

Apache Spark Through Email

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Intro, Slides And Code

- mostly Java, big data with Hadoop
- big data with Spark, Databricks, Scala
- Now Asymmetrik - Scala, Spark, Elasticsearch, Akka...
- Slides: <https://github.com/medale/spark-mail/blob/master/presentation/ApacheSparkThroughEmail.pdf>
- Spark Code Examples:
<https://github.com/medale/spark-mail/>
- README.md
 - describes how to get and parse Enron email dataset
 - Spark, Docker notebook setup

Goals

- Scalable data processing
- Spark structured processing
- Spark concepts - RDD, Dataset/DataFrame, partitions
- Notebooks, Spark UI

- Laptop
- Explore subset, develop approaches/algorithms, find *features*

- Standalone server - more memory, faster CPU, more storage

Data Science for Larger Dataset (Vertical Scaling)

- Big iron - lots of cores, memory, disk/SSDs, GPUs

Data Science for Large Datasets (Horizontal Scaling)

- Parallelize, coordinate compute among many “commodity” machines
- Deal with *failure*

Big Data Framework - Apache Hadoop

- Google GFS (2003), Google MapReduce (2004)
- Hadoop (Nutch - open source web crawler/Lucene) - Doug Cutting, Mike Cafarella
 - Yahoo, Cloudera, Hortonworks, MapR

- HDFS, YARN, MapReduce (Spark replaces MR)
- HBase (Google BigTable), Cassandra, Accumulo
- Pig, Hive - MR scripting DSL/SQL

Apache Spark Components

- Foundation: Resilient Distributed Datasets (RDD)
 - Broadcast variables, accumulators
 - Java objects, should use Kryo serialization
- *Structured APIs* (use these) - Datasets, DataFrames, SQL
 - Spark manages object layout in memory, schemas, code generation
- Streaming, MLlib (Advanced analytics)
- Scala, Java, Python, R + library ecosystems
- Submit (Batch/Stream) or Shell/Notebooks (e.g. Zeppelin, Jupyter)

Hello, Spark Email World!

- Jupyter Notebook with Apache Toret
- Comment out `spark.close`, Restart & run all
- See `ApacheSparkThroughEmail1`

Cluster Manager, Driver, Executors, Tasks

- *Cluster manager*: Spark Standalone, Hadoop YARN, AWS EMR, Kubernetes, Mesos
- *Driver* (start once)
 - Execute user code
 - Schedules tasks for executors
 - Serialize code (closures with data) as tasks to executors
- *Executors* on worker nodes (start once, restart)
 - Cache - distributed memory for partitions
 - Execute tasks (threads/core = parallelism of tasks)
 - Read/manage partitions (serialization - Kryo)

SparkSession: Entry to cluster

- Notebook: `spark` provided (`spark.close` or JVM shutdown)
- Code: `SparkSession.builder()` w/ `appName`, `master`, `getOrCreate`
 - `spark-submit` (by hand, or Airflow, or EMR)
- Show `SparkSession` in Spark ScalaDocs (clickable link)

DataFrameReader: Input for structured data

- `spark.read` - built-in: jdbc, csv, json, parquet, text
- 3rd party: <https://spark-packages.org> - Avro, Redshift, MongoDB...
- Also, any Hadoop InputFormat via RDD/SparkContext

Scaling Behind the Scenes

- Executing notebook - show Spark Application UI (stay on Jobs page)
- Job - series of transformations followed by action!
 - *transformations*: select (projection), where, limit, cache
 - *actions*: count, df.write..., df.collect/take/head/first (memory!)
- Task - ~1 task per partition
 - Serialize code (no classes, non-serializable)
 - Data in closure (e.g. HashMap - use broadcast variables!)

Stages: Pipeline work per stage - shuffle

- Click on description for job with 209 tasks
- 3 Stages: pipeline per stage
 - Wholestage Code Gen - Tungsten code gen engine
 - InMemoryTableScan
 - Exchange (shuffle data with same from value)
 - HashPartitioner (Strings etc.), RangePartitioner (values)
 - groupBy
 - TakeOrderedAndProject exchange data
 - orderBy, limit
- Behind the scenes - scaling!

Where clause, Column methods, Built-in functions

- methods on Dataset
- methods on Columns
- sql.functions
- Ensure Notebook 1 is closed (kill on UI or spark.close)
- Restart & run all

- Show Spark Scala API docs
- Show Spark documentation

Parallelism and Partitioning

- Goldilocks - not too many, not too few
 - Too many tasks - scheduling overhead, little work
 - Too few - tasks take very long
- Initial parallelism - number of input “blocks”
- Shuffle - `spark.sql.shuffle.partitions` configuration
- repartition (shuffle)/coalesce (combine on same executor)
 - e.g. write out 1 partition

And now for something completely different: Colon Cancer

- Screening saves lives! Colonoscopy

Questions?

- medale@asymmetrik.com
- Ping Pong
- Baltimore Scala meetup - January 10
- Spark mail repo for getting/parsing Enron data, presentations, code, notebooks