

Streaming on the Lakehouse

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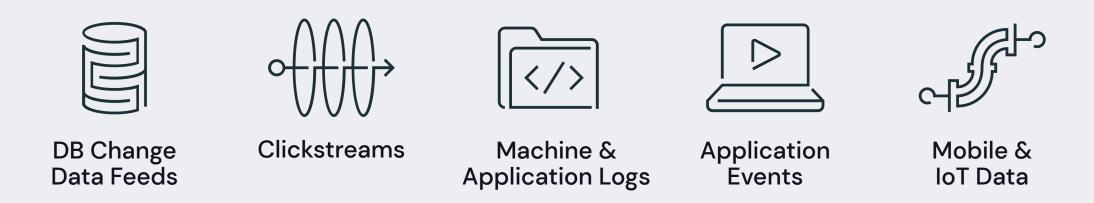
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Streaming Data

Continuously generated and unbounded data

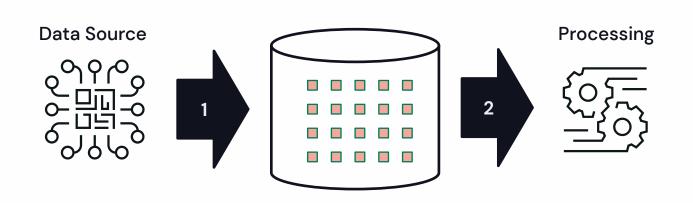




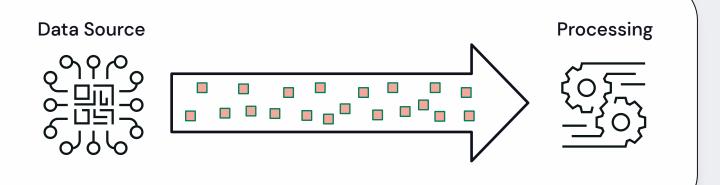
The vast majority of the data in the world is streaming data!

Stream Processing

Traditional Processing is one-off and bounded



Stream Processing is continuous and unbounded



Technical Advantages



A more intuitive way of capturing and processing continuous and unbounded data



Lower latency for time sensitive applications and use cases



Better fault-tolerance through checkpointing



Higher compute utilization and scalability through continuous and incremental processing

Business Benefits



Analytics

Fresher and faster insights

Quicker and better business decisions



Data Engineering

Sooner availability of cleaned data



More business use cases



Data Science and ML

More frequent model update and inference



Better model efficacy



Event Driven Application

Faster customized response and action



Better and differentiated customer experience



Common Misconceptions



Misconception #1



Stream processing is only for low latency use cases

```
spark.readStream
   .format("delta")
   .option("maxFilesPerTrigger", "1")
   .load(inputDir)
   .writeStream
   .trigger(Trigger.AvailableNow)
   .option("checkpointLocation",
checkpointDir)
   .start()
```

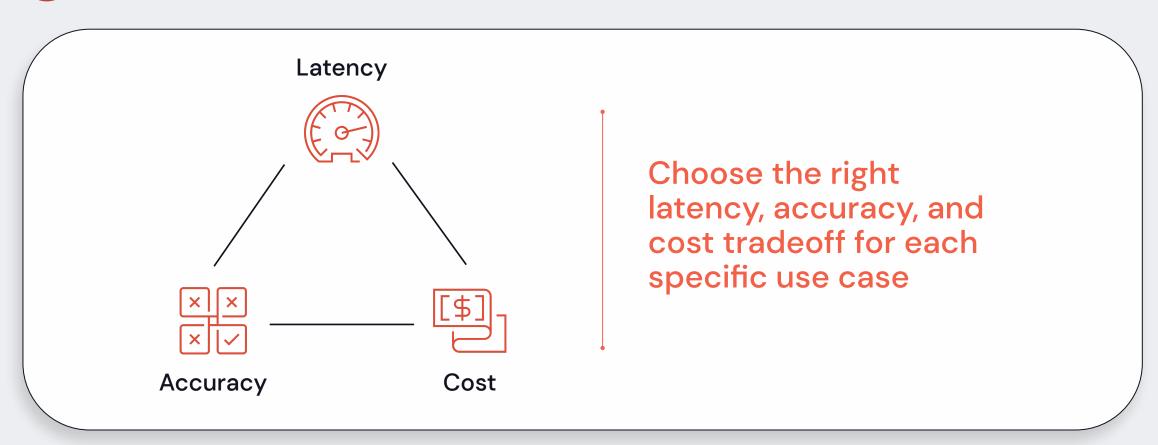
Stream processing can be applied to use cases of any latency

"Batch" is a special case of streaming

Misconception #2

X

The lower the latency, the better



Stream
Processing with
Structured
Streaming



Structured Streaming



A scalable and fault-tolerant stream processing engine built on the Spark SQL engine

Structured Streaming





Source

- Read from an initial offset position
- Keep tracking offset position as processing makes progress



Transformation

 Apply the same transformations using a normal Dataframe



Sink

- Write to a target
- Keep updating checkpoint as processing makes progress

Source

```
spark.readStream.format(<source>)
.option(<>,<>)...
.load()
```

Transformation

```
spark.readStream.format(<source>)
.option(<>>,<>)...
.load()
.select(cast("string").alias("jsonData"))
.select(from_json($"jsonData",jsonSchema).alias("payload"))
```

Sink

```
spark.readStream.format(<source>)
.option(<>>, <>)...
.load()
.select(cast("string").alias("jsonData"))
.select(from_json($"jsonData",jsonSchema).alias("payload"))
.writeStream
.format("delta")
.option("path",...)
```

Configuration

```
spark.readStream.format(<source>)
.option(<>,<>)...
.load()
.select(cast("string").alias("jsonData"))
.select(from json($"jsonData", jsonSchema).alias("payload"))
.writeStream
.format("delta")
.option("path",...)
.trigger("30 seconds")
.option("checkpointLocation",...)
.start()
```

Trigger Types

Default: Process as soon as the previous batch has been processed

Fixed interval: Process at a user-specified time interval

• One-time: Process all of the available data and then stop

Output Modes

 Append (Default): Only new rows added to the result table since the last trigger will be output to the sink

 Complete: The whole result table will be output to the sink after each trigger

 Update: Only the rows updated in the result table since the last trigger will be output to the sink

Structured Streaming Benefits



Unified Batch and Streaming

Unified API makes development and maintenance simple



High Throughput

Optimized for high throughput and low cost



Exactly Once Semantics

Fault-tolerance and exactly once semantics guarantee correctness



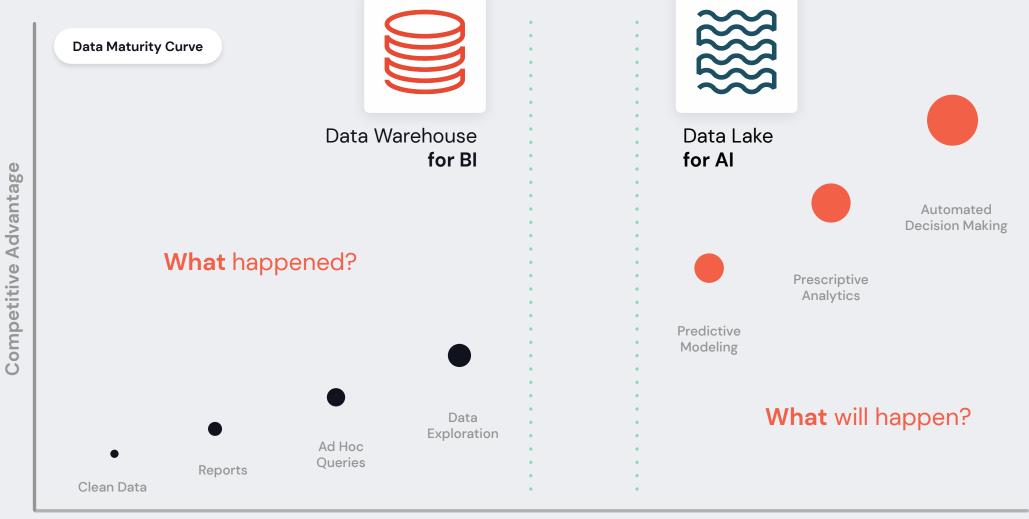
Rich Connector Ecosystem

Streaming connectors ranging from message buses to object storage services

Streaming on the Lakehouse

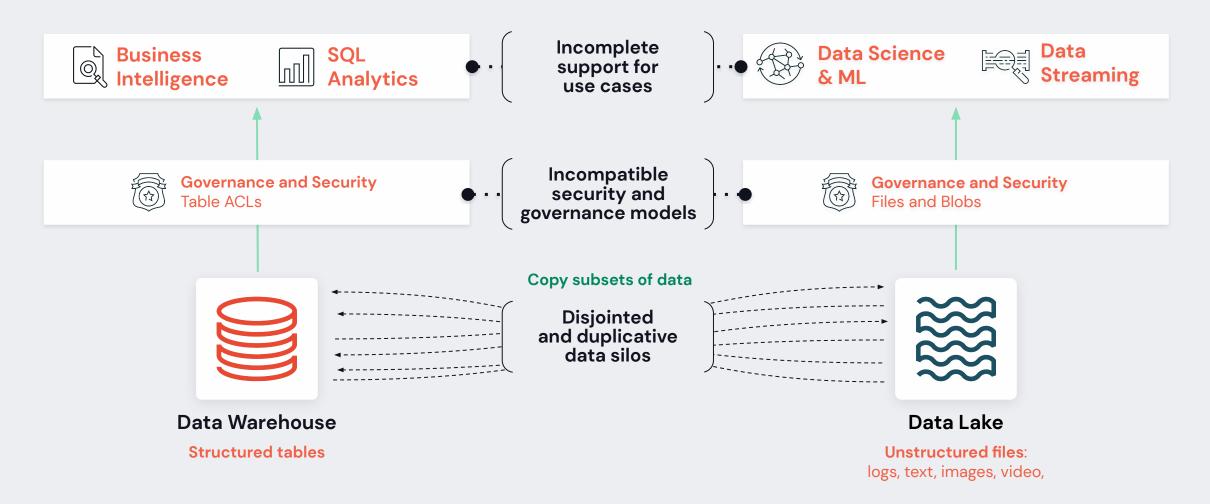


Realizing this requires two disparate, incompatible data platforms





Realizing this requires two disparate, incompatible data platforms



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Lakehouse Platform

All machine learning, SQL, BI, and streaming use cases

One security and governance approach for all data assets on all clouds

An open and reliable data platform to efficiently handle all data types



Data Lake





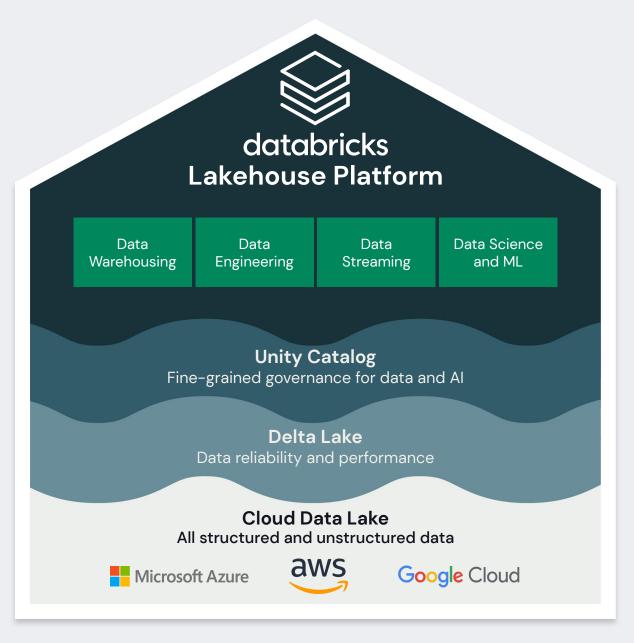


Incompatible security and governancé models



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Databricks Lakehouse Platform

Simple

Unify your data warehousing and Al use cases on a single platform

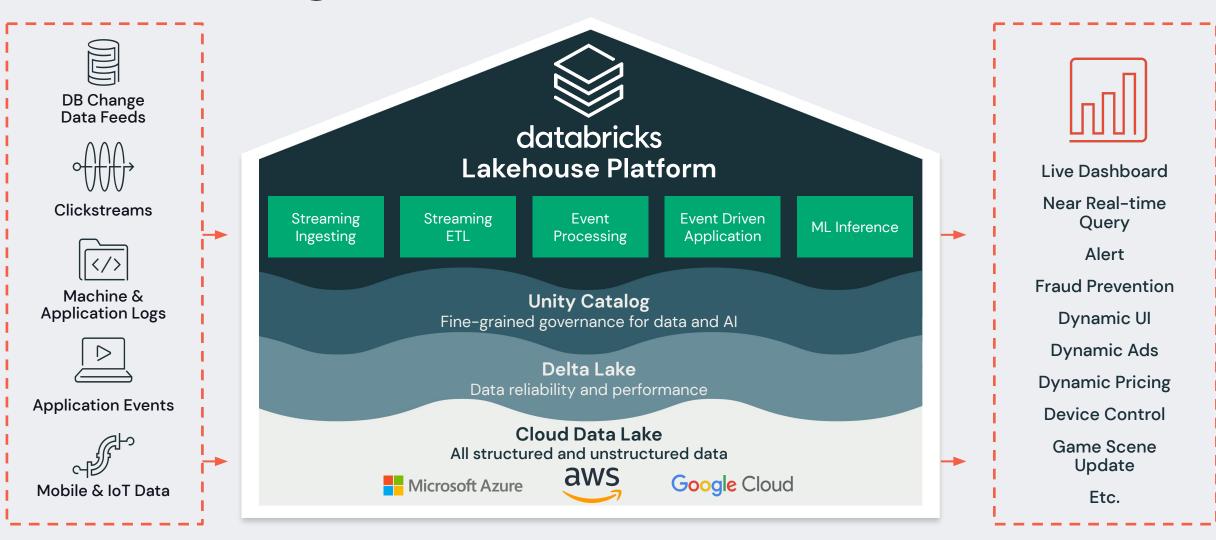
Open

Built on open source and open standards

Multicloud

One consistent data platform across clouds

Streaming on the Lakehouse



Lakehouse Differentiations



No overhead of learning, developing on, or maintaining two sets of APIs and data processing stacks



Favorite Tools

Provide diverse users
with their favorite tools
to work with streaming
data, enabling the
broader organization to
take advantage of
streaming



Optimal Cost Structure

Easily configure the right latency-cost tradeoff for each of your streaming workloads



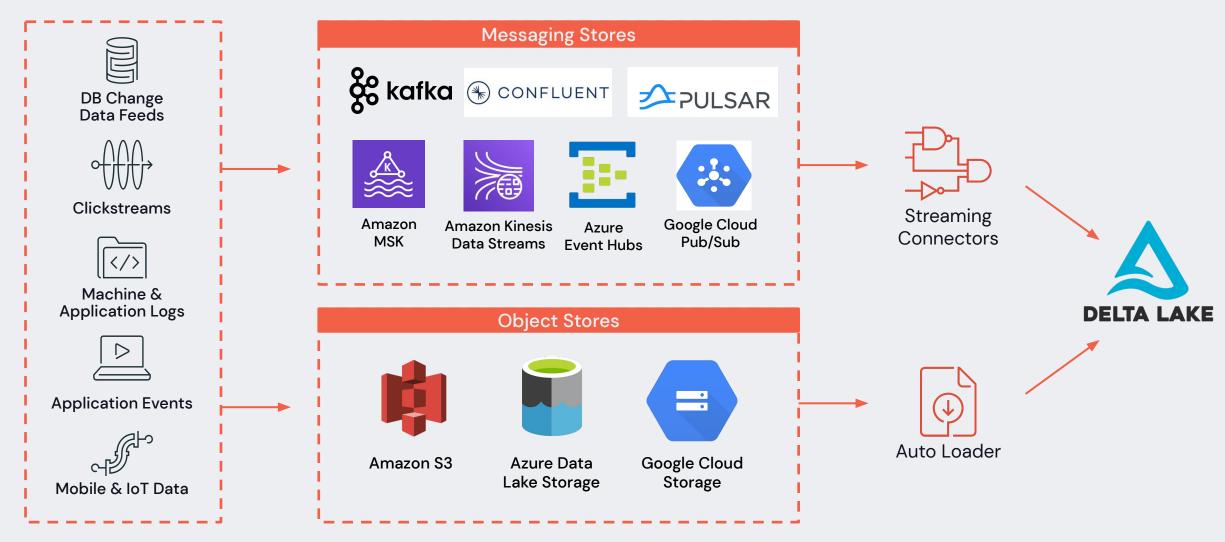
End-to-End Streaming

Has everything you need, no need to stitch together different streaming technology stacks or tune them to work together

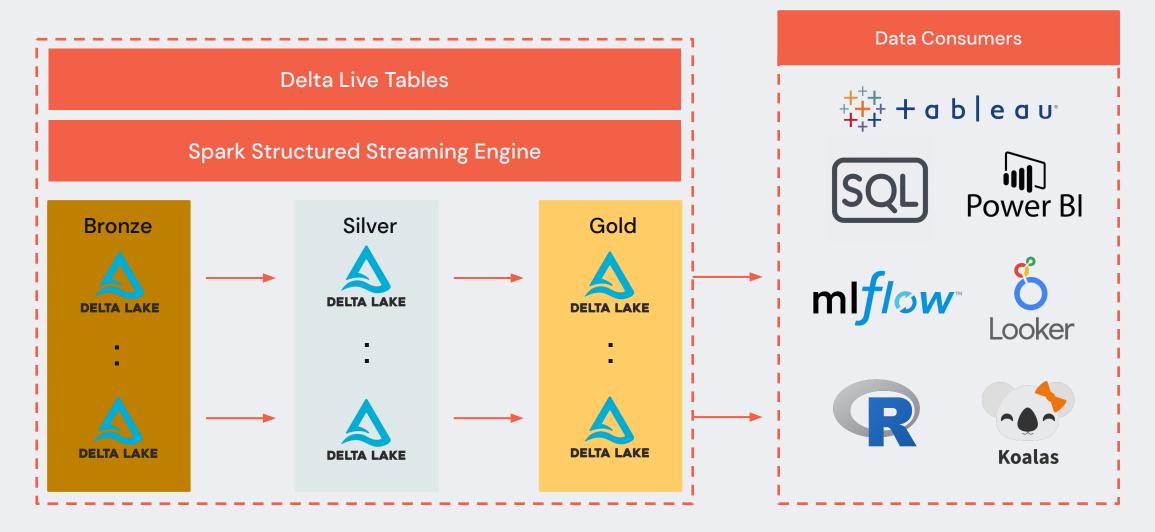
Streaming Patterns on the Lakehouse



Streaming Ingestion



Streaming ETL



Streaming ETL Choices

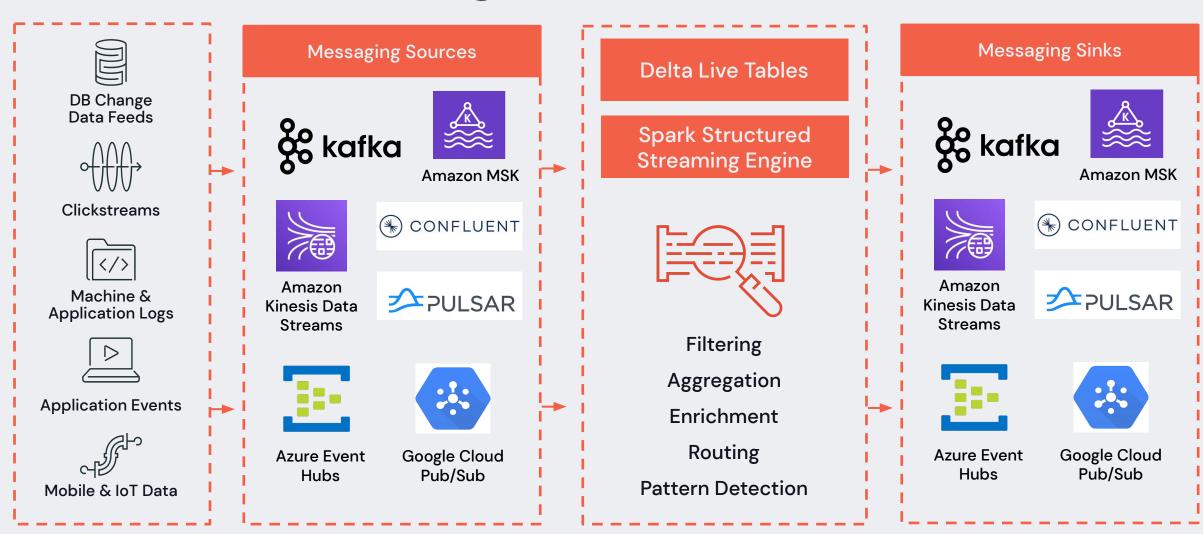
Delta Live Tables (DLT)

- Fully managed ETL service (batch + streaming) by Databricks
- The preferred way of doing streaming ETL in the Delta Lake
- Focus on ease of use
- Top choice for any new streaming ETL workloads

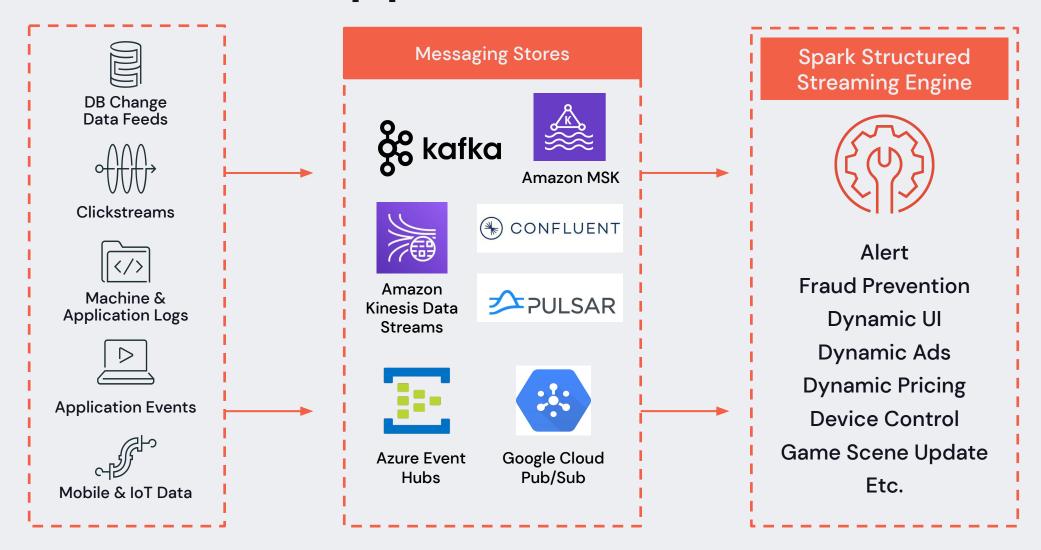
Structured Streaming

- The same Spark Structured Streaming API you have been using
- Roll your own ETL pipelines with Structured Streaming + Delta Connector + Workflow/Jobs
- Focus on flexibility
- Top choice for migrating existing Structured Streaming workloads

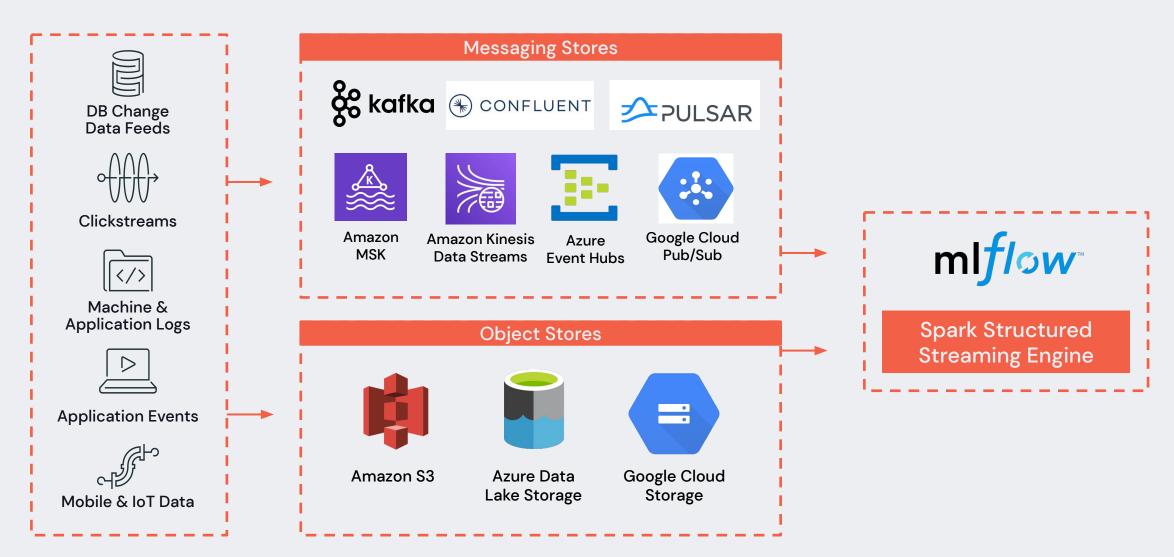
Event Processing



Event Driven Application



ML Inference



Upwork ML: Streaming Pipelines for Search & Discovery



Erica Lee
VP, ML at Upwork





Oleksii Diagiliev + Le Gu (contributors)

Engineering Manager, ML Infra + Director, ML S&D



ML Applications at Upwork

Upwork is the world's work marketplace to solve complex work serving 30% of Fortune 100 and Enterprise customers



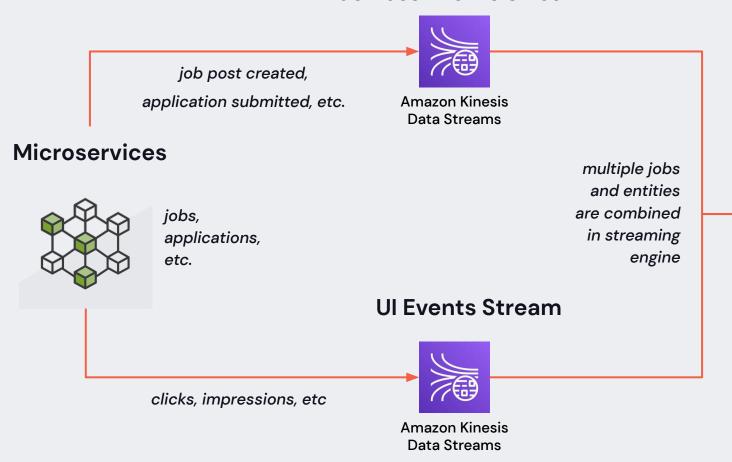


We use ML to automate and scale capabilities to our users.



Streaming Feature Computation

Business Events Stream



Streaming Feature Computation Jobs

Spark Structured Streaming Engine

job post impression sliding window counts (e.g. 1 min, 5 min, 30 mins)

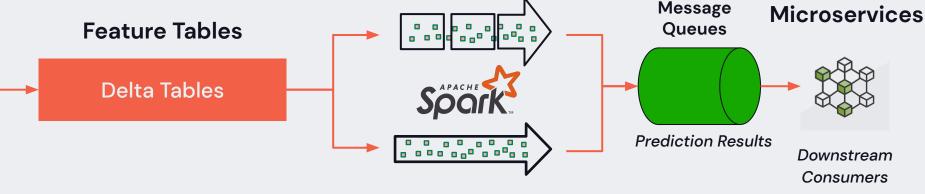




Model Serving

Spark Batch Prediction

Client Hiring State (XGBoost Classification) predicts whether client is actively hiring or not



Spark Structured Streaming Engine

Streaming Feature

Computation Jobs

Unified API for batch and stream processing



Spark Streaming Prediction

BestMatch (XGBoost Classification) ranks job applications so high quality ones are ranked first for a client

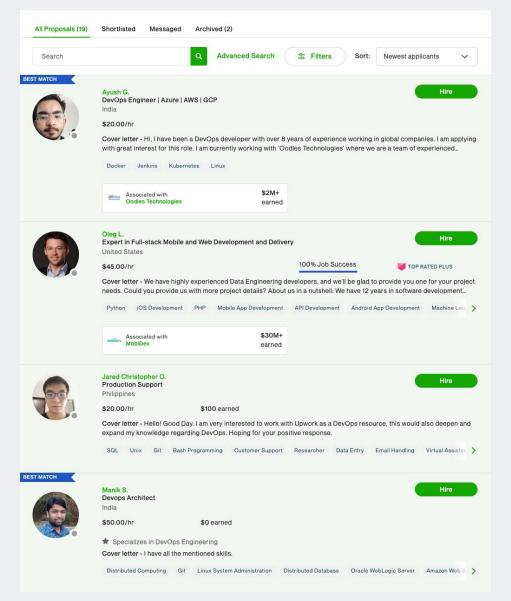


Job Value Prediction (XGBoost Regression) predicts the job post value in dollars





BestMatch Ranking Algorithm







BestMatch Ranking Algorithm





50%+ of the jobs posted on our platform receive 25+ proposal bids within 24 hours of posting.





Benefits of Databricks Platform

Benefits

- Unified platform that empowers our ML & Data and Engineers with 1 environment to run 2 workloads via Delta Tables (batch & streaming/real-time)
- Interactive and collaborative notebooks reduce dev. time (10%+)



Come Join Upwork ML!

Management

- 1 ML Manager Search & Discovery
- 1 ML Manager Trust & Safety
- 1 ML Manager Infrastructure

Individual Contributors

- 1 ML Ops Engineer Search & Discovery
- 1 Senior ML Engineer Trust & Safety





Contact Aaron White (aaronwhite@upwork.com) about openings





Thank you