

#### A Deeper Understanding of Spark's Internals

Aaron Davidson 07/01/2014

#### This Talk

 Goal: Understanding how Spark runs, focus on performance

- Major core components:
  - Execution Model
  - The Shuffle
  - Caching



#### This Talk

 Goal: Understanding how Spark runs, focus on performance

- Major core components:
  - Execution Model
  - The Shuffle
  - Caching



```
sc.textFile("hdfs:/names")

.map(name => (name.charAt(0), name))

.groupByKey()

.mapValues(names => names.toSet.size)

.collect()
```

```
sc.textFile("hdfs:/names")

.map(name => (name.charAt(0), name))

.groupByKey()

.mapValues(names => names.toSet.size)

.collect()
```

```
sc.textFile("hdfs:/names")
.map(name => (name.charAt(0), name))
.groupByKey()
.mapValues(names => names.toSet.size)
.collect()
Ahir Pat Andy
(A, Ahir) (P, Pat) (A, Andy)

.collect()
```

Goal: Find number of distinct names per "first letter"

.collect()

```
sc.textFile("hdfs:/names")
.map(name => (name.charAt(0), name))
.groupByKey()
.mapValues(names => names.toSet.size)
Ahir Pat Andy
(A, Ahir) (P, Pat) (A, Andy)
(A, [Ahir, Andy]) (P, [Pat])
```

Goal: Find number of distinct names per "first letter"

.collect()

```
sc.textFile("hdfs:/names")
.map(name => (name.charAt(0), name))
.groupByKey()
.mapValues(names => names.toSet.size)
Ahir Pat Andy
(A, Ahir) (P, Pat) (A, Andy)
(A, [Ahir, Andy]) (P, [Pat])
```

```
sc.textFile("hdfs:/names")
.map(name => (name.charAt(0), name))
.groupByKey()
.mapValues(names => names.toSet.size)
.collect()
Ahir Pat Andy
(A, Ahir) (P, Pat) (A, Andy)
(A, [Ahir, Andy]) (P, [Pat])
.collect()
```

Goal: Find number of distinct names per "first letter"

.collect()

```
sc.textFile("hdfs:/names")

.map(name => (name.charAt(0), name))

.groupByKey()

.mapValues(names => names.toSet.size)

Ahir Pat Andy

(A, Ahir) (P, Pat) (A, Andy)

(A, [Ahir, Andy]) (P, [Pat])

(A, [Ahir, Andy])

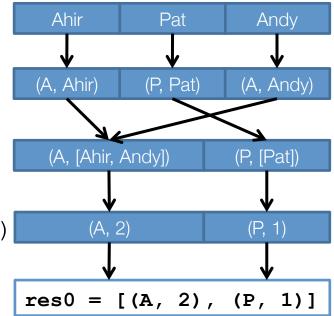
(B, [Pat])

(B, [Pat])

(C, Pat])

(C, Pat])
```

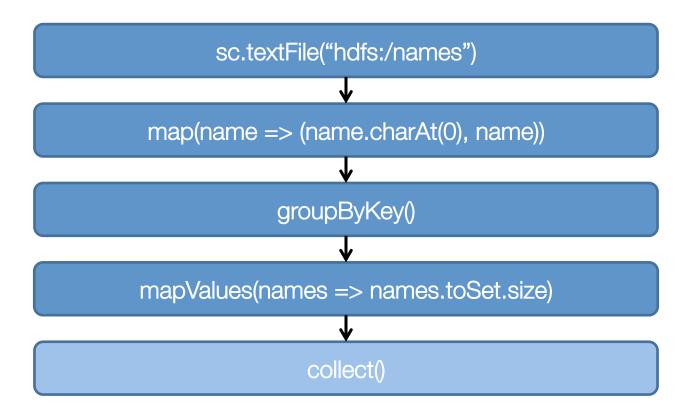
```
sc.textFile("hdfs:/names")
.map(name => (name.charAt(0), name))
.groupByKey()
.mapValues(names => names.toSet.size)
.collect()
```



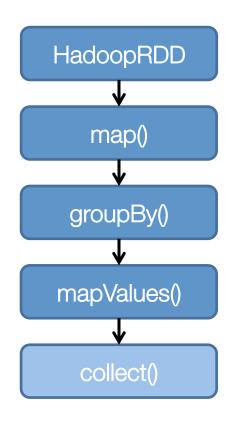
#### Spark Execution Model

- 1. Create DAG of RDDs to represent computation
- 2. Create logical execution plan for DAG
- 3. Schedule and execute individual tasks

#### Step 1: Create RDDs

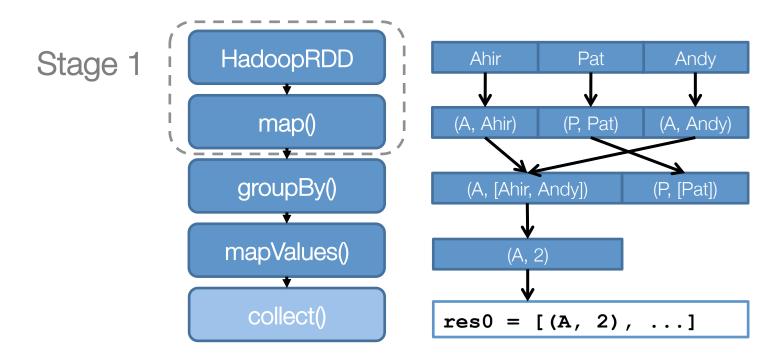


## Step 1: Create RDDs



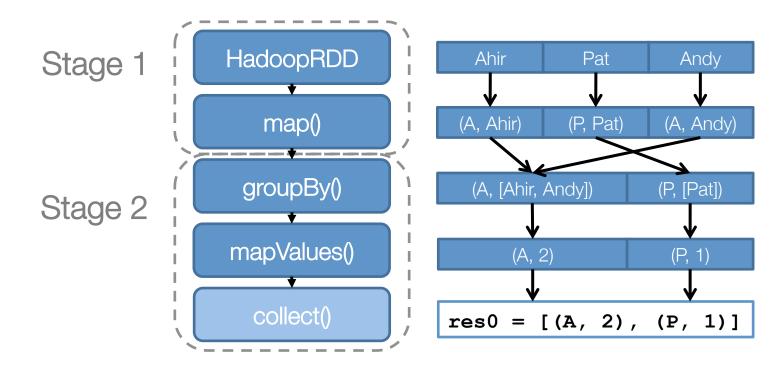
#### Step 2: Create execution plan

- Pipeline as much as possible
- Split into "stages" based on need to reorganize data

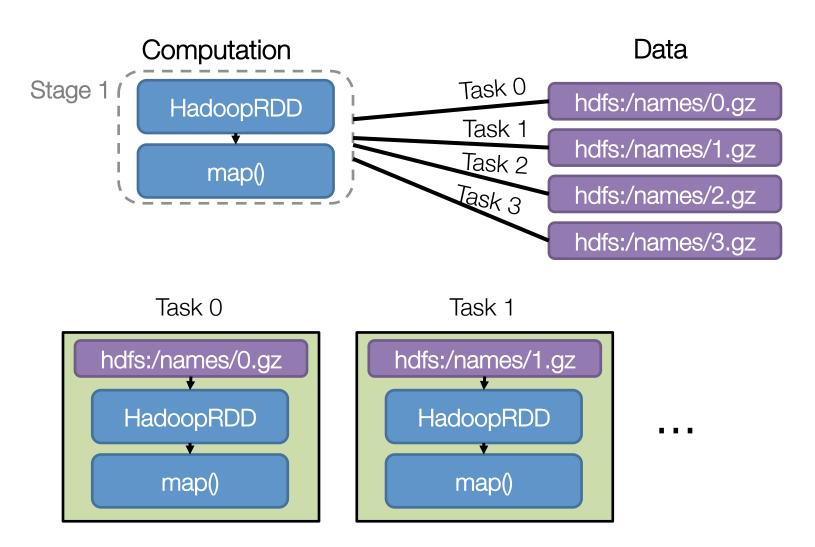


#### Step 2: Create execution plan

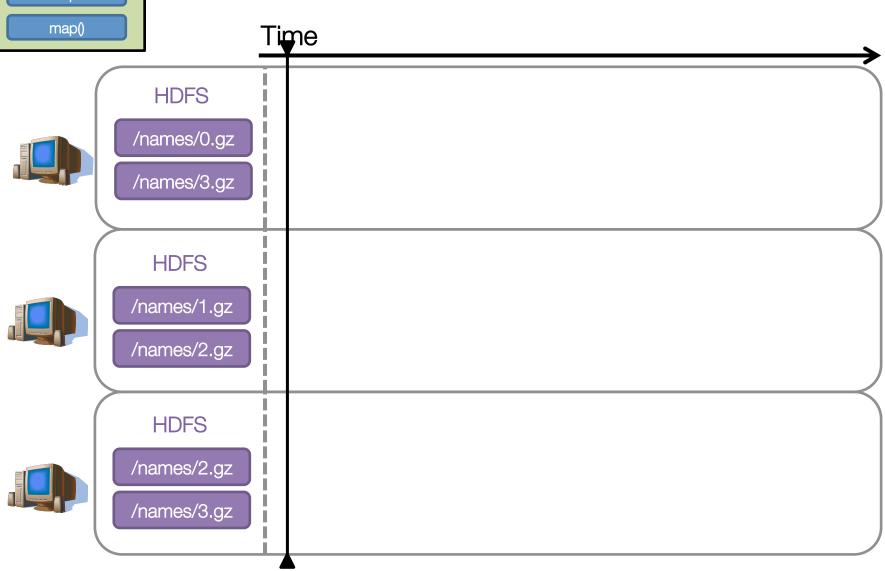
- Pipeline as much as possible
- Split into "stages" based on need to reorganize data



- Split each stage into tasks
- A task is data + computation
- Execute all tasks within a stage before moving on



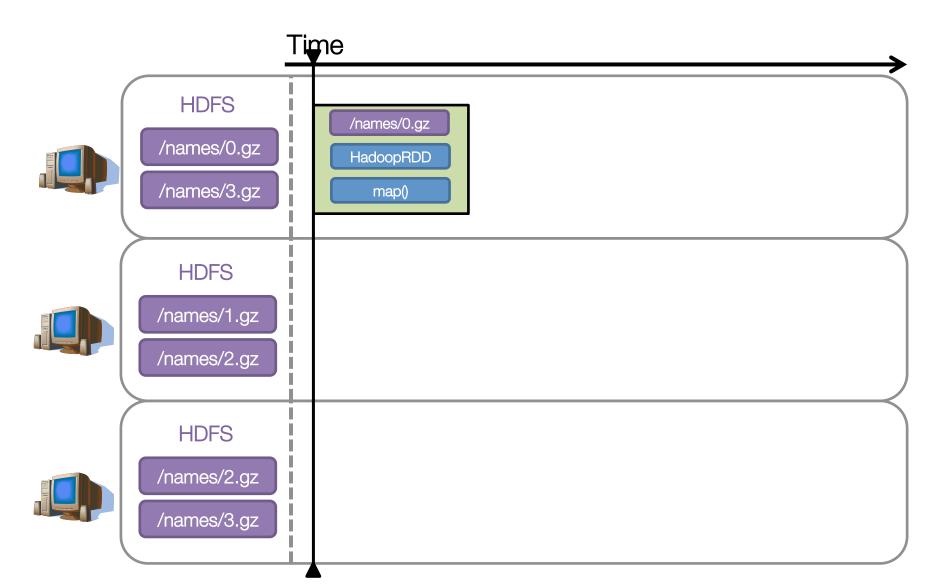
# /names/0.gz Step 3 HadoopRDD map() Time



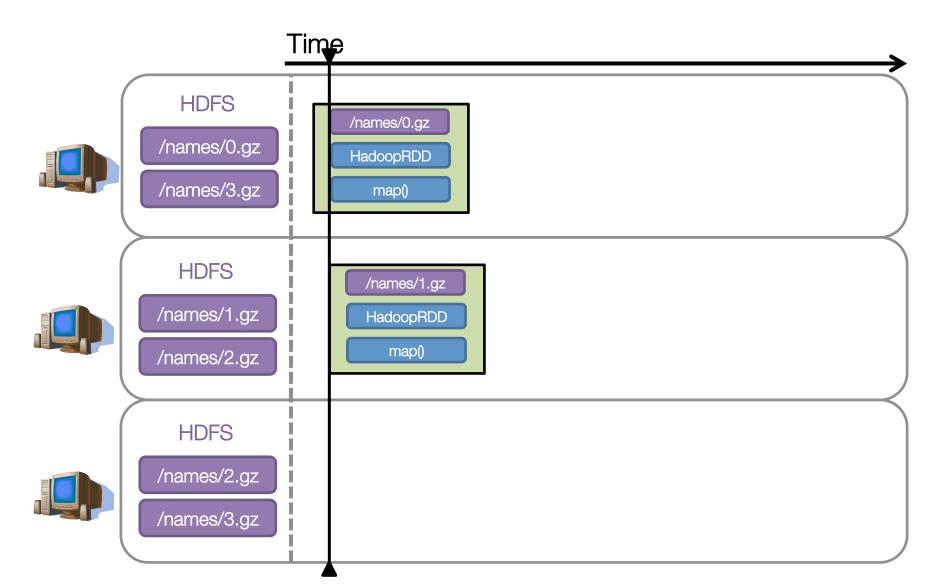
Step 3: Schedule tasks /names/0.gz HadoopRDD map() Time **HDFS** /names/0.gz /names/3.gz **HDFS** /names/1.gz /names/2.gz **HDFS** 

/names/2.gz

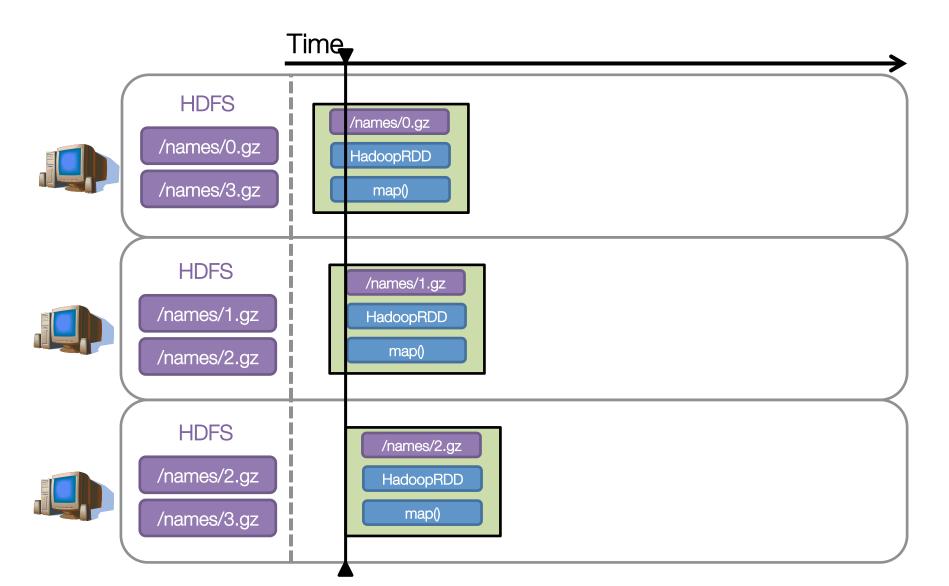
/names/3.gz



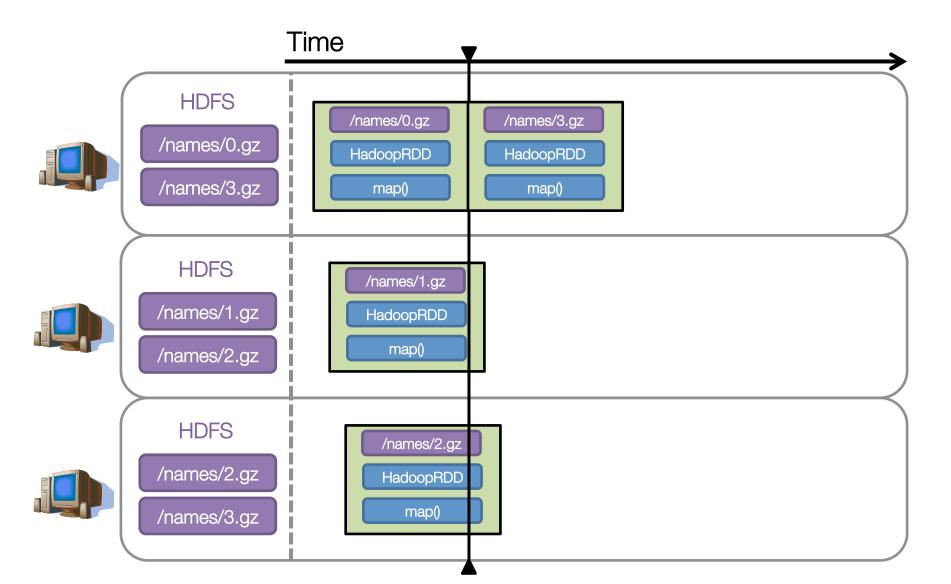
#### Step 3: Schedule tasks /names/1.gz HadoopRDD map() Time **HDFS** /names/0.gz /names/0.gz HadoopRDD /names/3.gz map() **HDFS** /names/1.gz /names/2.gz **HDFS** /names/2.gz /names/3.gz

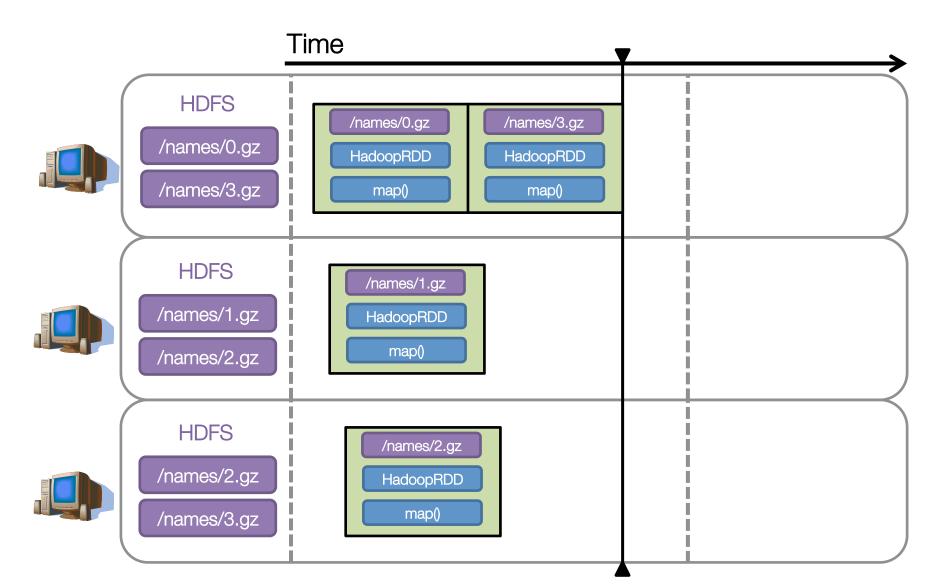


#### Step 3: Schedule tasks /names/2.gz HadoopRDD map() Time\_ **HDFS** /names/0.gz /names/0.gz HadoopRDD /names/3.gz map() **HDFS** /names/1.gz /names/1.gz HadoopRDD map() /names/2.gz **HDFS** /names/2.gz /names/3.gz

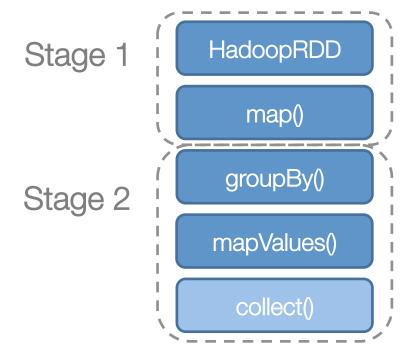


#### Step 3: Schedule tasks /names/3.gz HadoopRDD map() Time **HDFS** /names/0.gz /names/0.gz HadoopRDD /names/3.gz map() **HDFS** /names/1.gz /names/1.gz HadoopRDD map() /names/2.gz **HDFS** /names/2.gz /names/2.gz HadoopRDD map() /names/3.gz



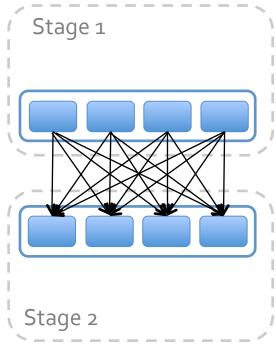


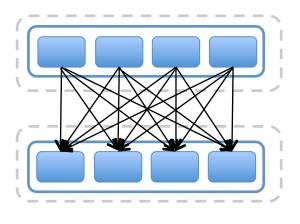
#### The Shuffle



#### The Shuffle

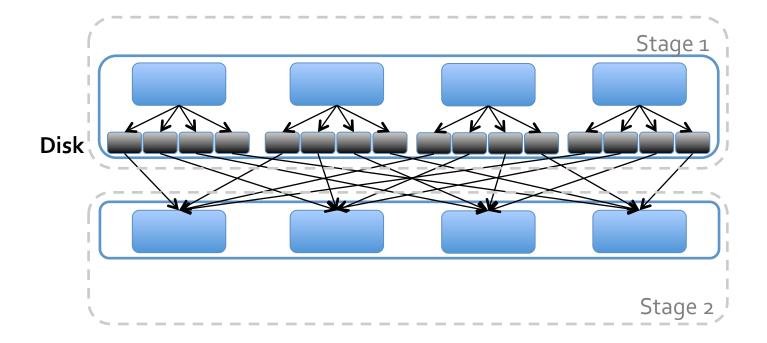
- Redistributes data among partitions
- Hash keys into buckets
- Optimizations:
  - Avoided when possible, if data is already properly partitioned
  - Partial aggregation reduces data movement





#### The Shuffle

- Pull-based, not push-based
- Write intermediate files to disk



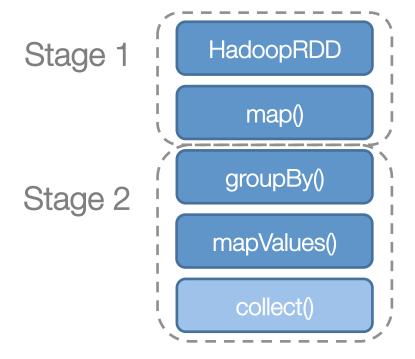
## Execution of a groupBy()

Build hash map within each partition

```
A => [Arsalan, Aaron, Andrew, Andrew, Andy, Ahir, Ali, ...], E => [Erin, Earl, Ed, ...]
```

 Note: Can spill across keys, but a single key-value pair must fit in memory

#### Done!



#### What went wrong?

- Too few partitions to get good concurrency
- Large per-key groupBy()
- Shipped all data across the cluster

#### Common issue checklist

- 1. Ensure enough partitions for concurrency
- 2. Minimize memory consumption (esp. of sorting and large keys in groupBys)
- 3. Minimize amount of data shuffled
- 4. Know the standard library

1 & 2 are about tuning number of partitions!

## Importance of Partition Tuning

- Main issue: too few partitions
  - Less concurrency
  - More susceptible to data skew
  - Increased memory pressure for groupBy, reduceByKey, sortByKey, etc.
- Secondary issue: too many partitions
- Need "reasonable number" of partitions
  - Commonly between 100 and 10,000 partitions
  - Lower bound: At least ~2x number of cores in cluster
  - Upper bound: Ensure tasks take at least 100ms

#### Memory Problems

- Symptoms:
  - Inexplicably bad performance
  - Inexplicable executor/machine failures (can indicate too many shuffle files too)
- Diagnosis:
  - Set spark.executor.extraJavaOptions to include
    - -XX:+PrintGCDetails
    - -XX:+HeapDumpOnOutOfMemoryError
  - Check dmesg for oom-killer logs
- Resolution:
  - Increase spark.executor.memory
  - Increase number of partitions
  - Re-evaluate program structure (!)

```
sc.textFile("hdfs:/names")
.map(name => (name.charAt(0), name))
.groupByKey()
.mapValues { names => names.toSet.size }
.collect()
```

- Ensure enough partitions for concurrency
- 2. Minimize memory consumption (esp. of large groupBys and sorting)
- 3. Minimize data shuffle
- 4. Know the standard library

```
sc.textFile("hdfs:/names")
    .repartition(6)
    .map(name => (name.charAt(0), name))
    .groupByKey()
    .mapValues { names => names.toSet.size }
    .collect()
```

- Ensure enough partitions for concurrency
- 2. Minimize memory consumption (esp. of large groupBys and sorting)
- 3. Minimize data shuffle
- 4. Know the standard library

```
sc.textFile("hdfs:/names")
    repartition(6)
    distinct()
    .map(name => (name.charAt(0), name))
    .groupByKey()
    .mapValues { names => names.toSet.size }
    .collect()
```

- Ensure enough partitions for concurrency
- 2. Minimize memory consumption (esp. of large groupBys and sorting)
- 3. Minimize data shuffle
- 4. Know the standard library

```
sc.textFile("hdfs:/names")
    .repartition(6)
    .distinct()
    .map(name => (name.charAt(0), name))
    .groupByKey()
    .mapValues { names => names.size }
    .collect()
```

- Ensure enough partitions for concurrency
- 2. Minimize memory consumption (esp. of large groupBys and sorting)
- 3. Minimize data shuffle
- 4. Know the standard library

```
sc.textFile("hdfs:/names")
.distinct(numPartitions = 6)
.map(name => (name.charAt(0), name))
.groupByKey()
.mapValues { names => names.size }
.collect()
```

- Ensure enough partitions for concurrency
- 2. Minimize memory consumption (esp. of large groupBys and sorting)
- 3. Minimize data shuffle
- 4. Know the standard library

```
sc.textFile("hdfs:/names")
.distinct(numPartitions = 6)
.map(name => (name.charAt(0), 1))
.reduceByKey(_ + _)
.collect()
```

- Ensure enough partitions for concurrency
- 2. Minimize memory consumption (esp. of large groupBys and sorting)
- 3. Minimize data shuffle
- 4. Know the standard library

```
sc.textFile("hdfs:/names")
  .distinct(numPartitions = 6)
  .map(name \Rightarrow (name.charAt(0), \mathbf{1}))
  .reduceByKey( + )
  .collect()
Original:
sc.textFile("hdfs:/names")
  .map(name => (name.charAt(0), name))
  .groupByKey()
  .mapValues { names => names.toSet.size }
  .collect()
```



Questions?