Apache Spark Release 1.6

Patrick Wendell



About Me apwendell

U.C. Berkeley PhD, left to co-found Databricks

Coordinate community roadmap

Frequent release manager for Spark





About Databricks

Founded by Spark team, donated Spark to Apache in 2013 and lead development today.

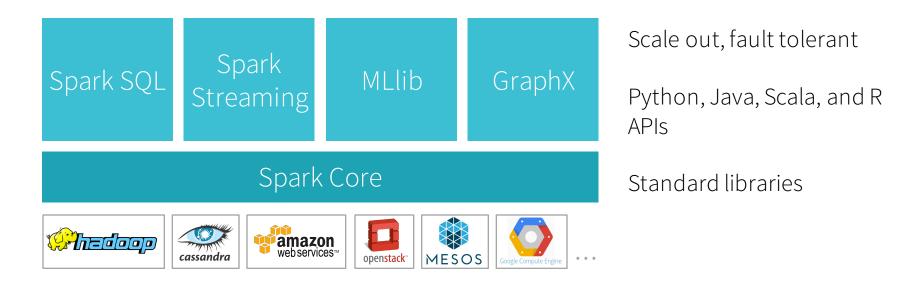
Collaborative, cloud-hosted data platform powered by Spark

Free trial to check it out

https://databricks.com/



Apache Spark Engine



Unified engine across diverse workloads & environments



Open Source Ecosystem

Applications











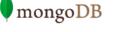


MESOS











cassandra



HBASE

TACHYON













Environments

Users











































Distributors & Apps

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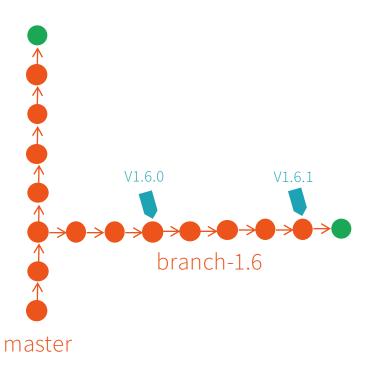








Spark's 3 Month Release Cycle



For production jobs, use the latest release

To try out unreleased features or fixes, use nightly builds
people.apache.org/~pwendell/spark-nightly/



Spark 1.6

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Spark 1.6 Release

Will ship upstream through Apache foundation in December (likely)

Key themes

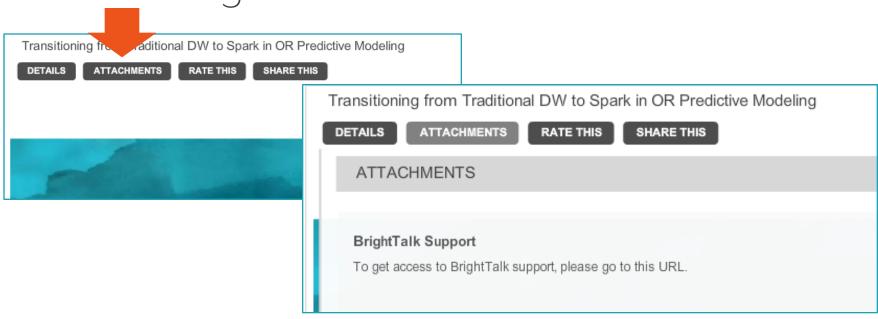
Out of the box performance

Previews of key new API's

Follow along with me at http://bit.ly/10BkjMM

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Follow_along



http://bit.ly/1lrvdLc



Memory Management in Spark: <= 1.5

- Two separate memory managers:
 - Execution memory: computation of shuffles, joins, sorts, aggregations
 - Storage memory: caching and propagating internal data sources across cluster
- Challenges with this:
 - Manual intervention to avoid unnecessary spilling
 - No good defaults for all workloads meaning lost efficiency
- Goal: Allow memory regions to shrink/grow dynamically



Unified Memory Management in Spark 1.6

- Can cross between execution and storage memory
 - When execution memory exceeds its own region, it can borrow as much of the storage space as is free and vice versa
 - Borrowed storage memory can be evicted at any time
- Significantly reduces configuration
 - Can define low water mark for storage (below which we won't evict)
- Reference: [SPARK-10000]

History of Spark API's

RDD API (2011)

Distribute collection of JVM objects

Functional

operators (map, filter, etc)

DataFrame API (2013)

Distribute collection of Row objects

Logical plans and

Expression-based

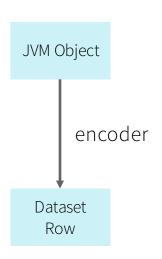
operations and UDF's optimizer

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Dataset

"Encoder" converts from JVM Object into a Dataset Row

Checkout [SPARK-9999]





Dataset API in Spark 1.6

Typed interface over DataFrames / Tungsten

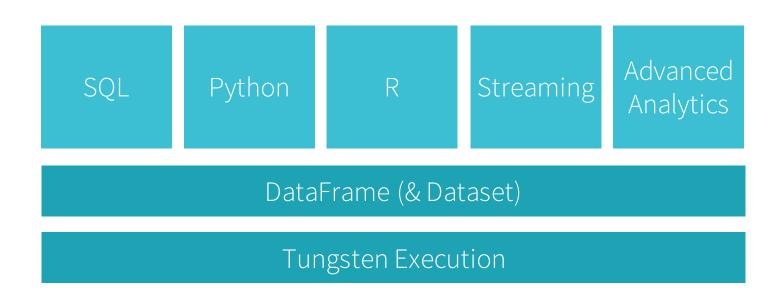
```
case class Person(name: String, age: Long)

val dataframe = read.json("people.json")

val ds: Dataset[Person] = dataframe.as[Person]

ds.filter(p => p.name.startsWith("M"))
    .groupBy($"name")
    .avg("age")
```







Other Notable Core Engine Features

SQL directly over files

Advanced JSON parsing

Better instrumentation for SQL operators

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Demos of What We Learned So Far



Advanced Layout of Cached Data

Storing partitioning and ordering schemes in In-memory table scan allows for performance improvements: e.g. in Joins, an extra partition step can be saved based on this information

Adding distributeBy and localSort to DF API

Similar to HiveQL's **DISTRIBUTE BY**

allows the user to control the partitioning and ordering of a data set

Checkout [SPARK-4849]



[Streaming] New improved state management

Introducing a DStream transformation for stateful stream processing

Does not scan every key

Easier to implement common use cases

timeout of idle data

returning items other than state

Supercedes updateStateByKey in functionality and performance. trackStateByKey (note, this name may change)



[Streaming] trackStateByKey example (name may change)

```
// Initial RDD input
    val initialRDD = ssc.sparkContext.parallelize(...)
    // ReceiverInputDStream
    val lines = ssc.socketTextStream(...)
    val words = lines.flatMap(...)
    val wordDStream = words.map(x \Rightarrow (x, 1))
    // stateDStream using trackStateByKey
    val trackStateFunc = (...) { ... }
    val stateDStream =
       wordDStream.trackStateByKey(StateSpec.function(trackStateFunc).initialSta
       te(initialRDD))
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```

[Streaming] Display the failed output op in Streaming

Checkout: [SPARK-10885] PR#8950

Duration	Status	Job Id	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total	Error
-	Succeeded	-	-	-	-	-
-	Failed due to error: java.lang.RuntimeException: xxx +details		-	-	-	-
	java.lang.RuntimeExcep tion: xxx					
17 ms	Succeeded	81	13 ms	2/2	9/9	
		82	4 ms	1/1 (1 skipped)	5/5 (4 skipped)	
8 ms	Failed due to Spark job error +details	83	4 ms	1/1 (1 skipped)	5/5 (4 skipped)	
		84	4 ms	0/1 (1 failed) (1 skipped)	0/5 (1 failed) (4	Job aborted due to stage failure: Task 2 in stage 168.0 failed 1 times, most recent failure: Lost task 2.0 in stage 168.0 (TID 517, localhost): java.lang.RuntimeException xxx +details



[MLlib]: Pipeline persistence

Persist ML Pipelines to:

Save models in the spark.ml API

Re-run workflows in a reproducible manner

Export models to non-Spark apps (e.g., model server)

This is more complex than ML model persistence because:

Must persist Transformers and Estimators, not just Models.

We need a standard way to persist Params.

Pipelines and other meta-algorithms can contain other Transformers and Estimators, including as Params.

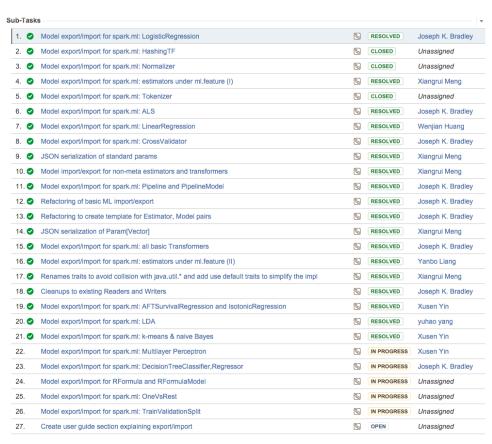
We should save feature metadata with Models



[MLlib]: Pipeline persistence

Reference [SPARK-6725] Adding model export/import to the spark.ml API.

Adding the internal Saveable/Loadable API and Parquet-based format





R-like statistics for GLMs

```
> # Model summary are returned in a similar format to R's native glm().
  summary(model)
 (1) Spark Jobs
$devianceResiduals
 Min
          Max
 -1.307112 1.412532
$coefficients
                  Estimate Std. Error t value Pr(>|t|)
(Intercept) 2.251393 0.3697543 6.08889 9.568102e-09
Sepal_Width
                  0.8035609 0.106339 7.556598 4.187317e-12
Species_versicolor 1.458743 0.1121079 13.01195 0
Species_virginica 1.946817 0.100015
                                       19.46525 0
Command took 0.90s
```

Provide R-like summary statistics for ordinary least squares via normal equation solver

Check out [SPARK-9836]



Performance

<u>SPARK-10000</u> **Unified Memory Management** - Shared memory for execution and caching instead of exclusive division of the regions.

<u>SPARK-10917</u>, <u>SPARK-11149</u> In-memory Columnar Cache Performance - Significant (up to 14x) speed up when caching data that contains complex types in DataFrames or SQL.

<u>SPARK-11389</u> **SQL Execution Using Off-Heap Memory** - Support for configuring query execution to occur using off-heap memory to avoid GC overhead

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Performance (continued)

<u>SPARK-4849</u> Advanced Layout of Cached Data - storing partitioning and ordering schemes in In-memory table scan, and adding distributeBy and localSort to DF API

<u>SPARK-9858</u> Adaptive query execution - Initial support for automatically selecting the number of reducers for joins and aggregations.

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Spark SQL

SPARK-9999 Dataset API

SPARK-11197 SQL Queries on Files

SPARK-11745 Reading non-standard JSON files

SPARK-10412 Per-operator Metrics for SQL Execution

SPARK-11329 Star (*) expansion for StructTypes

SPARK-11111 Fast null-safe joins

SPARK-10978 Datasource API Avoid Double Filter



Spark Streaming

API Updates

<u>SPARK-2629</u> New improved state management

SPARK-11198 Kinesis record deaggregation

SPARK-10891 Kinesis message handler function

<u>SPARK-6328</u> Python Streaming Listener API

UI Improvements

Made failures visible in the streaming tab, in the timelines, batch list, and batch details page.

Made output operations visible in the streaming tab as progress bars



MLlib: New algorithms / models

SPARK-8518 Survival analysis - Log-linear model for survival analysis

<u>SPARK-9834</u> Normal equation for least squares - Normal equation solver, providing R-like model summary statistics

<u>SPARK-3147</u> Online hypothesis testing - A/B testing in the Spark Streaming framework

<u>SPARK-9930</u> New feature transformers - ChiSqSelector, QuantileDiscretizer, SQL transformer

<u>SPARK-6517</u> Bisecting K-Means clustering - Fast top-down clustering variant of K-Means



MLlib: API Improvements

ML Pipelines

<u>SPARK-6725</u> Pipeline persistence - Save/load for ML Pipelines, with partial coverage of spark.ml algorithms

<u>SPARK-5565</u> LDA in ML Pipelines - API for Latent Dirichlet Allocation in ML Pipelines

RAPI

<u>SPARK-9836</u> R-like statistics for GLMs - (Partial) R-like stats for ordinary least squares via summary(model)

<u>SPARK-9681</u> Feature interactions in R formula - Interaction operator ":" in R formula

Python API - Many improvements to Python API to approach feature parity



MLlib: Miscellaneous Improvements

<u>SPARK-7685</u>, <u>SPARK-9642</u> Instance weights for GLMs - Logistic and Linear Regression can take instance weights

<u>SPARK-10384</u>, <u>SPARK-10385</u> <u>Univariate and bivariate statistics in DataFrames - Variance, stddev, correlations, etc.</u>

SPARK-10117 LIBSVM data source - LIBSVM as a SQL data source



For More Information

Apache Spark 1.6.0 Release Preview: http://apache-spark-developers-list.1001551.n3.nabble.com/ANNOUNCE-Spark-1-6-0-Release-Preview-td15314.html

Spark 1.6 Preview available in Databricks:

https://databricks.com/blog/2015/11/20/announcing-spark-1-6-preview-indatabricks.html



Notebooks

Spark 1.6 Improvements Notebook:

http://cdn2.hubspot.net/hubfs/438089/notebooks/Spark_1.6_Improvements.html?t=1448929686268

Spark 1.6 R Improvements Notebook:

http://cdn2.hubspot.net/hubfs/438089/notebooks/Spark_1.6_R_Improvements.html?t=1448946977231





Join us at Spark Summit East

February 16-18, 2016 | New York City





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