Correctness and Performance of Apache Spark SQL

Spark + Al Summit, London



About us



BOGDAN GHIT

Databricks, Software Engineer

SQL performance optimizations

IBM T.J. Watson, Research Intern

Bid advisor for cloud spot markets

Delft University of Technology, PhD in Computer Science

- Resource management in datacenters
- Performance of Spark, Hadoop



NICOLAS POGGI

Databricks, Performance Engineer

Spark benchmarking



Barcelona Supercomputing - Microsoft Research Centre

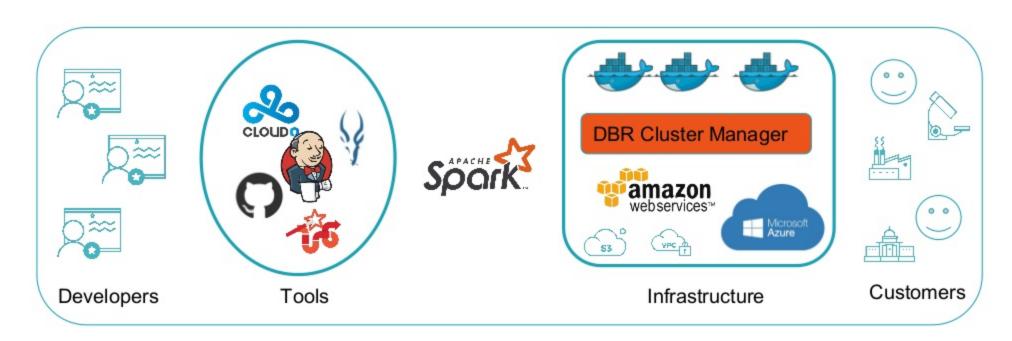
- Lead researcher ALOJA project
- New architectures for Big Data

BarcelonaTech (UPC), PhD in Computer Architecture

- Autonomic resource manager for the cloud
- Web customer modeling



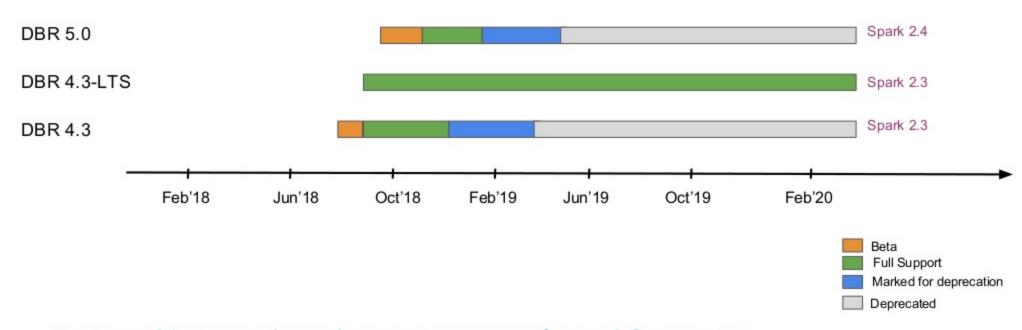
Databricks ecosystem







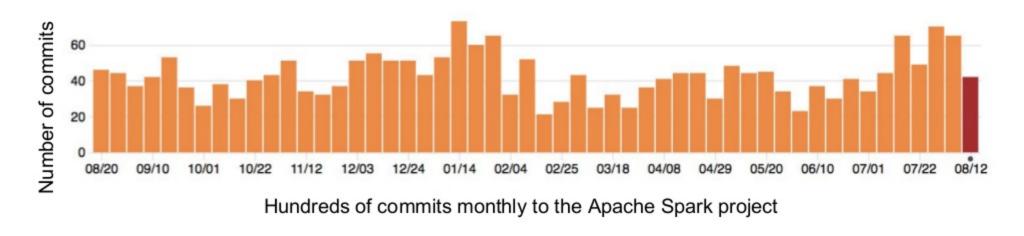
Databricks runtime (DBR) releases



Our goal is to make releases automatic and frequent



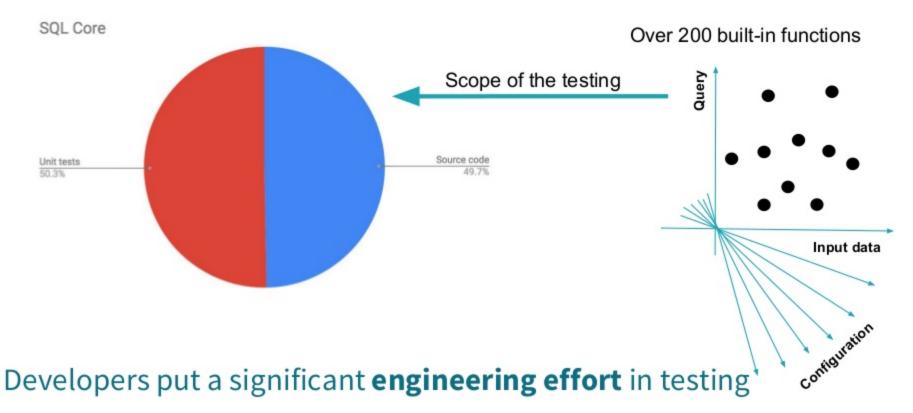
Apache Spark contributions



At this pace of development, **mistakes** are bound to happen

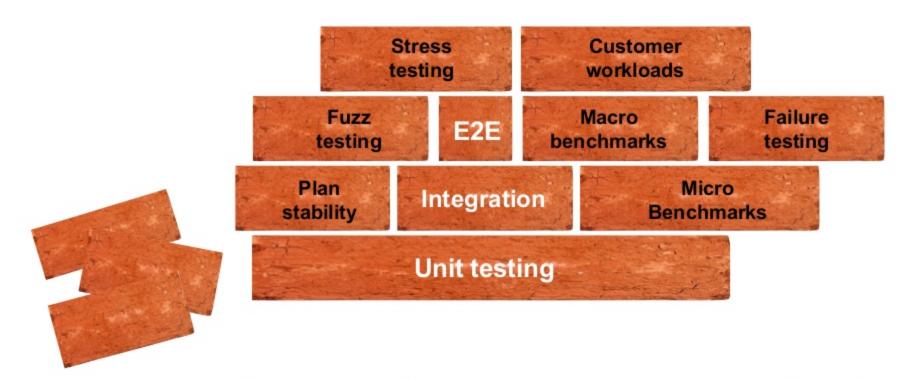


Where do these contributions go?





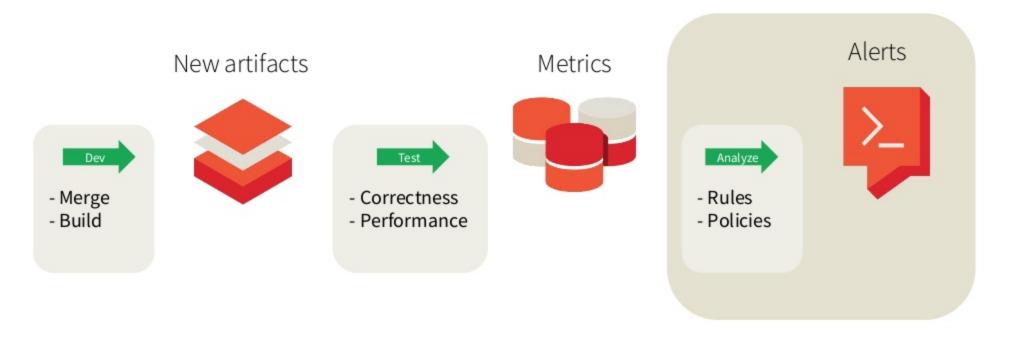
Yet another brick in the wall



Unit testing *is not enough* to guarantee correctness and performance

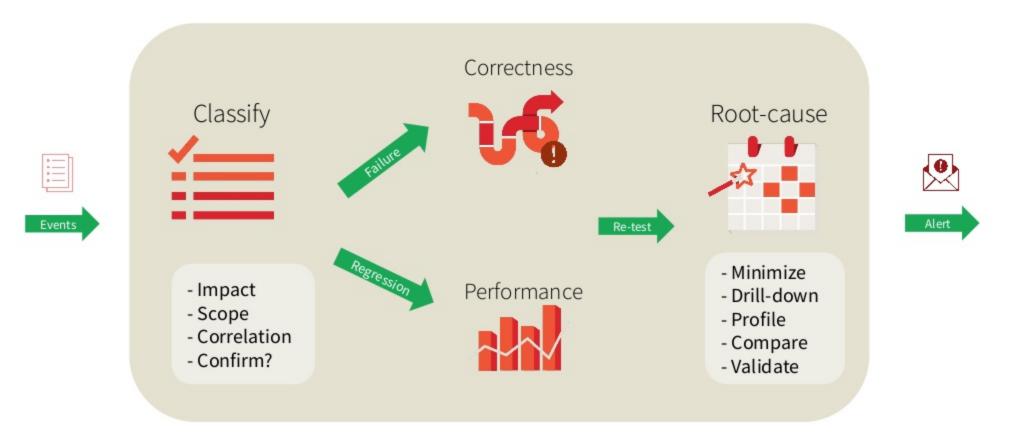


Continuous Integration pipeline



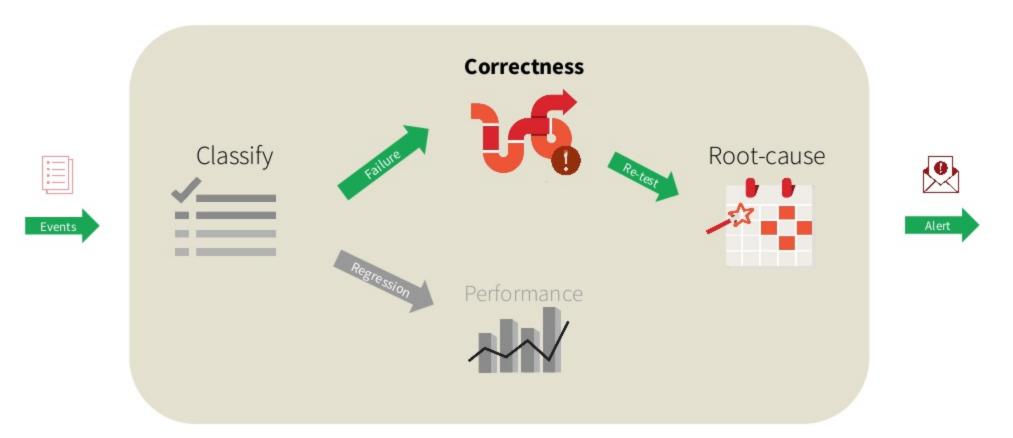


Classification and alerting

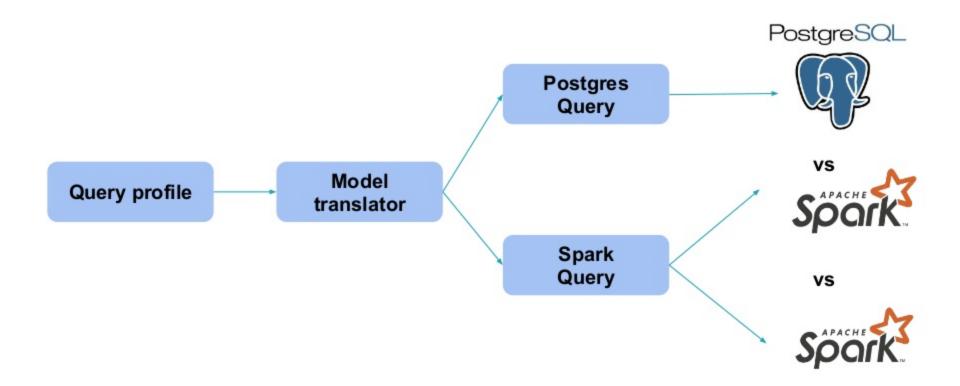




Correctness



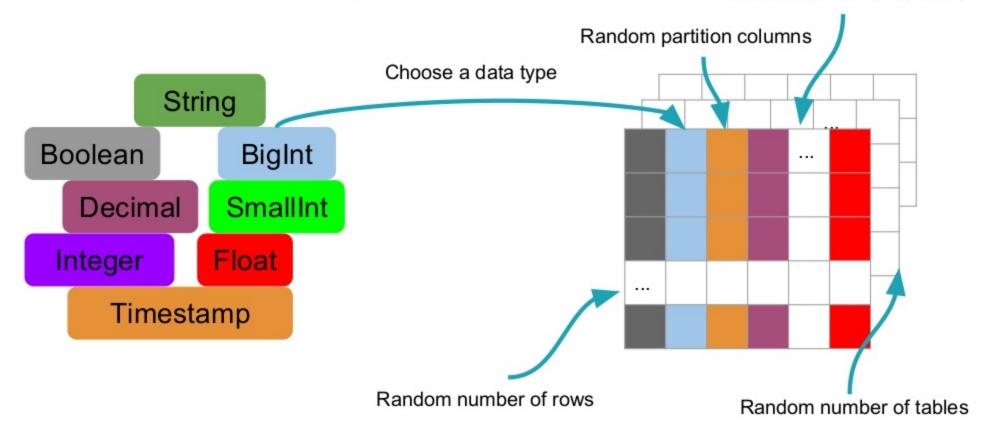
Random query generation



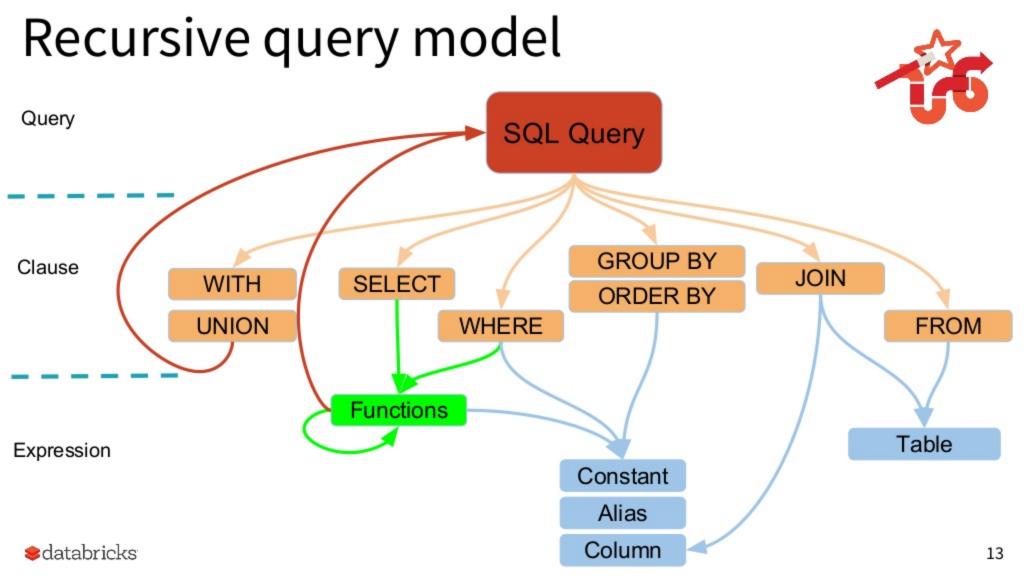


DDL and datagen

Random number of columns







Probabilistic query profile

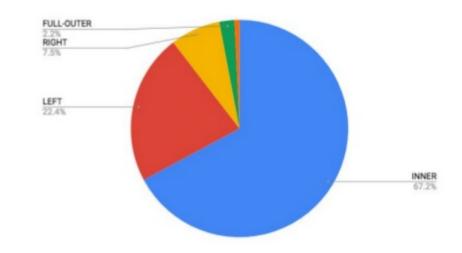
Independent weights

Optional query clauses



Inter-dependent weights

- Join types
- Select functions





Coalesce flattening (1/4)

```
SELECT COALESCE (t2.smallint_col_3, t1.smallint_col_3, t2.smallint_col_3) AS int_col,
    IF (NULL, VARIANCE (COALESCE (t2.smallint_col_3, t1.smallint_col_3, t2.smallint_col_3)),
    COALESCE (t2.smallint_col_3, t1.smallint_col_3, t2.smallint_col_3)) AS int_col_1,
    STDDEV (t2.double_col_2) AS float_col,
    COALESCE (MIN ((t1.smallint_col_3) - (COALESCE (t2.smallint_col_3, t1.smallint_col_3, t2.smallint_col_3, t2.smallint_col_3, t2.smallint_col_3, t2.smallint_col_3, t2.smallint_col_3),
    COALESCE (t2.smallint_col_3, t1.smallint_col_3, t2.smallint_col_3)) AS int_col_2

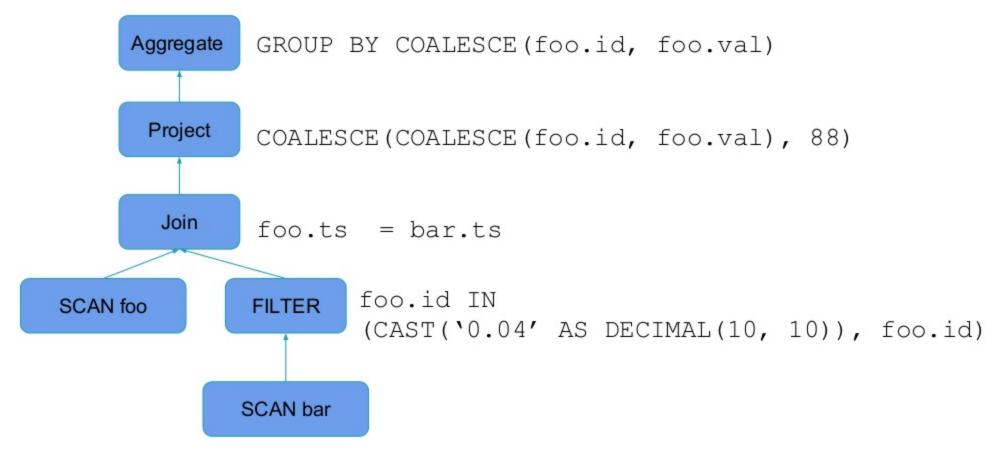
FROM table_4 t1
INNER JOIN table_4 t2 ON (t2.timestamp_col_7) = (t1.timestamp_col_7)
WHERE (t1.smallint_col_3) IN (CAST ('0.04' AS DECIMAL (10,10)), t1.smallint_col_3)
GROUP BY COALESCE (t2.smallint_col_3, t1.smallint_col_3, t2.smallint_col_3)
```

Small dataset with 2 tables of 5x5 size Within 10 randomly generated queries

Error: Operation is in ERROR_STATE

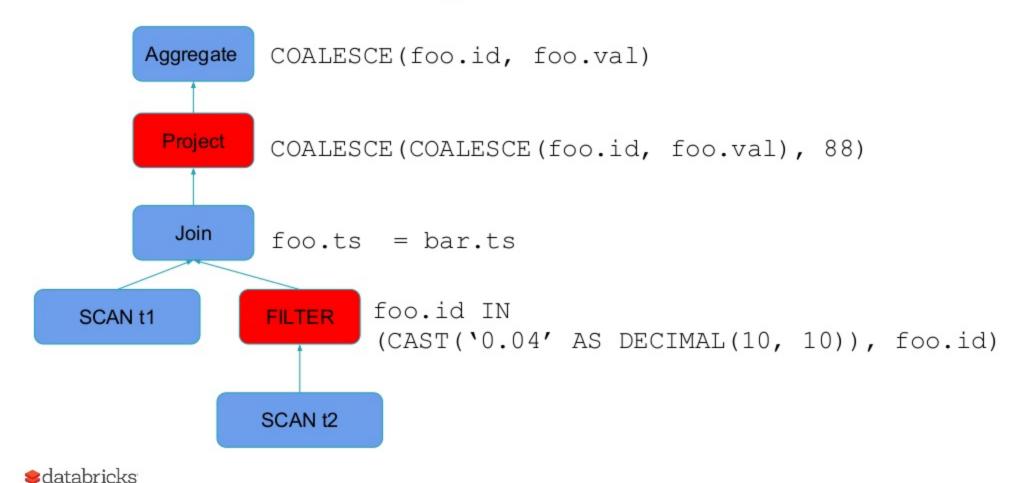


Coalesce flattening (2/3)

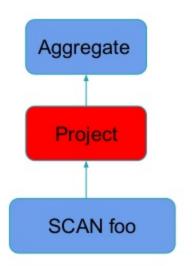




Coalesce flattening (3/4)



Coalesce flattening (4/4)



Minimized query:

```
SELECT

COALESCE (COALESCE (foo.id, foo.val), 88)

FROM foo

GROUP BY

COALESCE (foo.id, foo.val)
```

Analyzing the error

- The optimizer flattens the nested coalesce calls
- The SELECT clause doesn't contain the GROUP BY expression
- Possibly a problem with any GROUP BY expression that can be optimized



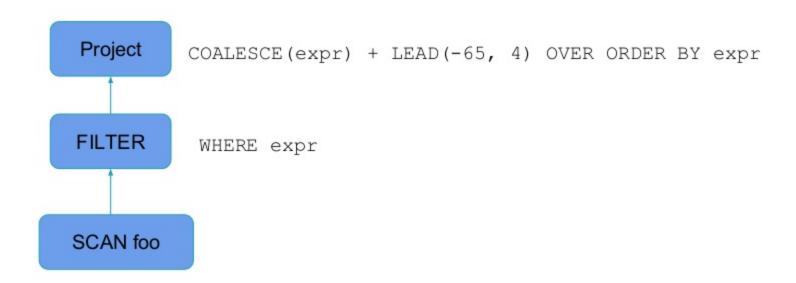
Lead function (1/3)

Error: Column 4 in row 10 does not match:

[1.0, 696, -871.81, <<-64.98>>, -349] SPARK row [1.0, 696, -871.81, <<None>>, -349] POSTGRESQL row

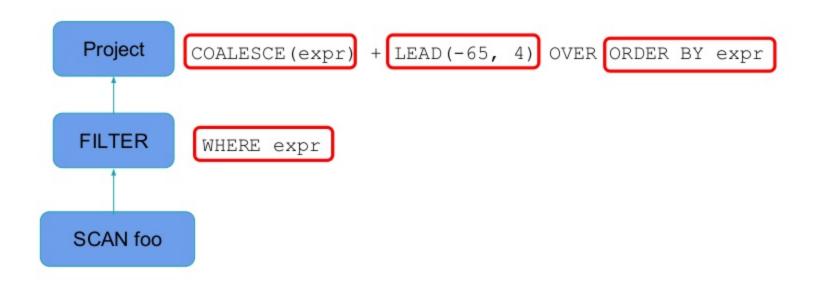


Lead function (2/3)





Lead function (3/3)

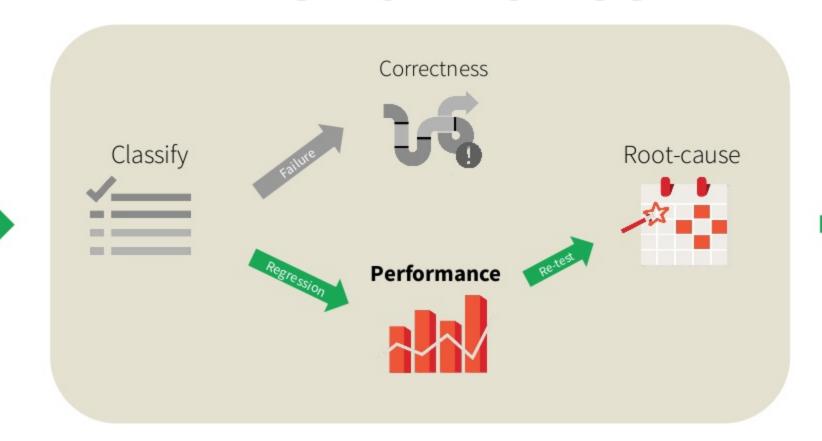


Analyzing the error

- Using constant input values breaks the behaviour of the LEAD function
- SC-16633: https://github.com/apache/spark/pull/14284



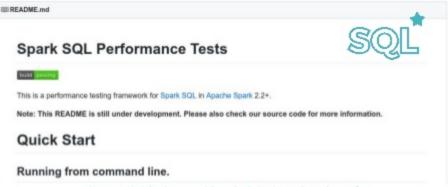
Performance



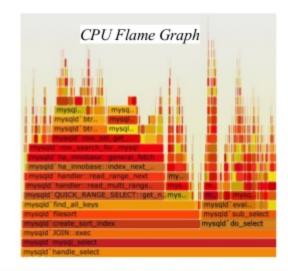


Benchmarking tools

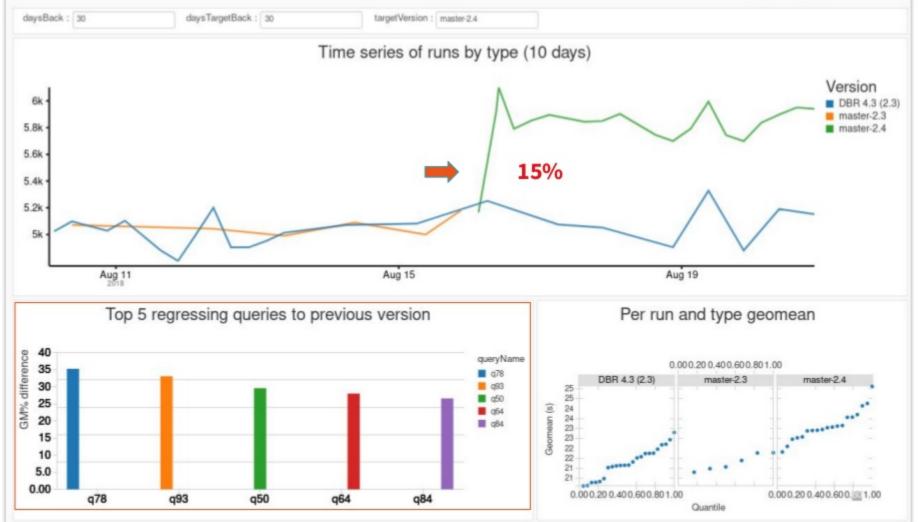
- We use spark-sql-perf public library for TPC workloads
 - Provides datagen and import scripts
 - local, cluster, S3
 - Dashboards for analyzing results
- The Spark micro benchmarks
- And the async-profiler
 - to produce flamegraphs



https://github.com/databricks/spark-sql-perf



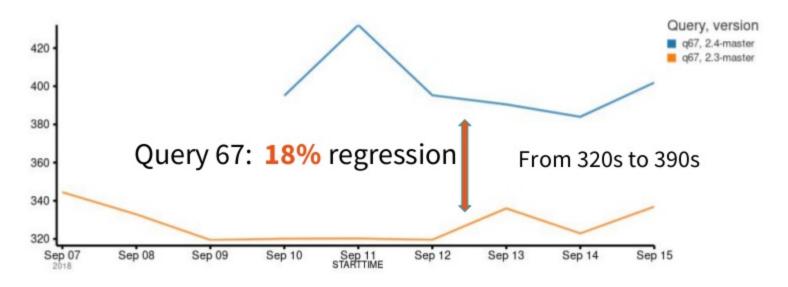
DBR 5.0-beta (v2.4) performance tracking-- journey



Per query drill-down: 67

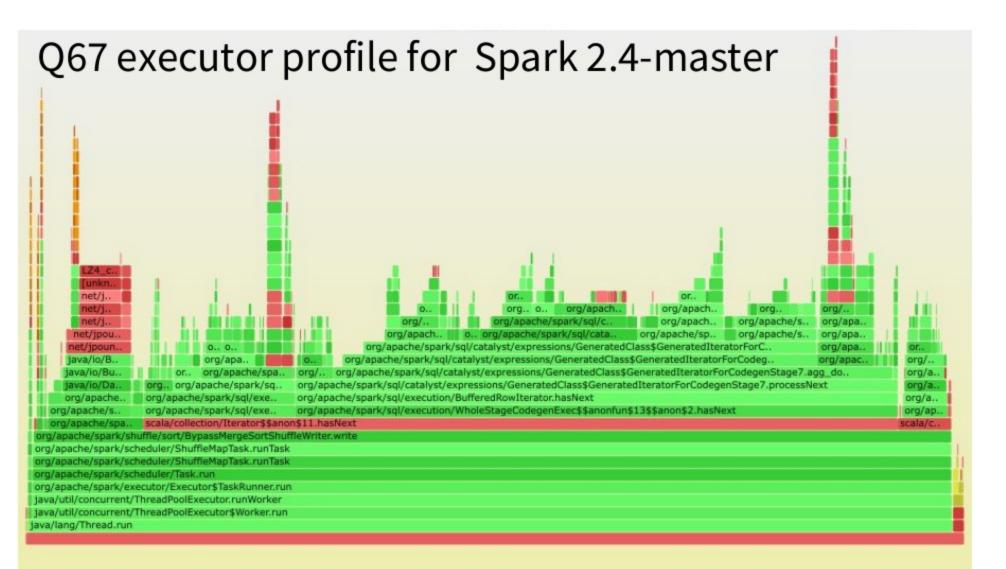


First, **scope** and **validate**



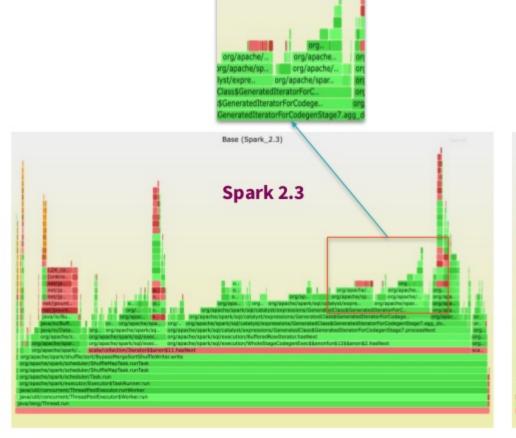
- in 2.4-master (dev) compared
- to 2.3 in DBR 4.3 (prod)

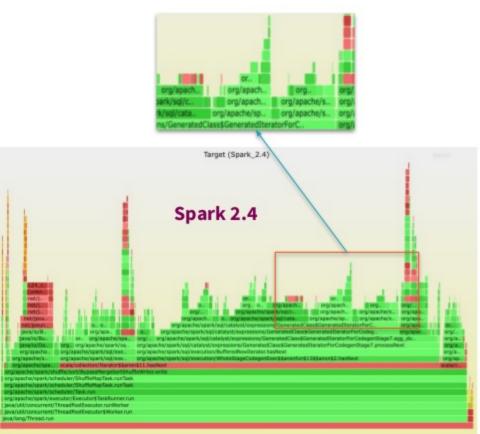




Side-by-side 2.3 vs 2.4: find the differences





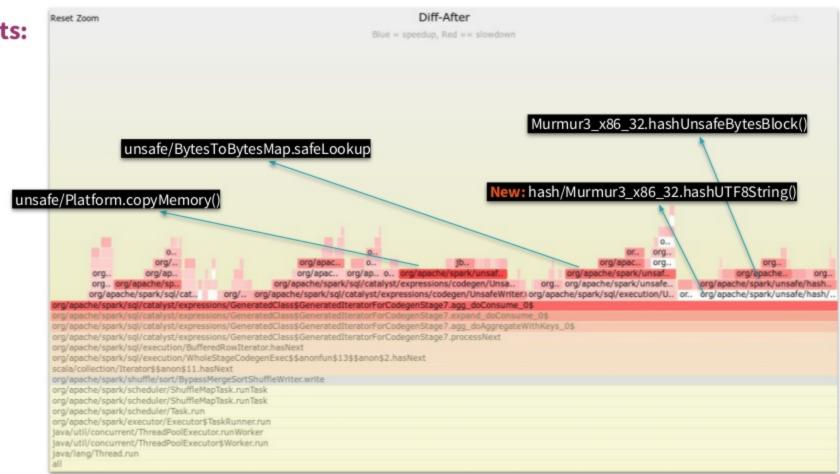


Framegraph diff zoom



Look for hints:

- Mem mgmt
- Hashing
- unsafe





Root-causing

Microbenchmark for UTF8String

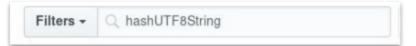
```
test("hashing") {
   import org.apache.spark.unsafe.hash.Murmur3_x86_32
   import org.apache.spark.unsafe.types.UTF8String
   val hasher = new Murmur3_x86_32(0)
   val str = UTF8String.fromString("b" * 10001)
   val numIter = 100000
   val start = System.nanoTime
   for(i <- 0 until numIter) {
      Murmur3_x86_32.hashUTF8String(str, 0)</pre>
```

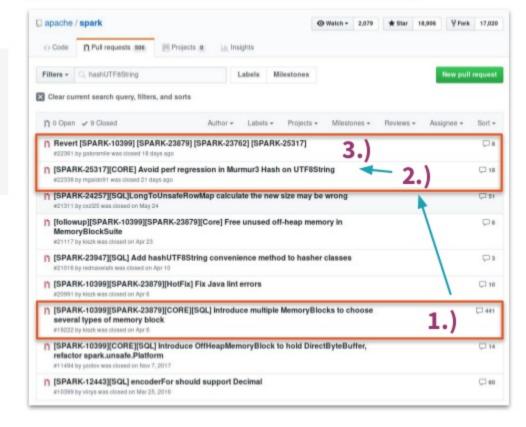
Results:

- Spark 2.3: hashUnsafeBytes() -> 40μs
- Spark 2.4 hashUnsafeBytesBlock() -> 140μs
- also slower UTF8String.getBytes()

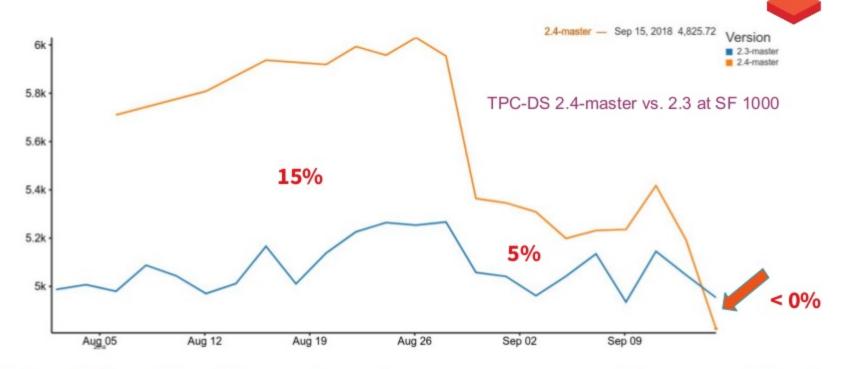


GIT BISECT





It is a journey to get a release out



DBR and Spark testing and performance are a continuous effort

Over a month effort to bring performance to improving

Conclusion

Spark in production is *not just the framework*Unit and integration testing are not enough

We need Spark specific tools to automate the process to ensure both correctness and performance



Thanks!



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databricks