

with Advanced Performance Tuning

Xiao Li & Wenchen Fan Spark Summit | SF | Jun 2018



#### About US

- Software Engineers at adatabricks
- Apache Spark Committers and PMC Members





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### Databricks' Unified Analytics Platform

Unifies Data Engineers and Data Scientists COLLABORATIVE NOTEBOOKS

Data Engineers

Data Scientists

Unifies Data and Al Technologies Data Engineers

Data Scientists

DATABRICKS RUNTIME

Powered by Spork

Delta SQL Streaming PYTORCH R Studio TensorFlow Constitution

Eliminates infrastructure complexity



**CLOUD NATIVE SERVICE** 



### Spark SQL

A highly scalable and efficient relational processing engine with ease-to-use APIs and mid-query fault tolerance.







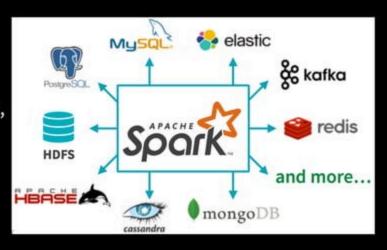






#### Run Everywhere

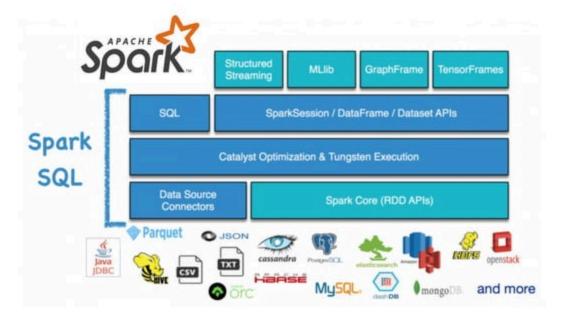
Processes, integrates and analyzes the data from diverse data sources (e.g., Cassandra, Kafka and Oracle) and file formats (e.g., Parquet, ORC, CSV, and JSON)



#### The not-so-secret truth...







### Not Only SQL

Powers and optimizes the other Spark applications and libraries:

- Structured streaming for stream processing
- MLlib for machine learning
- GraphFrame for graph-parallel computation
- Your own Spark applications that use SQL, DataFrame and Dataset APIs

#### Lazy Evaluation

Optimization happens as late as possible, therefore Spark SQL can optimize across functions and libraries

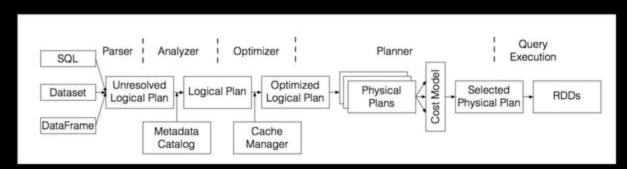
Holistic optimization when using these libraries and SQL/DataFrame/Dataset APIs in the same Spark application.

### New Features of Spark SQL in Spark 2.3

- PySpark Pandas UDFs [SPARK-22216] [SPARK-21187]
- Stable Codegen [SPARK-22510] [SPARK-22692]
- Advanced pushdown for partition pruning predicates [SPARK-20331]
- Vectorized ORC reader [SPARK-20682] [SPARK-16060]
- Vectorized cache reader [SPARK-20822]
- Histogram support in cost-based optimizer [SPARK-21975]
- Better Hive compatibility [SPARK-20236] [SPARK-17729] [SPARK-4131]
- More efficient and extensible data source API V2

### Spark SQL

#### A compiler from queries to RDDs.





#### Performance Tuning for Optimal Plans

Run EXPLAIN Plan. Interpret Plan. Tune Plan. Get the plans by running Explain command/APIs, or the SQL tab in either Spark UI or Spark History Server

```
Jobs Stages Storage Environment Decutors

Details for Query 111

Submitted Time: 2018/05/27 18:14:46

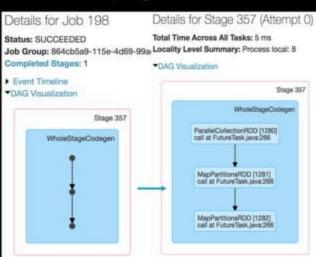
Duration: 0.2 s

Succeeded Jobs: 198 199

▼Details
```

```
= Physical Plan ==
*(3) HashAggregate(keys=[], functions=[finalmerge_count(merge count#2695L) AS count(1)#2691L], output=[count#2692L])
+- Exchange SinglePartition
+- *(2) HashAggregate(keys=[], functions=[partial_count(1) AS count#2695L], output=[count#2695L])
+- *(2) Project
+- *(2) BroadcastNestedLoopJoin BuildRight, LeftSemi, (a1#2679L < b1#2683L)
:- *(2) Project [id#2677L AS a1#2679L]
: +- *(2) Range (0, 20000, step=1, splits=8)
+- BroadcastExchange IdentityBroadcastMode
+- *(1) Project [id#2681L AS b1#2683L]
+- *(1) Range (0, 10000, step=1, splits=8)</pre>
```

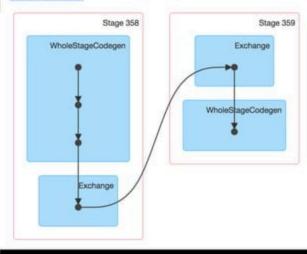
# More statistics from the Job page



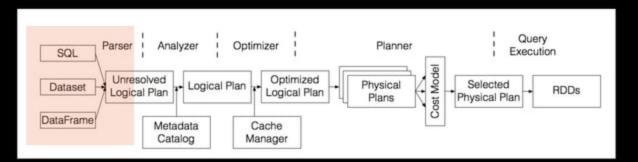
Details for Job 199

Status: SUCCEEDED Completed Stages: 2

- Event Timeline
- ▼DAG Visualization



#### Declarative APIs





#### Declarative APIs

#### Declare your intentions by

- SQL API: ANSI SQL:2003 and HiveQL.
- Dataset/DataFrame APIs: richer, languageintegrated and user-friendly interfaces

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#### Declarative APIs

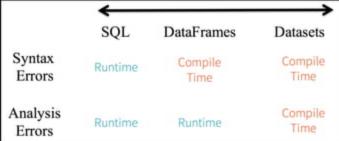
#### When should I use SQL, DataFrames or Datasets?

- The DataFrame API provides untyped relational operations
- The Dataset API provides a typed version, at the cost of performance due to heavy reliance on user-defined

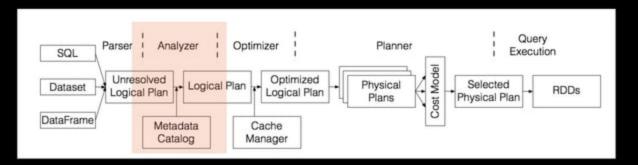
closures/lambdas.

[SPARK-14083]

http://dbricks.co/29xYnqR



### Metadata Catalog





### Metadata Catalog

- Persistent Hive metastore [Hive 0.12 Hive 2.3.3]
- Session-local temporary view manager
- Cross-session global temporary view manager
- Session-local function registry

### Metadata Catalog

#### Session-local function registry

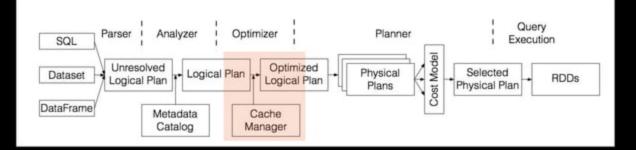
- Easy-to-use lambda UDF
- Vectorized PySpark Pandas UDF
- Native UDAF interface
- Support Hive UDF, UDAF and UDTF
- Almost 300 built-in SQL functions
- Next, <u>SPARK-23899</u> adds 30+ high-order built-in functions.
- Blog for high-order functions: <a href="https://dbricks.co/2rR8vAr">https://dbricks.co/2rR8vAr</a>

### Performance Tips - Catalog

#### Time costs of partition metadata retrieval:

- Upgrade your Hive metastore
- Avoid very high cardinality of partition columns
- Partition pruning predicates (improved in [SPARK-20331])

#### Cache Manager





#### Cache Manager

- Automatically replace by cached data when plan matching
- Cross-session
- Dropping/Inserting tables/views invalidates all the caches that depend on it
- Lazy evaluation

#### Performance Tips

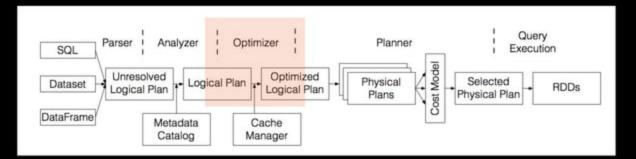
Cache: not always fast if spilled to disk.

- Uncache it, if not needed.

#### Next releases:

 A new cache mechanism for building the snapshot in cache. Querying stale data. Resolved by names instead of by plans. [SPARK-24461]

### Optimizer





### Optimizer

Rewrites the query plans using heuristics and cost.

- Column pruning
   Outer join elimination
- Predicate push down
   Constraint propagation
- Constant folding
   Join reordering
   and many more.

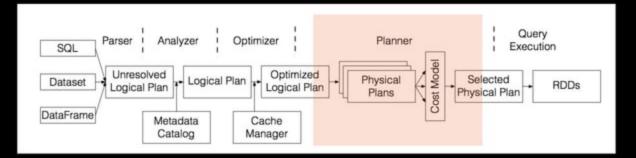


### Performance Tips

#### Roll your own Optimizer and Planner Rules

- In class ExperimentalMethods
  - var extraOptimizations: Seq[Rule[LogicalPlan]] = Nil
  - var extraStrategies: Seq[Strategy] = Nil
- Examples in the Herman's talk <u>Deep Dive into Catalyst</u> <u>Optimizer</u>
  - Join two intervals: <a href="http://dbricks.co/2etjIDY">http://dbricks.co/2etjIDY</a>

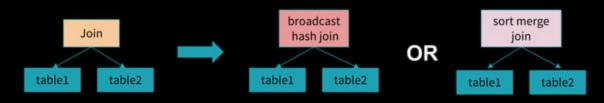
#### Planner





#### Planner

- Turn logical plans to physical plans. (what to how)
- Pick the best physical plan according to the cost



broadcast join has lower cost if one table can fit in memory



#### Performance Tips - Join Selection

## broadcast join table 1 ioin result broadcast table 2

### shuffle join table 1 shuffled join result table 2 shuffled



#### Performance Tips - Join Selection

broadcast join vs shuffle join (broadcast is faster)

- spark.sql.autoBroadcastJoinThreshold
- Keep the statistics updated
- broadcastJoin Hint

### Performance Tips - Equal Join

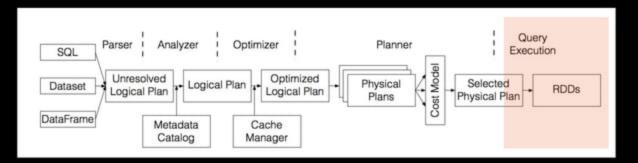
```
... t1 JOIN t2 ON t1.id = t2.id AND t1.value < t2.value
... t1 JOIN t2 ON t1.value < t2.value
```

Put at least one equal predicate in join condition

#### Performance Tips - Equal Join

```
... t1 JOIN t2 ON t1.id = t2.id AND t1.value < t2.value
... t1 JOIN t2 ON t1.value < t2.value
for l in left:
  for r in right:
                                               O(n ^ 2)
    if (satisfy join condition(l, r)):
      yield join(l, r)
hash_relation = build_relation(l)
for r in right:
  if (hash_relation.contains(r.joinKey)):
                                               O(n)
    l = hash relation.get(r.joinKey)
    if (satisfy_join_condition(l, r)):
      yield join(l, r)
```

### Query Execution





#### Query Execution

- Memory Manager: tracks the memory usage, efficiently distribute memory between tasks/operators.
- Code Generator: compiles the physical plan to optimal java code.
- Tungsten Engine: efficient binary data format and data structure for CPU and memory efficiency.

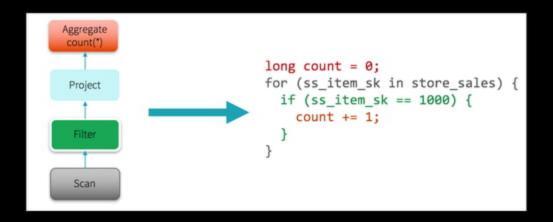
### Performance Tips - Memory Manager

Tune spark.executor.memory and spark.memory.fraction to leave enough space for unsupervised memory. Some memory usages are NOT tracked by Spark(netty buffer, parquet writer buffer).

Set spark.memory.offHeap.enabled and spark.memory.offHeap.size to enable offheap, and decrease spark.executor.memory accordingly.

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#### Whole Stage Code Generation



### Performance Tip - WholeStage codegen

Tune spark.sql.codegen.hugeMethodLimit to avoid big method(> 8k) that can't be compiled by JIT compiler.

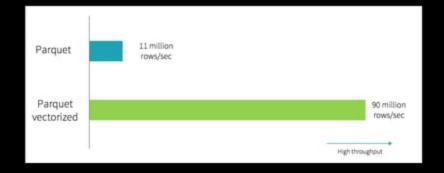
#### **Data Sources**

- Spark separates computation and storage.
- Complete data pipeline:
  - External storage feeds data to Spark.
  - · Spark processes the data
- Data source can be a bottleneck if Spark processes data very fast.

#### Scan Vectorization

- More efficient to read columnar data with vectorization.
- More likely for JVM to generate SIMD instructions.

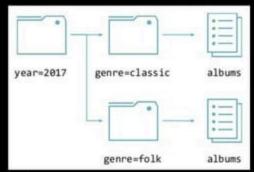
• ......

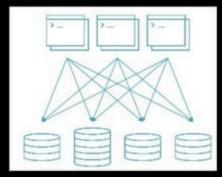




### Partitioning and Bucketing

- A special file system layout for data skipping and pre-shuffle.
- Can speed up query a lot by avoid unnecessary IO and shuffle.
- The summit talk: <a href="http://dbricks.co/2oG6ZBL">http://dbricks.co/2oG6ZBL</a>



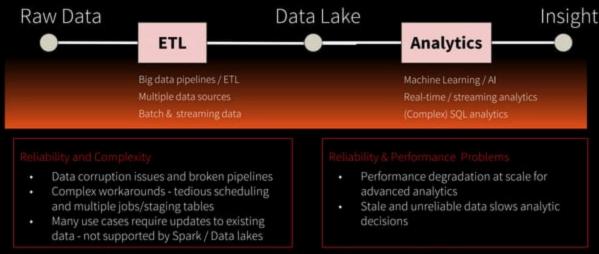




### Performance Tips

- Pick data sources that supports vectorized reading. (parquet, orc)
- For file-based data sources, creating partitioning/bucketing if possible.

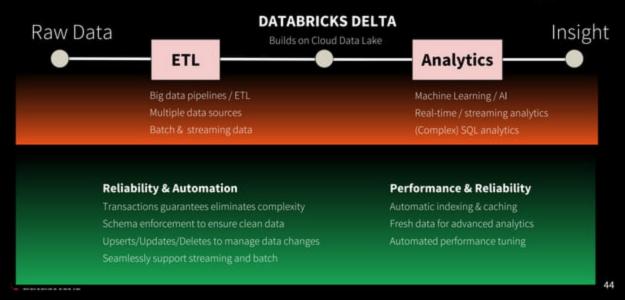
#### Yet challenges still remain



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Streaming magnifies these challenges

#### Databricks Delta address these challenges



# Thank you

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