatMapValue



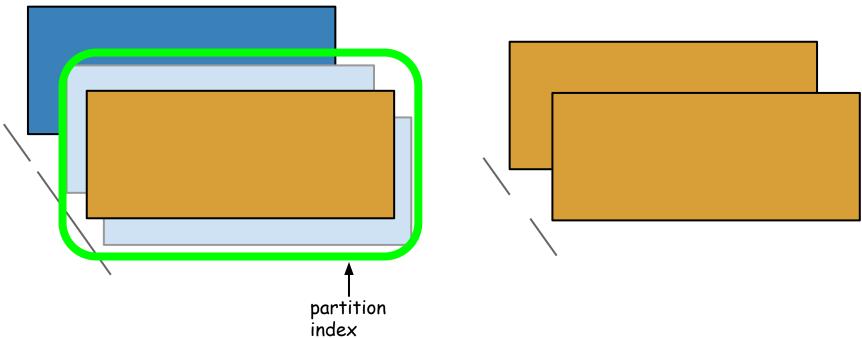




# mapPartitions

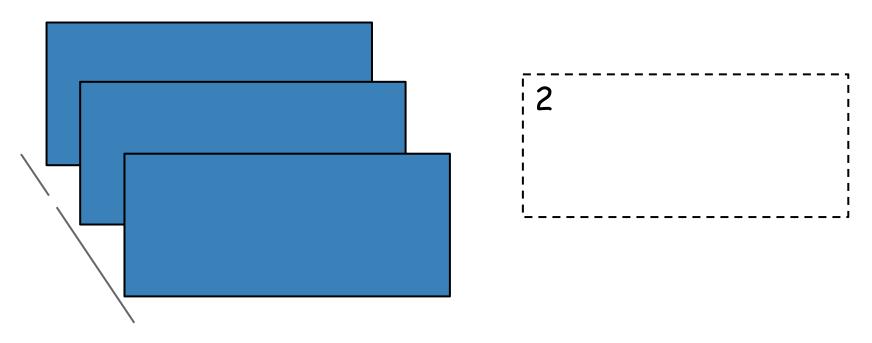


### mapPartitionsWithIndex



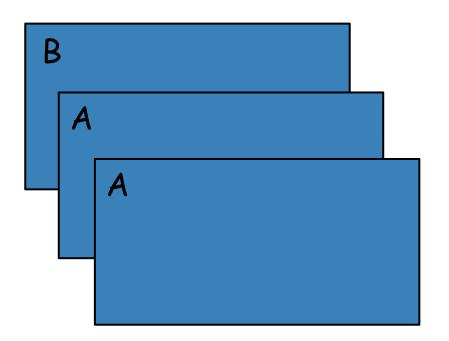


### getNumPartitions



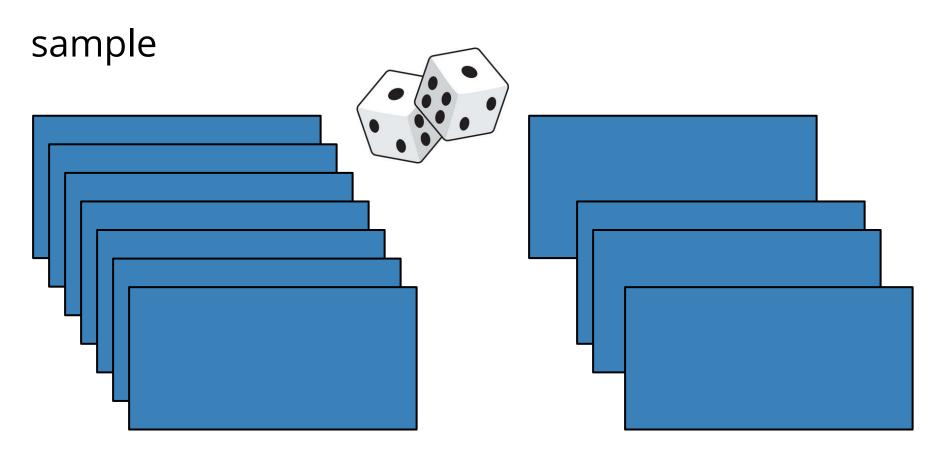


### distinct





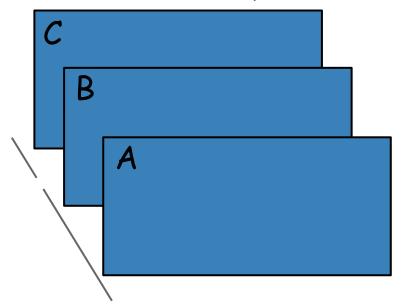


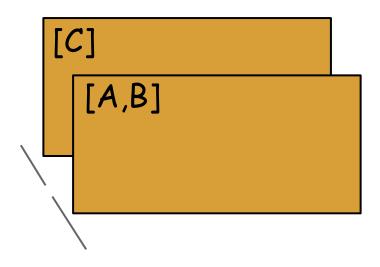




### glom

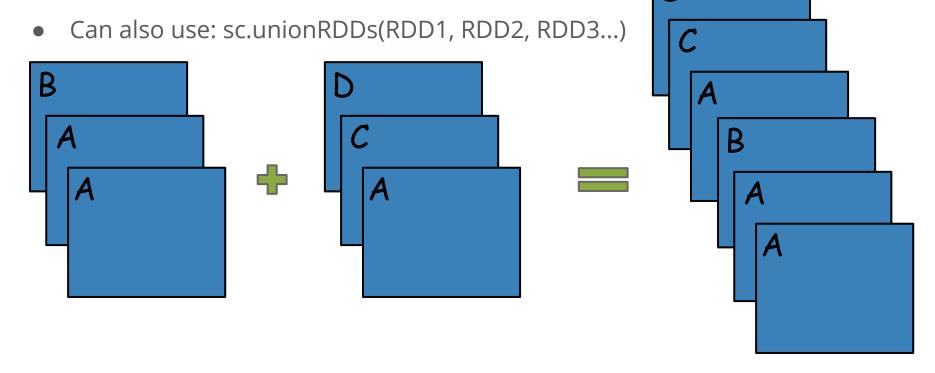
• I have never used it:)





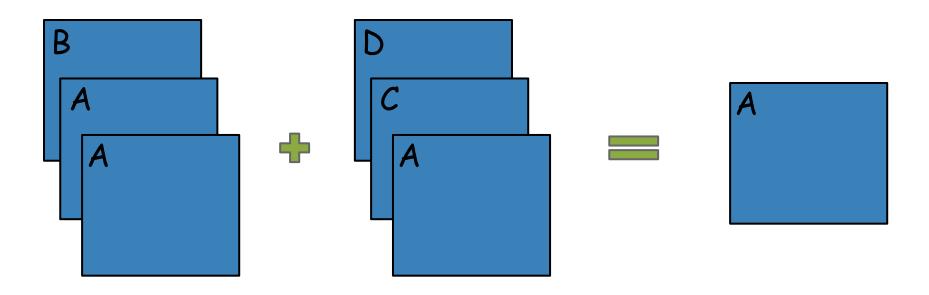


### Union (RDD1.union(RDD2))





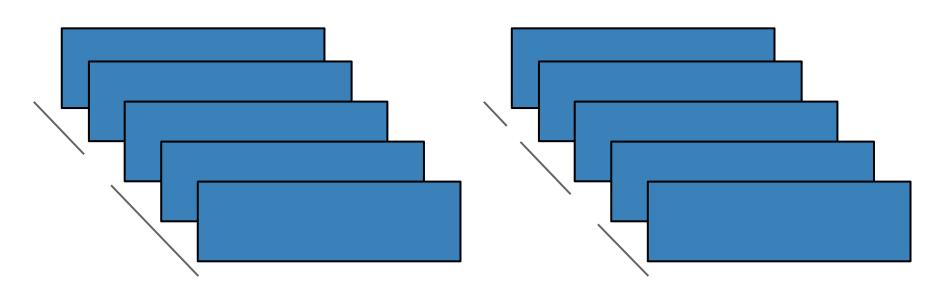
### Intersection (RDD1.intersection(RDD2))





## repartition

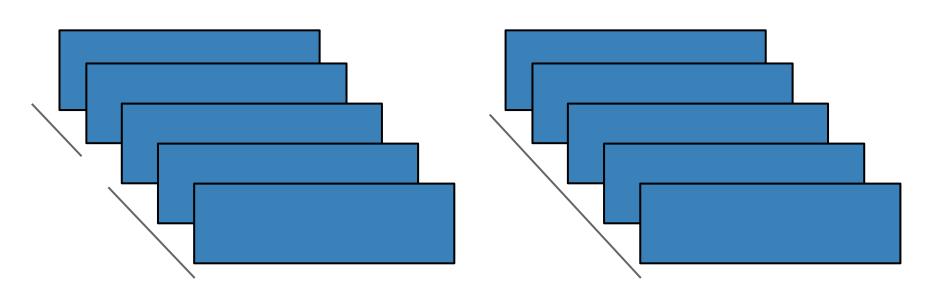
numPartitions = 3





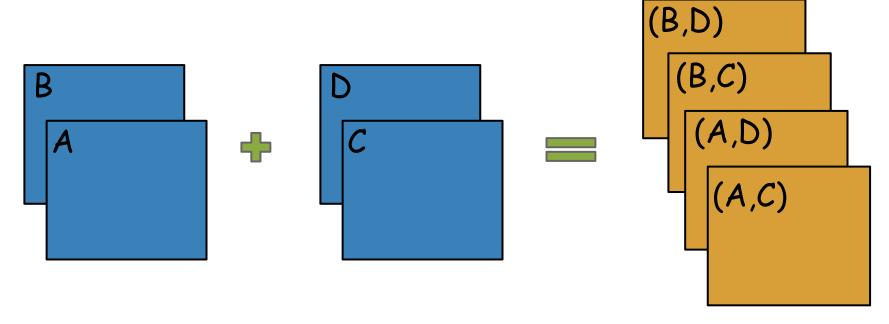
### coalesce

numPartitions = 1



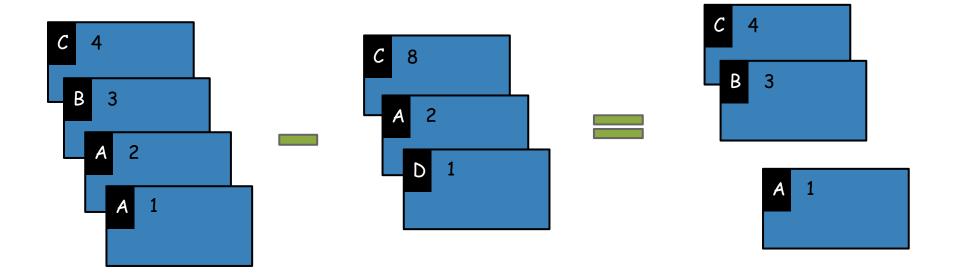


#### cartesian



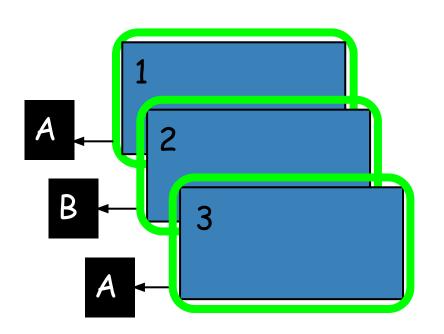


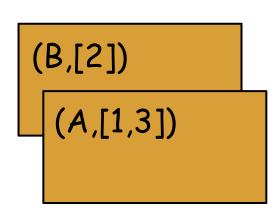
### substruct(RDD1.substruct(RDD2))





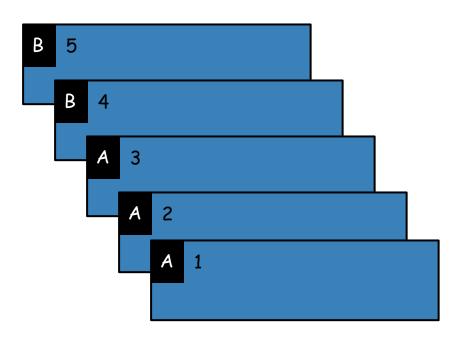
### groupBy

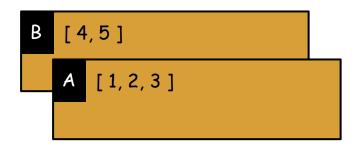






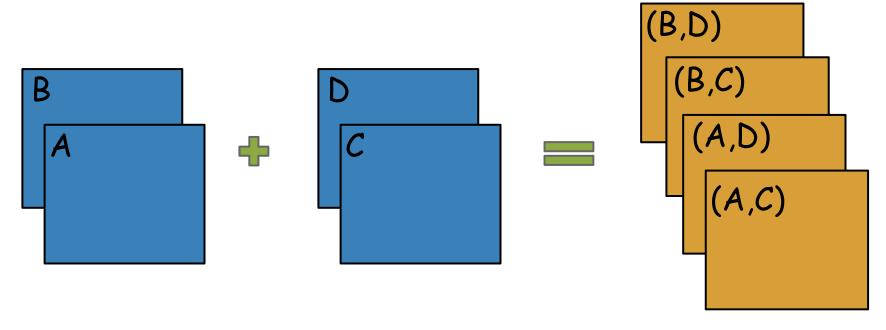
## groupByKey





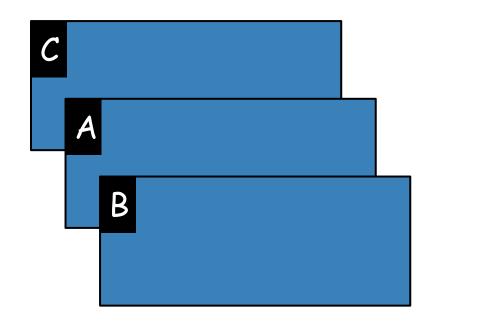


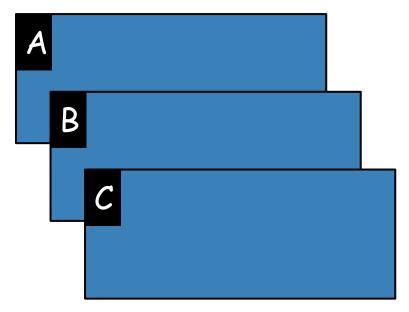
### cartesian (RDD1.cartesian(RDD2))





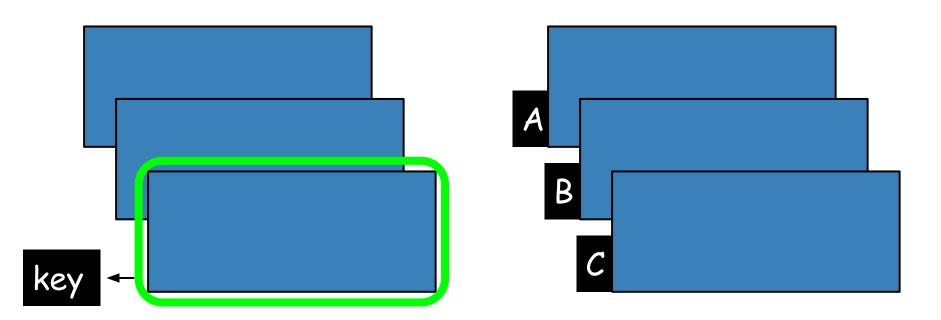
# sortByKey





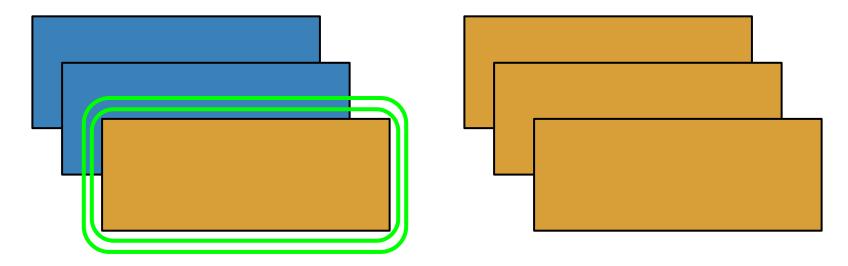


# sortBy





### pipe

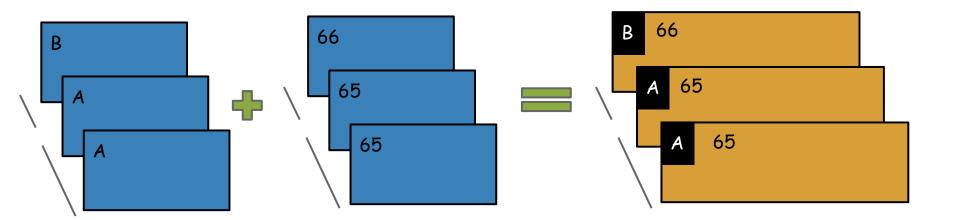


#### external command

- Pipe each partition of the RDD through a shell command, e.g. a Perl or bash script. RDD elements are written to the process's stdin and lines output to its stdout are returned as an RDD of strings.
- (I've never used it)



### zip (RDD1.zip(RDD2))

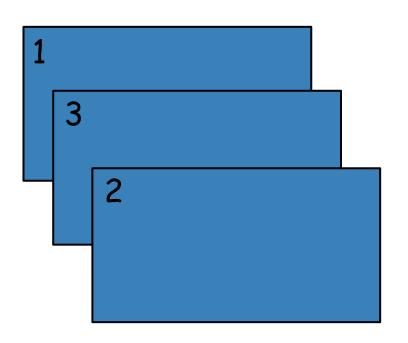




# **Actions**



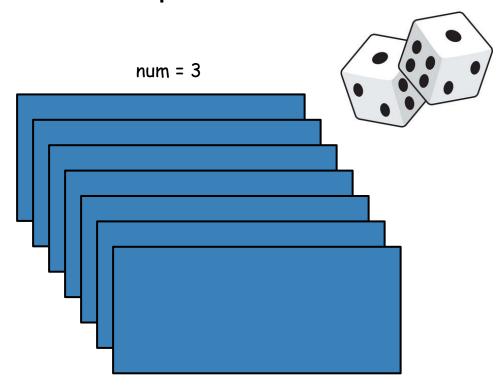
#### count

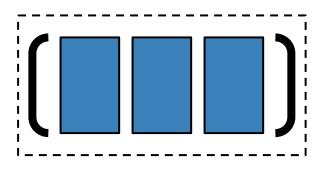






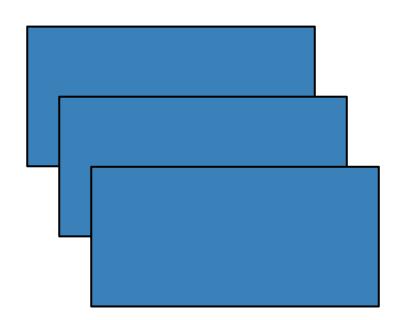
# takeSample

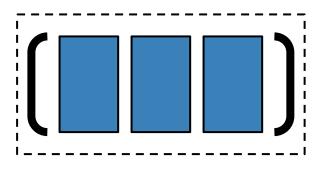






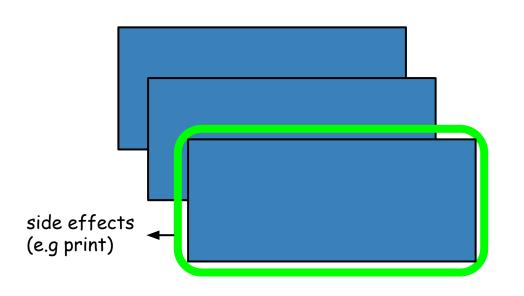
### Collect







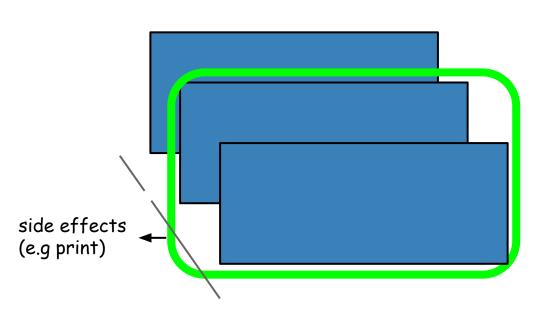
#### foreach



\*no return value, original RDD unchanged



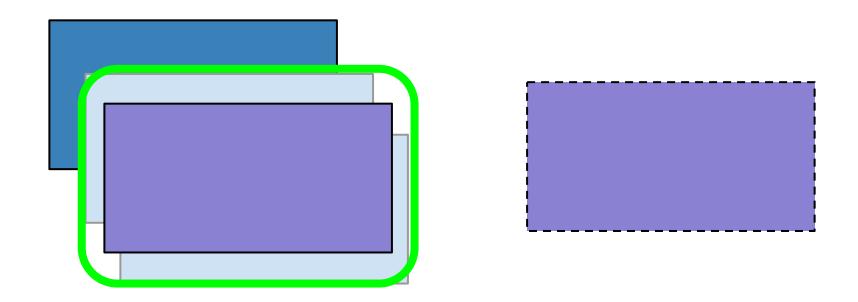
#### foreachPartition



\*no return value, original RDD unchanged



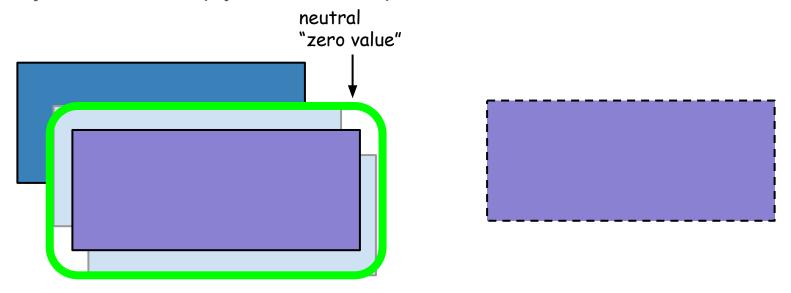
### reduce





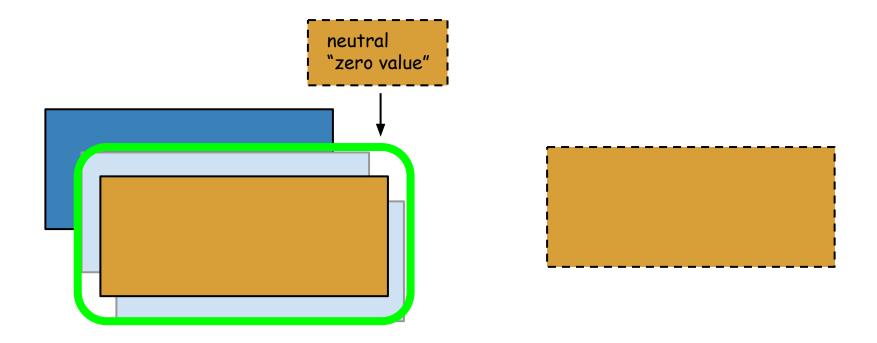
### fold

If you have an empty RDD, it's a replacement to reduce.



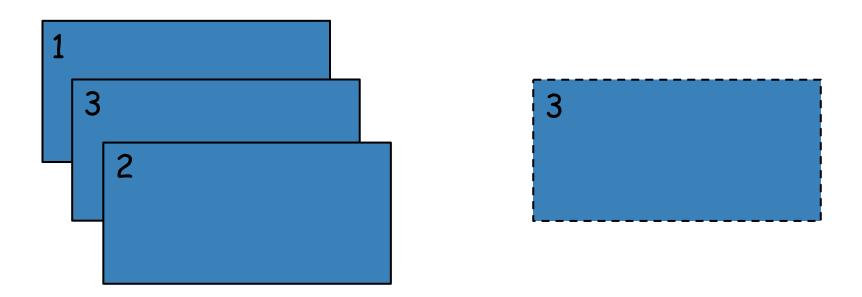


### aggregate()



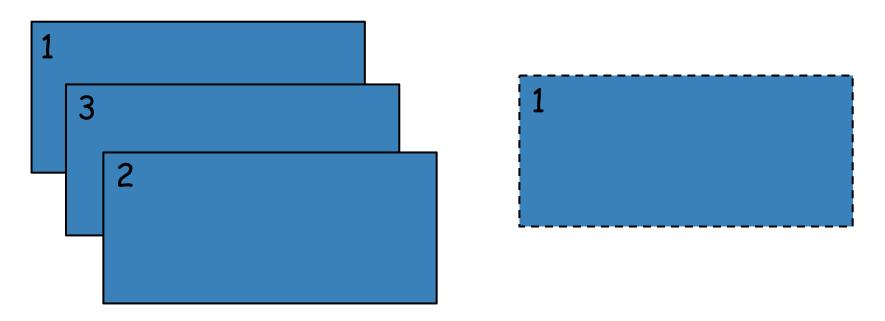


#### max



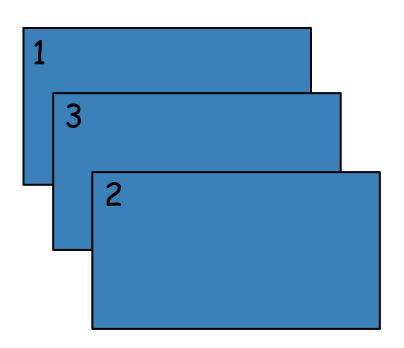


### min





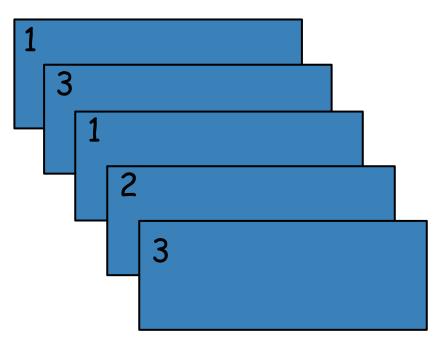
#### sum

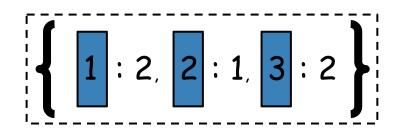






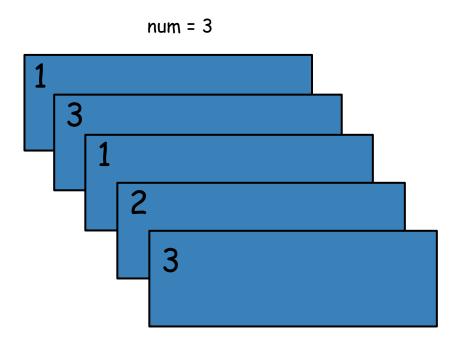
### countByValue()

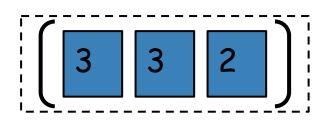






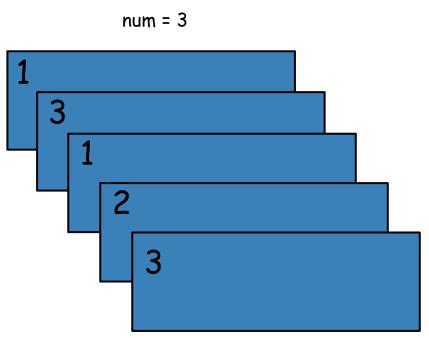
## top

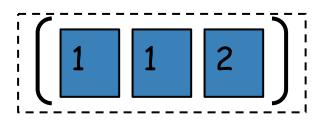






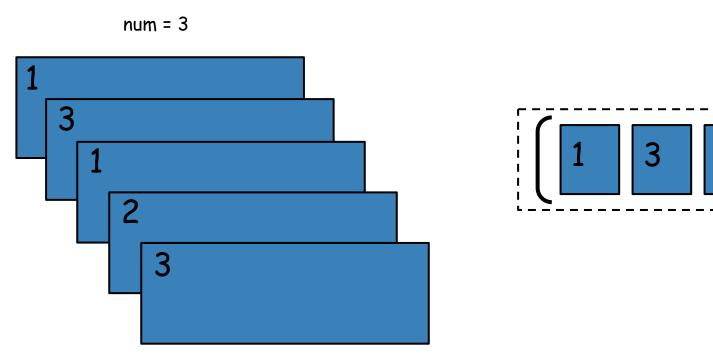
# topOrdered





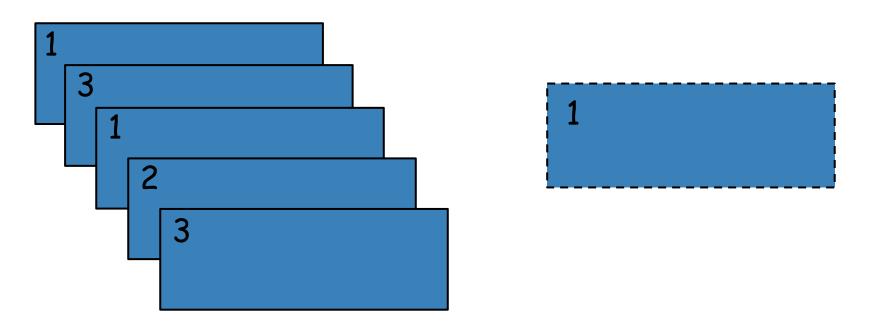


### take





### first





#### More Actions

- takeSample()
- mean()
- variance()
- tsdev()
- histogram()
- sampleStdev()
- sampleVariance()

Again, to tell you the truth, I've never used these.



### Persistence



#### **Actions**

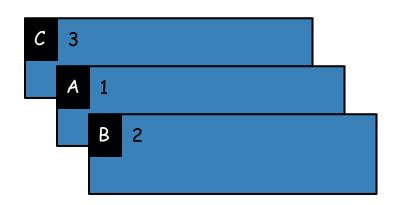
- saveAsObjectFile(path)
- saveAsTextFile(path)
- ExternalConnector
- foreach()
  - forEachPartition()

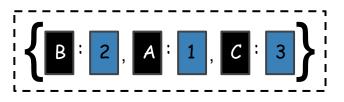


# Key Value Methods



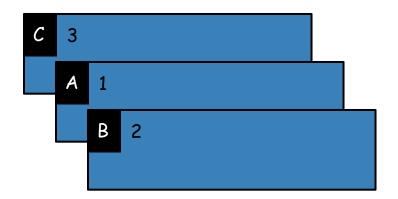
### collectAsMap







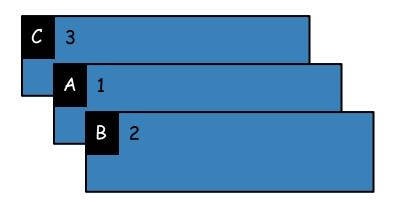
# keys

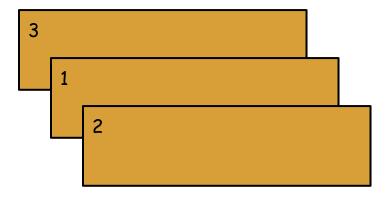






### values

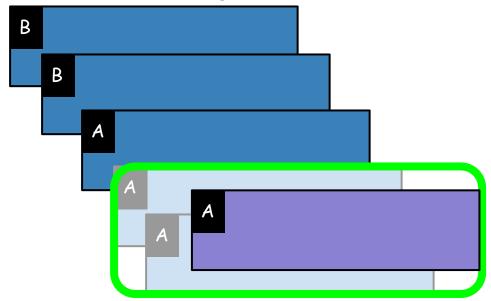






### reduceByKey

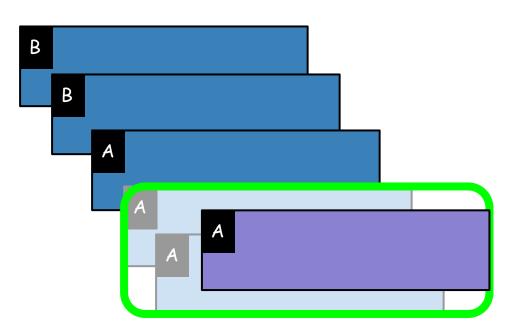
Occurs locally

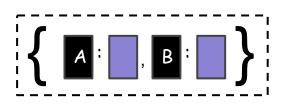






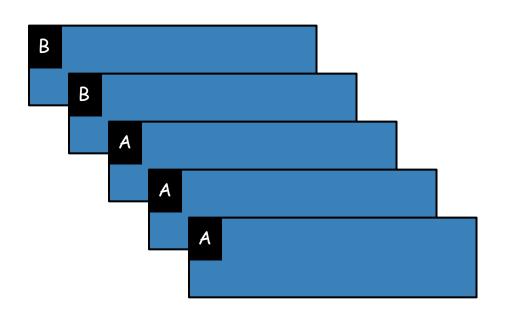
### reduceByKeyLocally

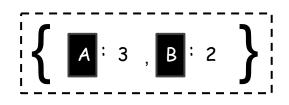






### countByKey



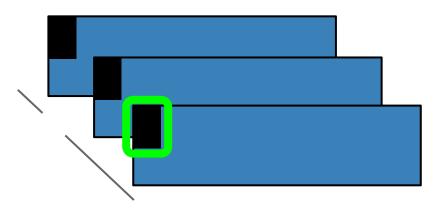


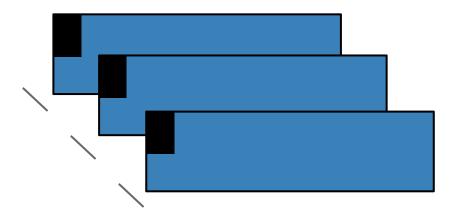


## partitionBy

```
new partition = key % numPartitions index
```

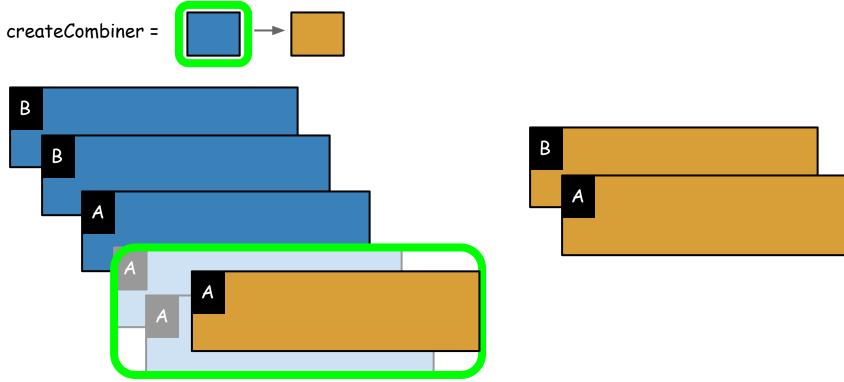
numPartitions = 3





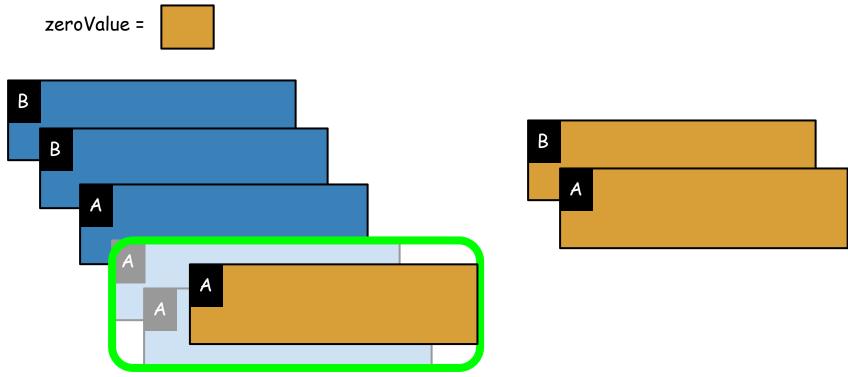


## combineBy



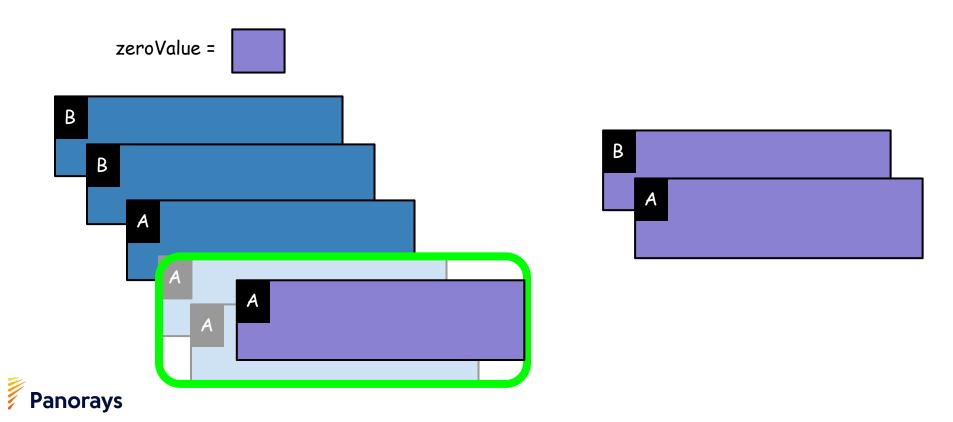


## aggregateByKey





# foldByKey



## Regular Methods

- collectAsMap
- mapValues
- reduceByKey
  - Occurs locally
- foldByKey
- aggregateByKey
- combineByKey

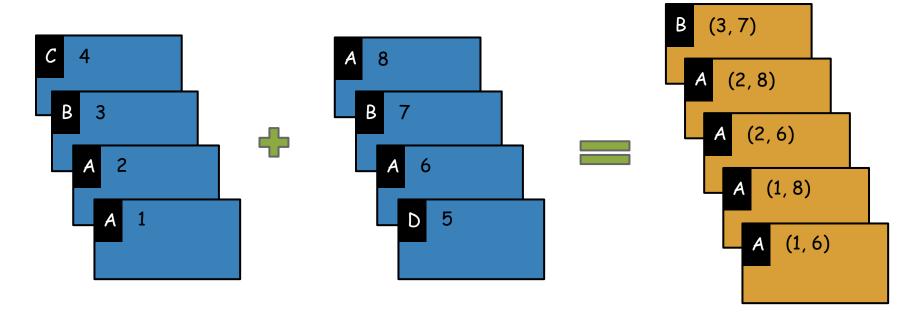
- groupByKey
- countByKey
- sampleByKey
- substractByKey
- sortByKey



# SQL-like Pairings

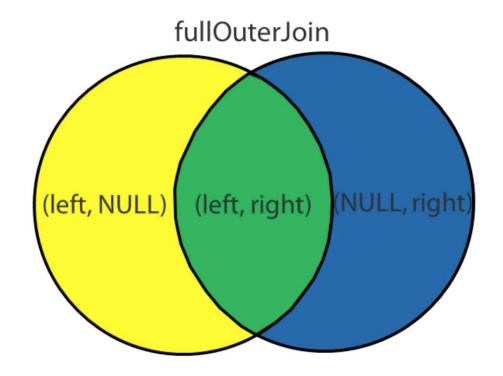


## join



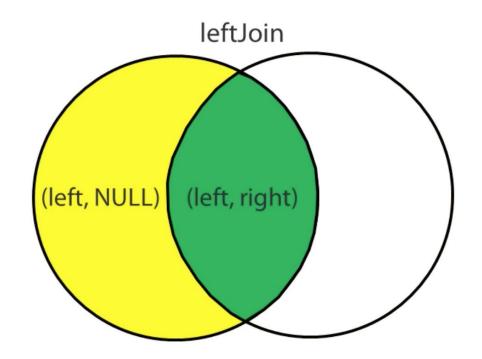


## fullOuterJoin



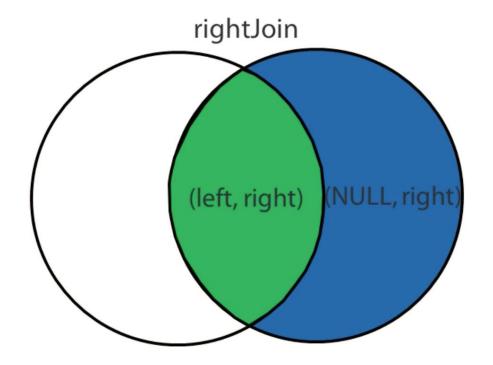


## leftJoin





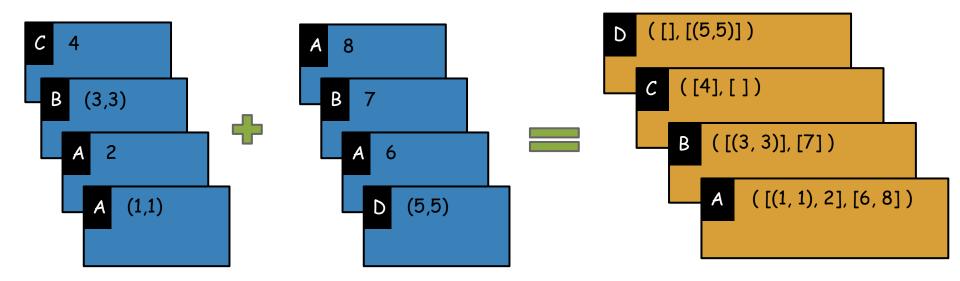
## rightJoin





## cogroup

When called on datasets of type (K, V) and (K, W), returns a dataset of (K, (Iterable<V>, Iterable<W>)) tuples. This operation is also called groupWith.





### List of Transformations and Actions

· sample

randomSplit



### General

### Math / Statistical

### Set Theory / Relational

### Data Structure / I/O

- map
- · filter
- · flatMap
- · mapPartitions
- mapPartitionsWithIndex
- groupBy
- · sortBy

- union
  - intersection
  - · subtract
  - · distinct
  - cartesian
  - · zip

- keyBy
- · zipWithIndex
- · zipWithUniqueID
- · zipPartitions
- · coalesce
- · repartition
- repartitionAndSortWithinPartitions
- · pipe





- · collect
- · aggregate
- · fold
- · first
- · take
- forEach
- · top
- treeAggregate
- treeReduce collectAsMap
- · forEachPartition

- count
- takeSample
- max
- · min
- Sum
- histogram
- nean
- variance
- stdev
- sampleVariance
- countApprox
- countApproxDistinct

takeOrdered

- saveAsTextFile
- saveAsSequenceFile
- saveAsObjectFile
- saveAsHadoopDataset
- saveAsHadoopFile
- saveAsNewAPIHadoopDataset
- saveAsNewAPIHadoopFile

# Caching Data



### Cache

- Cache / Persist
  - org.apache.spark.storage.StorageLevel.MEMORY\_ONLY
- persist(newLevel: StorageLevel)
  - MEMORY ONLY
  - MEMORY\_AND\_DISK
  - DISK ONLY
  - MEMORY ONLY SER
  - MEMORY\_AND\_DISK\_SER
  - ...\_2 (Can be replicated to another node)
  - o OFF HEAP
- unpersist(blocking: boolean = true)



## Caching pitfalls

- Always unpersist your cached RDDs, but not too soon.
- Try to cache right after wide transformations
  - o To avoid the shuffle in case of failure



## Accumulator & Broadcast



### Accumulator

val accumulator = sc.accumulator(0, "Accumulator Name")

```
rdd.foreach(x=> {
    doSomething();
    accumulator += 1;
})
```

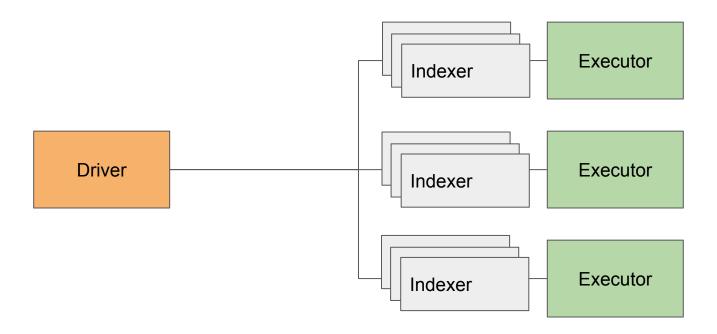
- In the driver:
  - val accumulatorValue = accumulator.value;



```
val indexer = Map(...)
rdd.flatMap(rddVal => indexer.get(rddVal))
```

What will happen?

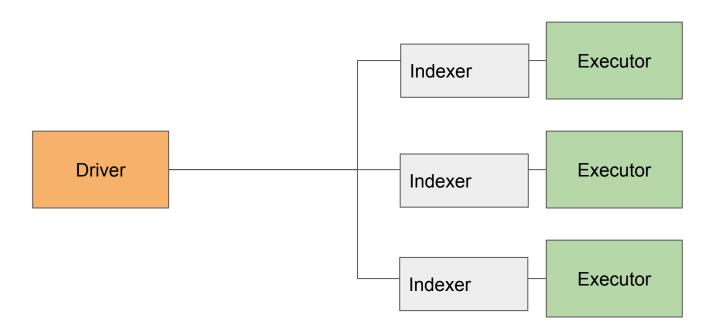






val indexer = sc.broadcast((Map(....))) // Map = 10Mb; indexer < 10Mb
rdd.flatMap(rddVal => indexer.value.get(rddVal))







## Java API

- All implemented with:
  - o Org.apache.spark.api.java.function
- JavaPairRDD
  - mapToPair
- Almost all of the API is pretty identical, all of the changes can be found in the documentation



### Resources

- RDD Research Paper
  - https://www.cs.berkeley.edu/~matei/papers/2012/nsdi\_spark.pdf
- Lambdas
  - https://docs.oracle.com/javase/tutorial/java/javaOO/lambdaexpressions.html
- Official Documentation
  - http://spark.apache.org/docs/latest/programming-guide.html







# Mind the Attack Surface