Introduction to Big Data with Apache Spark







This Workshop

Programming Spark

Resilient Distributed Datasets (RDDs)

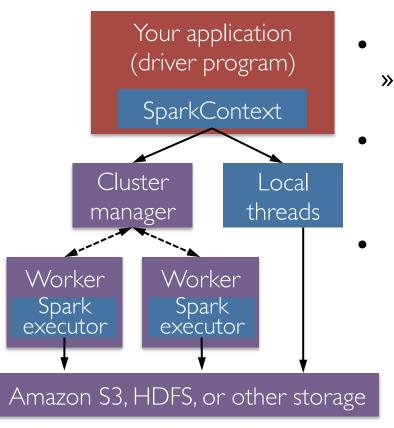
Creating an RDD

Spark Transformations and Actions

Python Spark (pySpark)

- We are using the Python programming interface to Spark (<u>pySpark</u>)
- pySpark provides an easy-to-use programming abstraction and parallel runtime:
 - » "Here's an operation, run it on all of the data"
- RDDs are the key concept

Spark Driver and Workers



- A Spark program is two programs:
 - » A driver program and a workers program
- Worker programs run on cluster nodes or in local threads
- RDDs are distributed across workers

Spark Context

- A Spark program first creates a **SparkContext** object
 - » Tells Spark how and where to access a cluster
 - » pySpark shell and Databricks Cloud automatically create the sc variable
 - » <u>iPython</u> and programs must use a constructor to create a new **SparkContext**
- Use SparkContext to create RDDs

Resilient Distributed Datasets

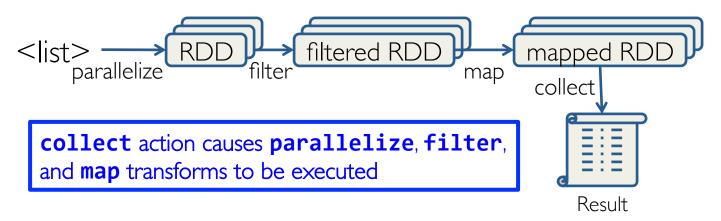
- The primary abstraction in Spark
 - » Immutable once constructed
 - » Track lineage information to efficiently recompute lost data
 - » Enable operations on collection of elements in parallel
- You construct RDDs
 - » by parallelizing existing Python collections (lists)
 - » by transforming an existing RDDs
 - » from files in HDFS or any other storage system

RDDs

- Two types of operations: transformations and actions
- Transformations are lazy (not computed immediately)
- Transformed RDD is executed when action runs on it
- Persist (cache) RDDs in memory or disk

Working with RDDs

- Create an RDD from a data source:
- <
- Apply transformations to an RDD: map filter
- Apply actions to an RDD: collect count



Creating an RDD

Create RDDs from Python collections (lists)

```
No computation occurs with sc.parallelize()
>>> data = [1, 2, 3, 4, 5]
>>> data
[1, 2, 3, 4, 5]

>>> rDD = sc.parallelize(data, 4)
>>> rDD
ParallelCollectionRDD[0] at parallelize at PythonRDD.scala:229
```

Creating RDDs

From HDFS, text files, <u>Hypertable</u>, <u>Amazon S3</u>, <u>Apache Hbase</u>,
 SequenceFiles, any other Hadoop <u>InputFormat</u>, and directory or glob wildcard: /data/201404*

```
>>> distFile = sc.textFile("/user/hadoop/README.md", 4)
>>> distFile
/user/hadoop/README.md
MapPartitionsRDD[1] at textFile at
NativeMethodAccessorImpl.java:0
```

Spark Transformations

- Create new datasets from an existing one
- Use lazy evaluation: results not computed right away instead Spark remembers set of transformations applied to base dataset
 - » Spark optimizes the required calculations
 - » Spark recovers from failures and slow workers
- Think of this as a recipe for creating result

Some Transformations

Transformation	Description
map(func)	return a new distributed dataset formed by passing each element of the source through a function func
filter(func)	return a new dataset formed by selecting those elements of the source on which <i>func</i> returns true
<pre>distinct([numTasks]))</pre>	return a new dataset that contains the distinct elements of the source dataset
<pre>flatMap(func)</pre>	similar to map, but each input item can be mapped to 0 or more output items (so <i>func</i> should return a Seq rather than a single item)

Review: Python lambda Functions

- Small anonymous functions (not bound to a name)
 lambda a, b: a + b
 - » returns the sum of its two arguments
- Can use lambda functions wherever function objects are required
- Restricted to a single expression

Transformations

```
>>> rdd = sc.parallelize([1, 2, 3, 4])
>>> rdd.map(lambda x: x * 2)
RDD: [1, 2, 3, 4] → [2, 4, 6, 8]

>>> rdd.filter(lambda x: x % 2 == 0)
RDD: [1, 2, 3, 4] → [2, 4]

>>> rdd2 = sc.parallelize([1, 4, 2, 2, 3])
>>> rdd2.distinct()
RDD: [1, 4, 2, 2, 3] → [1, 4, 2, 3]
```

Spark Actions

- Cause Spark to execute recipe to transform source
- Mechanism for getting results out of Spark

Some Actions

Action	Description
reduce(func)	aggregate dataset's elements using function func. func takes two arguments and returns one, and is commutative and associative so that it can be computed correctly in parallel
take(n)	return an array with the first <i>n</i> elements
<pre>collect()</pre>	return all the elements as an array WARNING: make sure will fit in driver program
<pre>takeOrdered(n, key=func)</pre>	return n elements ordered in ascending order or as specified by the optional key function

Getting Data Out of RDDs

```
>>> rdd = sc.parallelize([1, 2, 3])
>>> rdd.reduce(lambda a, b: a * b)
Value: 6

>>> rdd.take(2)
Value: [1,2] # as list

>>> rdd.collect()
Value: [1,2,3] # as list
```

Spark Program Lifecycle

- I. Create RDDs from external data or <u>parallelize</u> a collection in your driver program
- 2. Lazily <u>transform</u> them into new RDDs
- 3. cache() some RDDs for reuse
- 4. Perform <u>actions</u> to execute parallel computation and produce results

Spark Key-Value RDDs

- Similar to Map Reduce, Spark supports Key-Value pairs
- Each element of a Pair RDD is a pair tuple

```
>>> rdd = sc.parallelize([(1, 2), (3, 4)])
RDD: [(1, 2), (3, 4)]
```

Some Key-Value Transformations

Key-Value Transformation	Description
reduceByKey(<i>func</i>)	return a new distributed dataset of (K,V) pairs where the values for each key are aggregated using the given reduce function <i>func</i> , which must be of type $(V,V) \rightarrow V$
sortByKey()	return a new dataset (K,V) pairs sorted by keys in ascending order
<pre>groupByKey()</pre>	return a new dataset of (K, Iterable <v>) pairs</v>

Key-Value Transformations

Spark References

- http://spark.apache.org/docs/latest/programming-guide.html
- http://spark.apache.org/docs/latest/api/python/index.html