Understanding Spark Application Performance with Spark's Web UI





Lesson Objectives

- After completing this lesson, you should be able to:
 - Understand how concepts like cores, executors, tasks and stages apply to actual Spark applications
 - Be able to use the Spark Web UI to understand how your Spark application works
 - Be able to find the running time of each job, stage, and task





Spark Web UI

Monitor running applications

```
import org.apache.spark.sql.SparkSession
val sparkSession = SparkSession.builder.
  master("local[4]"). //start with 4 cores
  appName("My Spark Application").
```

This lesson uses Spark 2.0.0

16/07/06 13:32:27 INFO Utils: Successfully started service **'SparkUI' on port 4040.** 16/07/06 13:32:27 INFO SparkUI: Bound SparkUI to 0.0.0.0, and started at http://192.168.1.100:4040

sparkSession.sparkContext.uiWebUrl

Some(http://192.168.1.100:4040)

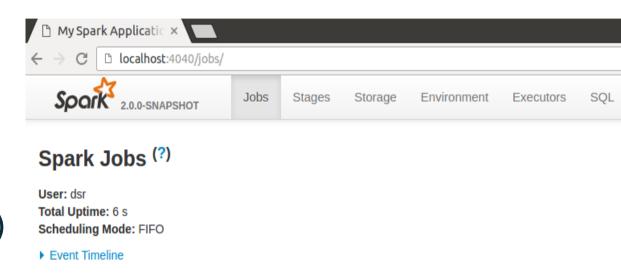


getOrCreate()



Spark Web UI

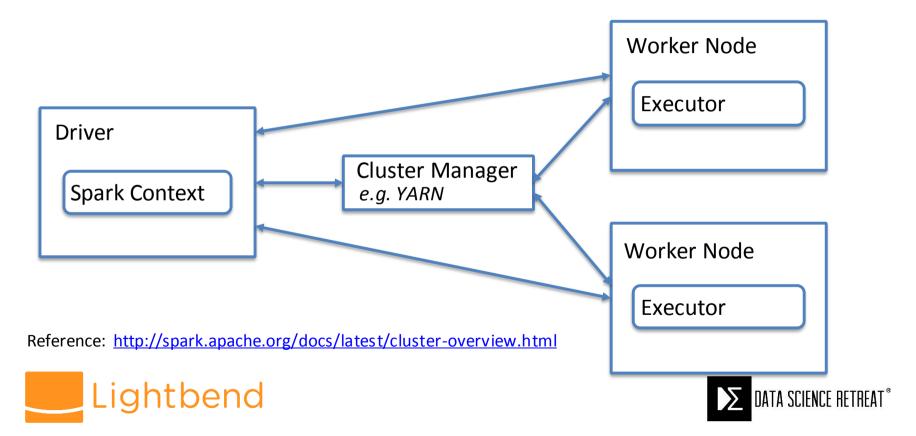
- Environment
- Jobs
- Stages
- Tasks
- Cache (Storage)
- SQL
- Streaming







Spark Architecture



Inside a Worker Node

Worker Node									
Executor									
Cache									
Block Manager									
Core	Core	Core	Core						
Task	Task	Task	Task						

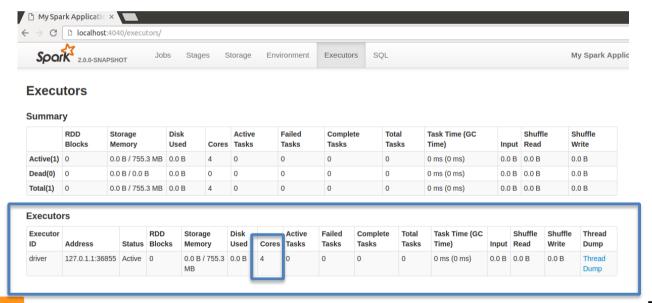




Executor and Cores in Web Ul

import org.apache.spark.sql.SparkSession

val sparkSession = SparkSession.builder.master("local[4]")







Spark Application Life Cycle

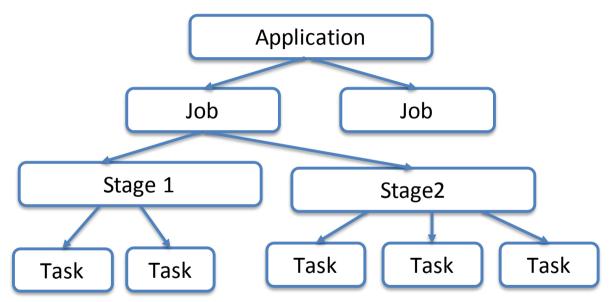
- Create a Spark context and obtain executors
- Spark application calls an action (such as collect() or count())
- Spark creates an execution schedule that consists of tasks
 - From DAG to stages to tasks
- The driver program asks the cluster manager to run tasks on executors
- Executors run the tasks
- The driver program collects the results
- The application stops the Spark context

Reference: https://spark.apache.org/docs/latest/job-scheduling.html





Jobs, Stages and Tasks



Spark Web UI shows breakdown of a running application





Spark Program Example

```
val sparkSession = . . .
val trans = sparkSession.sparkContext
.parallelize(List(1,2,3,4),
    numSlices = 4).map(_ + 1)
val output = trans.collect()
```

- Create a Spark context
- Use 4 partitions
- Define transformations (e.g. map, filter)
- Materialize result by applying an action (e.g. count, collect)





Jobs

User script:

```
val trans = sparkSession.sparkContext
.parallelize(List(1,2,3,4),
   numSlices = 4).map(_ + 1)
val output = trans.collect()
```

Implementation of collect:

```
def collect(): Array[T] = {
  val results = sc.runJob(/*...*/)
  //...
}
```







Stages

```
val trans = ...
  parallelize(...).
  map(...)
trans.toDebugString
```

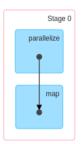
```
(4) MapPartitionsRDD[1] at map at <console>:13 []| ParallelCollectionRDD[0] at parallelize at <console>:13 []
```



Details for Job 0

Status: SUCCEEDED Completed Stages: 1

- ▶ Event Timeline
- ▼ DAG Visualization



Completed Stages (1)

Stage Id	Description	Submitted	Duration	Tasks: Succeeded/Total
0	collect at <console>:14 +details</console>	2016/07/06 13:39:51	0.2 s	4/4





Tasks

Aggregated Metrics by Executor

Executor ID A	Address	Task Time	Total Tasks	Failed Tasks	Killed Tasks	Succeeded Tasks
driver	127.0.1.1:36855	0.5 s	4	0	0	4

Tasks

Index •	ID	Attempt	Status	Locality Level	Executor ID / Host	Launch Time	Duration	Scheduler Delay	Task Deserialization Time	GC Time	Result Serialization Time	Getting Result Time	Peak Execution Memory	Errors
0	0	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 16:04:30	36 ms	99 ms	43 ms		0 ms	0 ms	0.0 B	
1	1	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 16:04:30	17 ms	42 ms	45 ms		0 ms	0 ms	0.0 B	
2	2	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 16:04:30	29 ms	28 ms	49 ms		0 ms	0 ms	0.0 B	
3	3	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 16:04:30	12 ms	53 ms	48 ms		0 ms	0 ms	0.0 B	

Number of tasks equals number of RDD





Output

```
adding task set 0.0 with 4 tasks
```

16/07/06 13:39:51 INFO TaskSetManager: Starting task 0.0 in stage 0.0 (TID 0, localhost, partition 0, PROCESS_LOCAL, 5288 bytes)

16/07/06 13:39:51 INFO TaskSetManager: Starting task 1.0 in stage 0.0 (TID 1, localhost, partition 1, PROCESS LOCAL, 5288 bytes)

16/07/06 13:39:51 INFO TaskSetManager: Starting task 2.0 in stage 0.0 (TID 2, localhost, partition 2, PROCESS LOCAL, 5288 bytes)

16/07/06 13:39:51 INFO TaskSetManager: Starting task 3.0 in stage 0.0 (TID 3, localhost, partition 3, PROCESS_LOCAL, 5288 bytes)

16/07/06 13:39:51 INFO Executor: Running task 1.0 in stage 0.0 (TID 1) 16/07/06 13:39:51 INFO Executor: Running task 3.0 in stage 0.0 (TID 3) 16/07/06 13:39:51 INFO Executor: Running task 0.0 in stage 0.0 (TID 0) 16/07/06 13:39:51 INFO Executor: Running task 2.0 in stage 0.0 (TID 2)

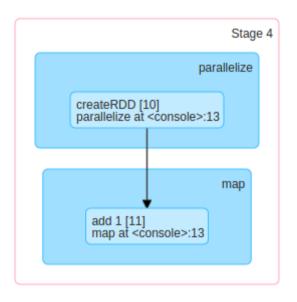
Task (**TID**) can be used to find tasks in Spark Web UI





Naming RDDs

```
val namedTrans =
  sparkSession.sparkContext.
  parallelize(List(1,2,3,4), 4).
    setName("createRDD").
  map(_ + 1).
  setName("add 1")
```





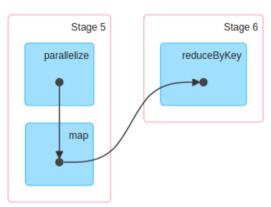


Example with Two Stages

```
val sc = sparkSession.sparkContext
val transformation2Stages =
 //stage 1
 sc.parallelize(List(1,2,3,4,1,2,3,4),4).
      setName("createRDD").
    map(value => (value, 1)).
      setName("makeKeyValue").
 //stage 2
    reduceByKey( + ).
      setName("reduceSum")
```

transformation2Stages.toDebugString

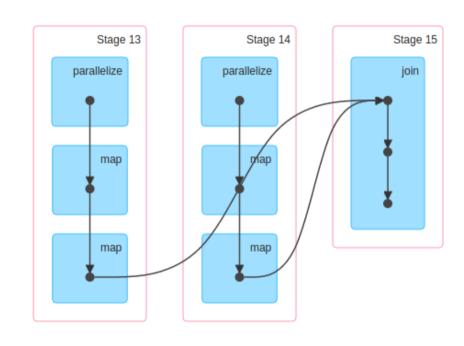






Repeated Computation: No Cache

```
val initialRDD =
  sc.parallelize(List(1,2,3,4), numSlices = 4).
    setName("createRDD")
val slowComputation = initialRDD.map(key => {
  Thread.sleep(5000); key}). /*cache()*/
  setName("slowComputation")
val remap1 = slowComputation.map(k => (k, 1)).
  setName("remap 1")
val remap2 = slowComputation.map(k => (k + 1, 1)).
  setName("remap 2")
val joined = remap1.join(remap2).setName("joined")
ioined.count()
```

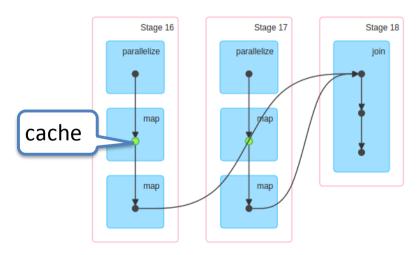


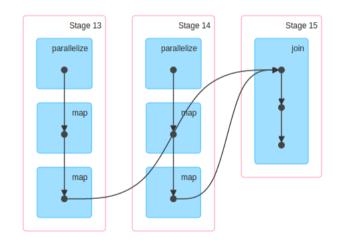




Cache vs No Cache

```
val slowComputation = initialRDD.map(key => {
    Thread.sleep(5000); key}).cache()
    setName("slowComputation")
```









Effect of Cache

Cache

Completed Jobs (11)

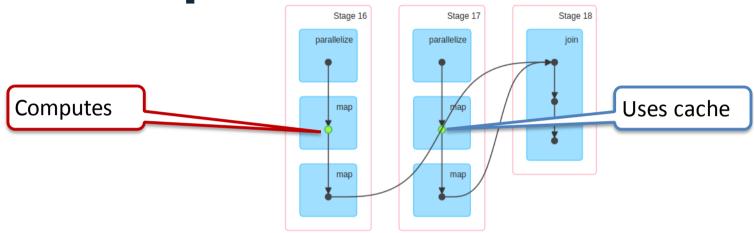
Job Id	Description	Submitted	Duration	stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
10	count at <console>:19</console>	2016/07/06 17:23:22	5 s	3/3	12/12
9	count at <console>:30</console>	2016/07/06 17:21:37	10 s	3/3	12/12

No Cache





Computation vs. Cache



Index	ID	Attempt	Status	Locality Level	Executor ID / Host	Launch Time	Duration
0	56	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 17:23:22	5 s
1	57	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 17:23:22	5 s
2	58	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 17:23:22	5 s
3	59	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 17:23:22	5 s

Index •	ID	Attempt	Status	Locality Level	Executor ID / Host		Duration
0	60	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 17:23:27	16 ms
1	61	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/06 17:23:27	2 ms
2	62	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/0 17:23:27	1 ms
3	63	0	SUCCESS	PROCESS_LOCAL	driver / localhost	2016/07/0 17:23:27	2 ms

Storage Tab

val slow Computation = initial RDD.map(key => { Thread.sleep(5000); key}).cache() setName("slowComputation") Stages Storage Environment My Spark Application application UI Executors SQL **Storage RDDs RDD Name** Storage Level **Cached Partitions Fraction Cached** Size in Memory Size on Disk slowComputation Memory Deserialized 1x Replicated 100% 96.0 B 0.0 B

Note: the effect of cache on performance depends on the available memory in Spark and on other jobs running and using the cache in the same Spark application. Consider using persist with a disk serialization option.





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