

Making Apache Spark™ Better with Delta Lake

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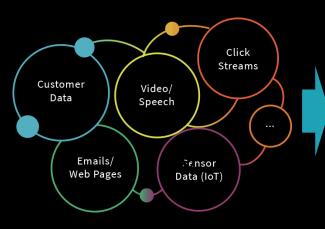


The Promise of the Data Lake

1. Collect Everything

2. Store it all in the Data Lake

3. Data Science & Machine Learning







- Recommendation Engines
- · Risk, Fraud Detection
- IoT & Predictive Maintenance
- Genomics & DNA Sequencing

Garbage In

Garbage Stored

Garbage Out



What does a typical data lake project look like?



Evolution of a Cutting-Edge Data Lake





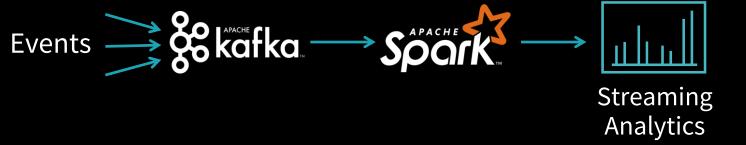








Evolution of a Cutting-Edge Data Lake

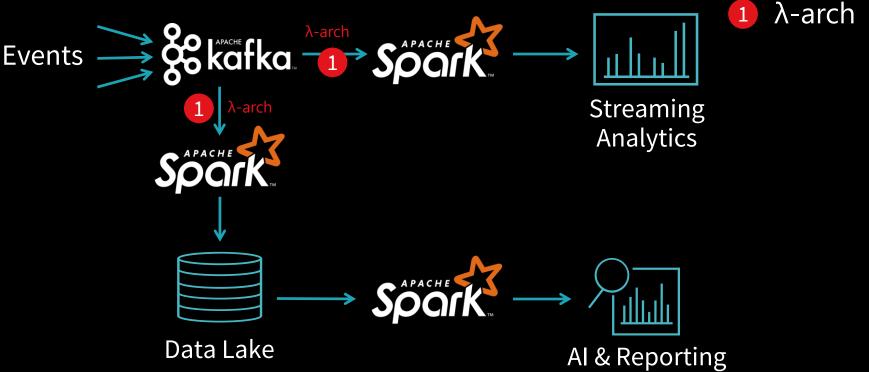






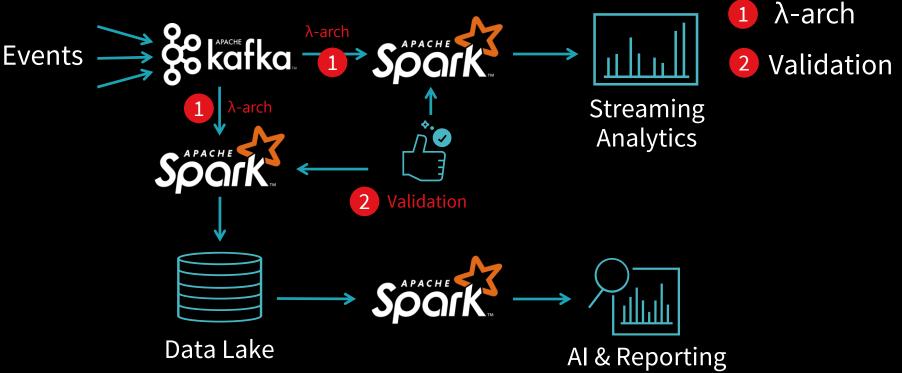


Challenge #1: Historical Queries?



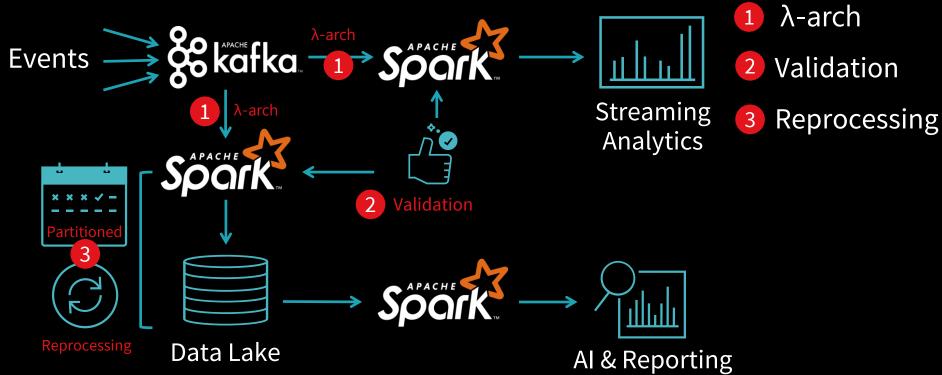


Challenge #2: Messy Data?



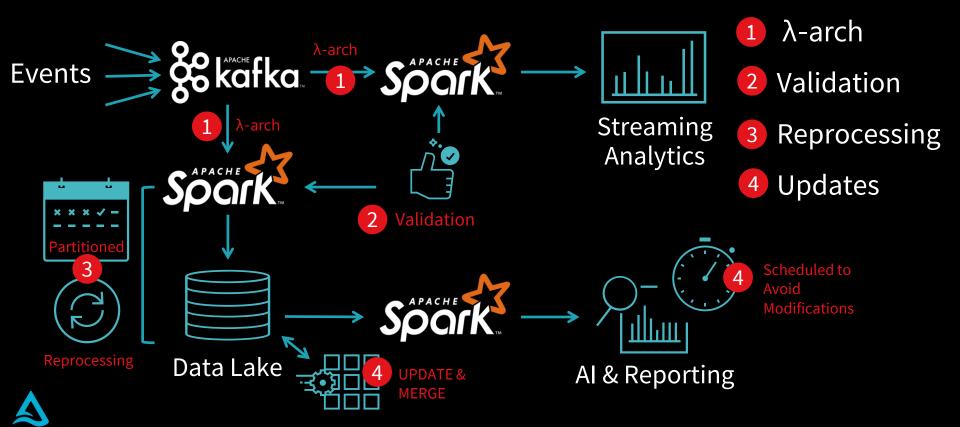


Challenge #3: Mistakes and Failures?





Challenge #4: Updates?



Wasting Time & Money

Solving Systems Problems

Instead of Extracting Value From Data



Data Lake Distractions



No atomicity means failed production jobs leave data in corrupt state requiring tedious recovery



No quality enforcement creates inconsistent and unusable data



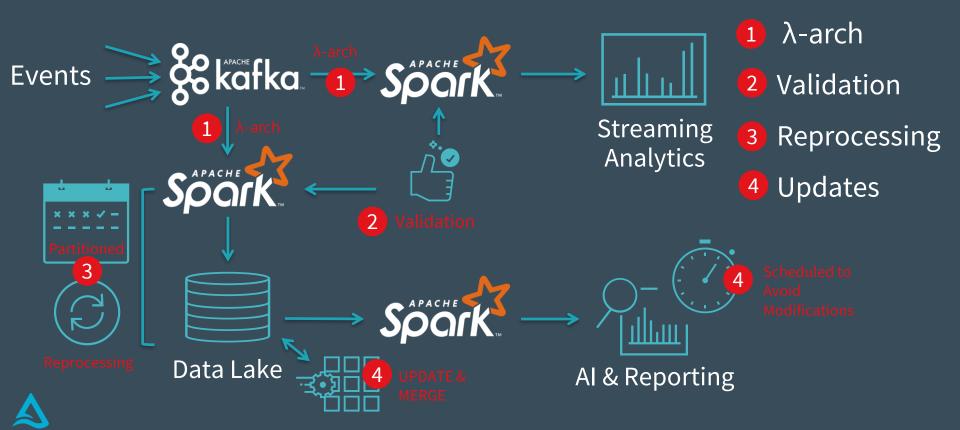
No consistency / isolation makes it almost impossible to mix appends and reads, batch and streaming

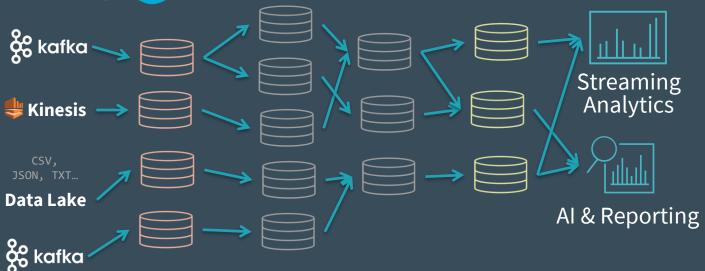


Let's try it instead with DELTA LAKE

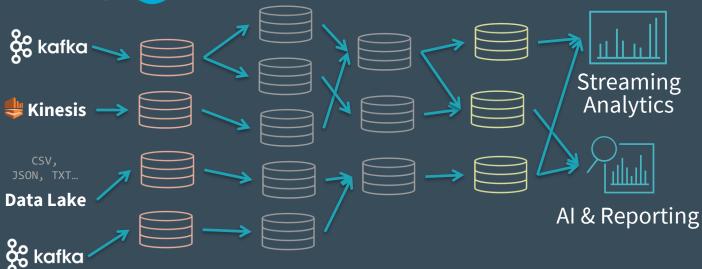


Challenges of the Data Lake





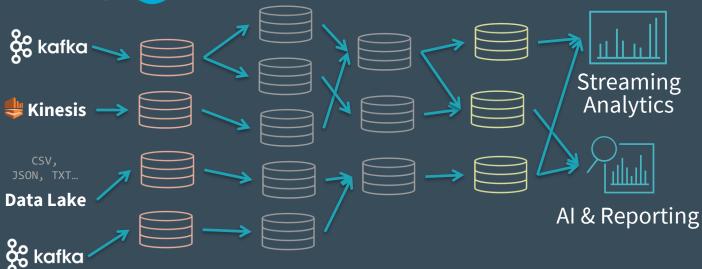




Full ACID Transaction

Focus on your data flow, instead of worrying about failures.

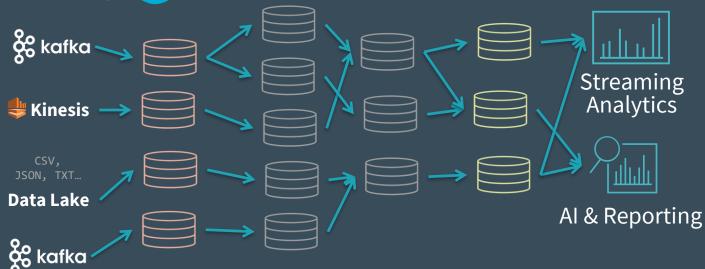




Open Standards, Open Source (Apache License)

Store petabytes of data without worries of lock-in. Growing community including Presto, Spark and more.







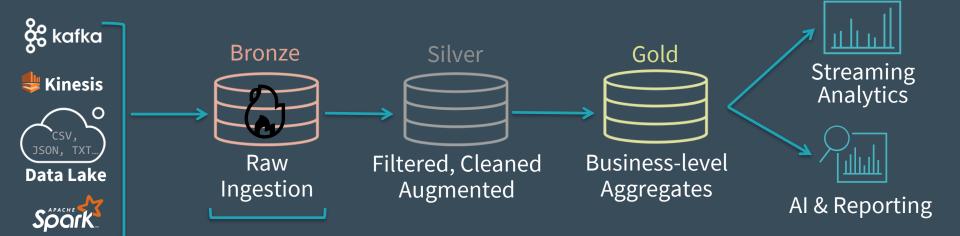
Unifies Streaming / Batch. Convert existing jobs with minimal modifications.





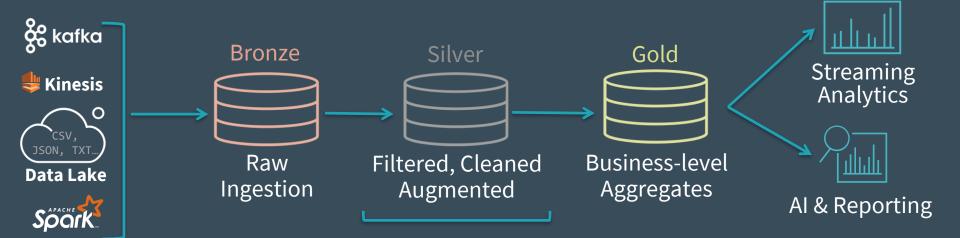
Delta Lake allows you to *incrementally* improve the quality of your data until it is ready for consumption.





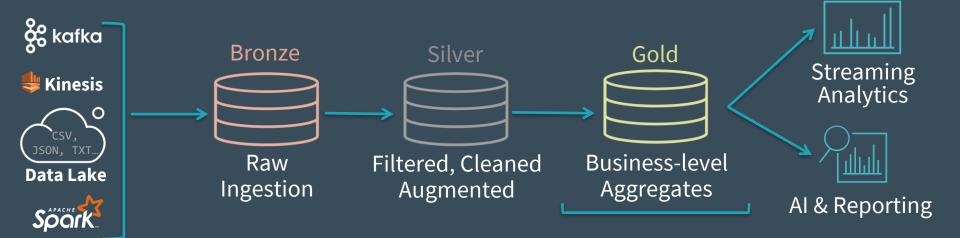
- Dumping ground for raw data
- Often with long retention (years)
- Avoid error-prone parsing





Intermediate data with some cleanup applied. Queryable for easy debugging!



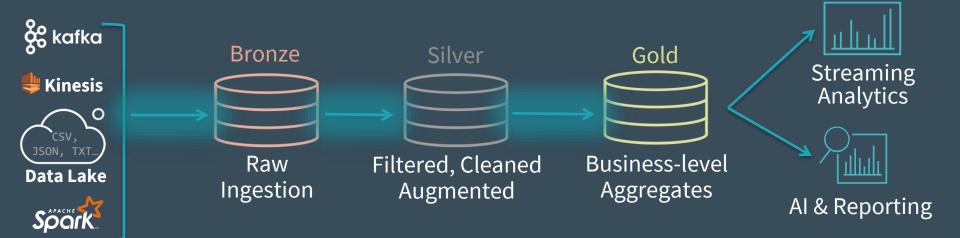


Clean data, ready for consumption.

Read with Spark or Presto*

*Coming Soon

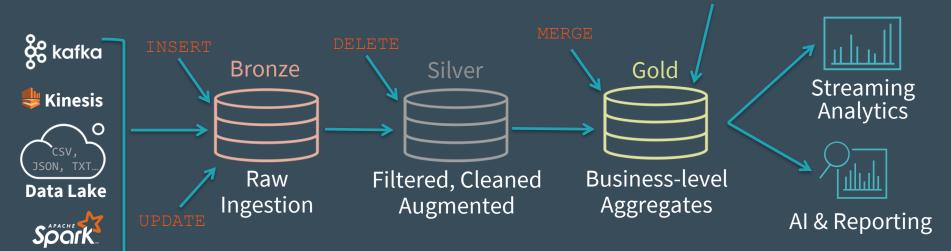




Streams move data through the Delta Lake

- Low-latency or manually triggered
- Eliminates management of schedules and jobs





Delta Lake also supports batch jobs and standard DML

Retention

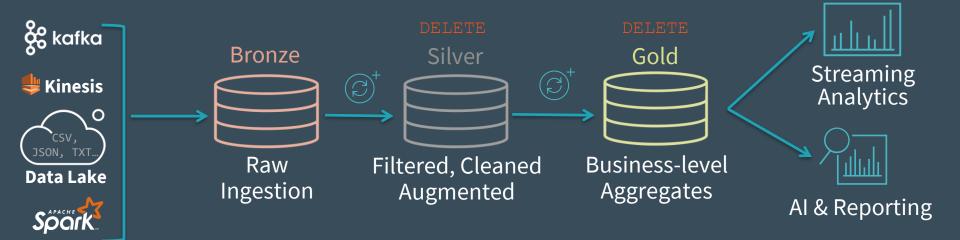
UPSERTS

Corrections

*DML Coming in 0.3.0



GDPR



Easy to recompute when business logic changes:

- Clear tables
- Restart streams



Who is using 🛕 DELTA LAKE?



Used by 1000s of organizations world wide

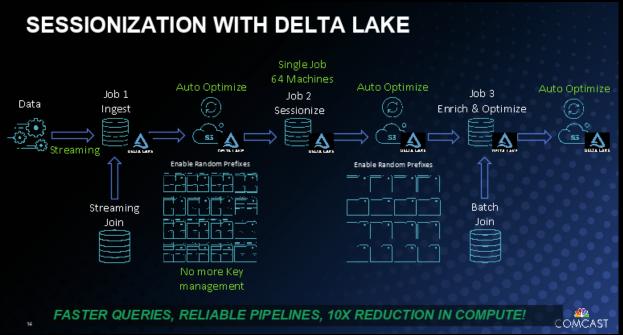
> 1 exabyte processed last month alone











Improved reliability:

Petabyte-scale jobs

10x lower compute:

640 instances to 64!

Simpler, faster ETL:

84 jobs \rightarrow 3 jobs

halved data latency



How do I use 🔔 delta lake?



Get Started with Delta using Spark APIs

Add Spark Package

```
pyspark --packages io.delta:delta-core_2.12:0.1.0
bin/spark-shell --packages io.delta:delta-core_2.12:0.1.0
```

Maven

```
<dependency>
  <groupId>io.delta</groupId>
  <artifactId>delta-core_2.12</artifactId>
  <version>0.1.0</version>
</dependency>
```

Instead of parquet...

```
dataframe
```

- .write
- .format("parquet")
- .save("/data")

... simply say delta

dataframe

- .write
- .format("delta")
- .save("/data")



Data Quality



Enforce metadata, storage, and quality declaratively.



Data Quality



Enforce metadata, storage, and quality declaratively.



How does take work?



Delta On Disk

Transaction Log
Table Versions

(Optional) Partition Directories

Data Files

```
my table/
  _delta_log/
   >00000.json
    00001.json
Ldate=2019-01-01/
  file-1.parquet
```



Table = result of a set of actions

Change Metadata – name, schema, partitioning, etc

Add File – adds a file (with optional statistics)

Remove File – removes a file

Result: Current Metadata, List of Files, List of Txns, Version



Implementing Atomicity

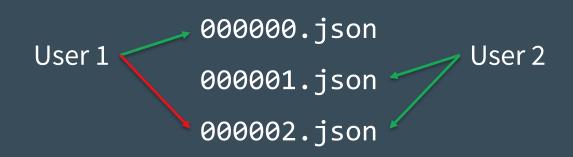
Changes to the table are stored as ordered, atomic units called commits





Ensuring Serializablity

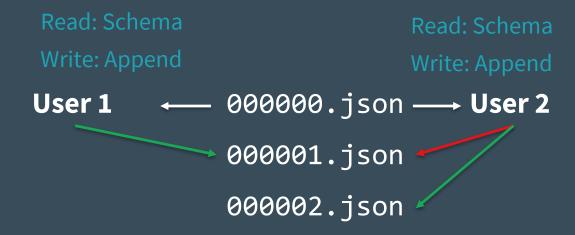
Need to agree on the order of changes, even when there are multiple writers.





Solving Conflicts Optimistically

- 1. Record start version
- 2. Record reads/writes
- 3. Attempt commit
- 4. If someone else wins, check if anything you read has changed.
- 5. Try again.





Handling Massive Metadata

Large tables can have millions of files in them! How do we scale the metadata? Use Spark for scaling!

Add 1.parquet
Add 2.parquet
Remove 1.parquet
Remove 2.parquet



Road Map

- 0.2.0 Released!
 - S3 Support
 - Azure Blob Store and ADLS Support
 SQL DML Support
- 0.3.0 (~July)
 - UPDATE (Scala)
 - DELETE (Scala)
 - MERGE (Scala)
 - VACUUM (Scala)

- Rest of Q3
 - DDL Support / Hive Metastore



Build your own Delta Lake at https://delta.io

