

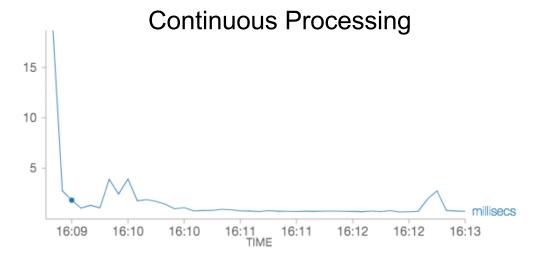
# **Continuous Processing in Structured Streaming**

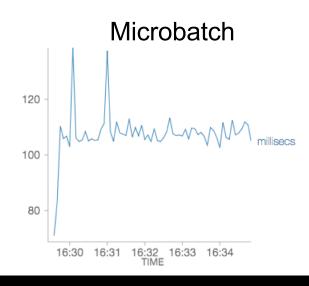
Jose Torres, Databricks

**#Dev4SAIS** 

# **Continuous Processing Overview**

- Unified Spark SQL API
- No microbatches
- Low (~1ms) latency







#### **DStream API**

- Non-declarative, similar to RDDs
- Scala/Java only
- Checkpoints only through complete snapshots
- No event time

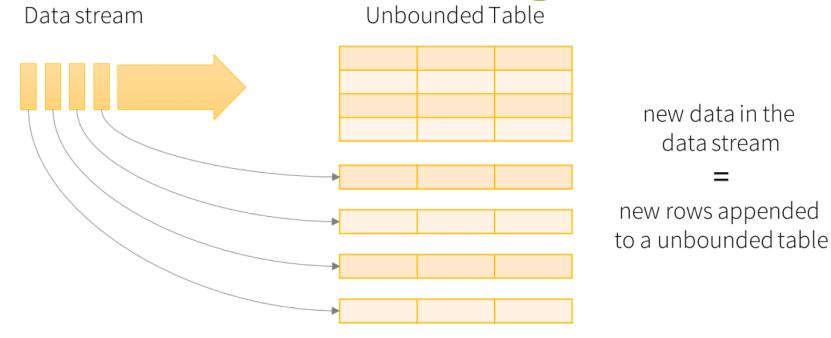


#### **Structured Streaming**

- Data represented as a virtual append-only table
- Unified Spark SQL query API
- Batch and streaming queries return same results



## **Structured Streaming**



Data stream as an unbounded table

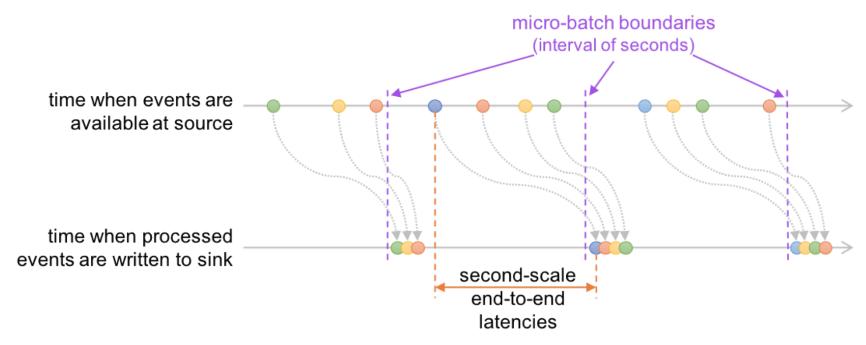


## Structured Streaming Features

- Dataframes and Datasets
- SQL, Python, and R language APIs
- Delta-based aggregation state



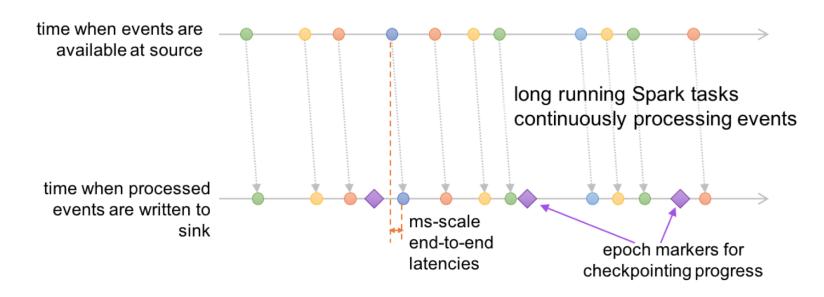
#### **Microbatches**



Second-scale end-to-end latencies with Micro-batch Processing



#### **Continuous Processing**



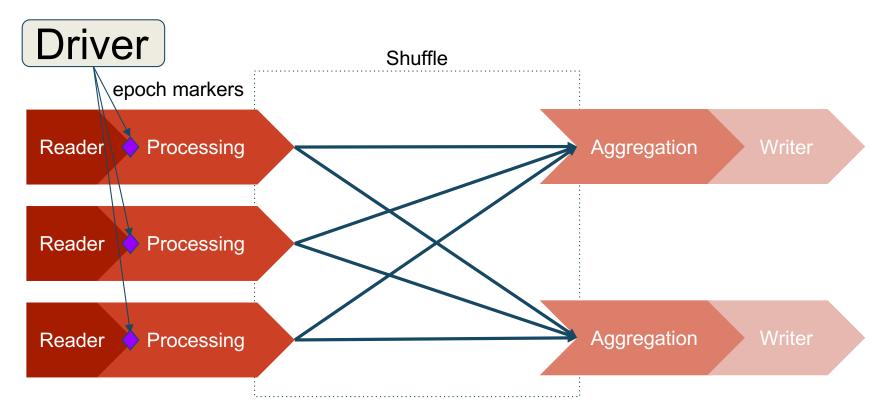
Millisecond-scale end-to-end latencies with Continuous Processing



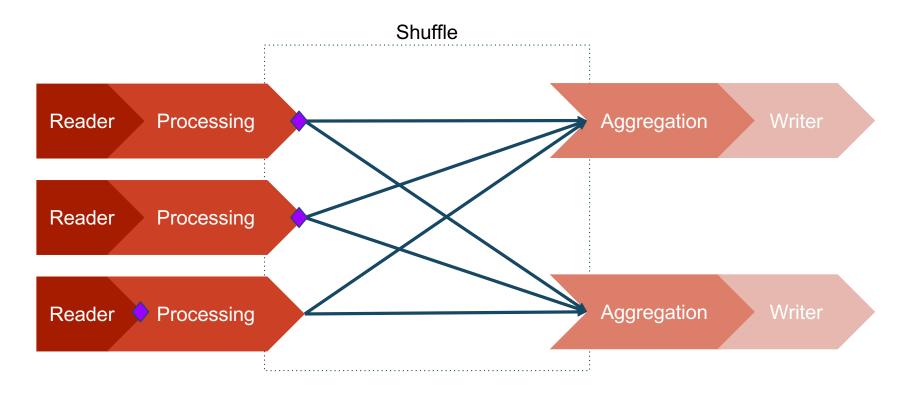
#### **Chandy-Lamport Checkpoints**

Asynchronous
Consistent
Driver checkpoint complete ready for global commit
Data Stream
Reader Aggregation Task
Save checkpoint to state store partition level commit

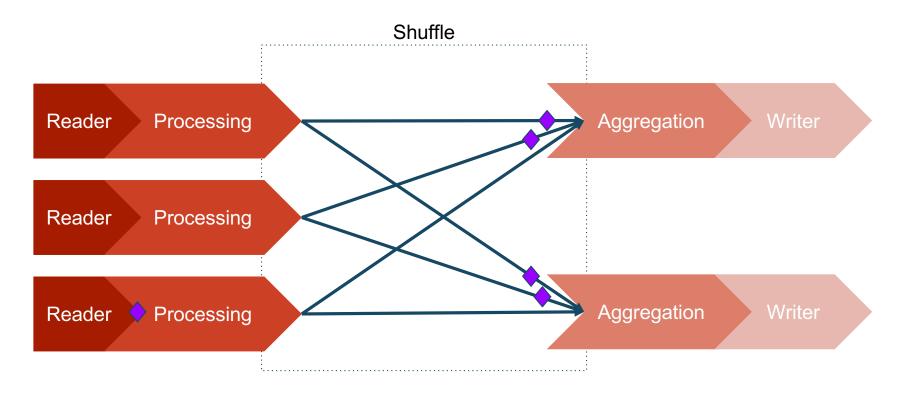




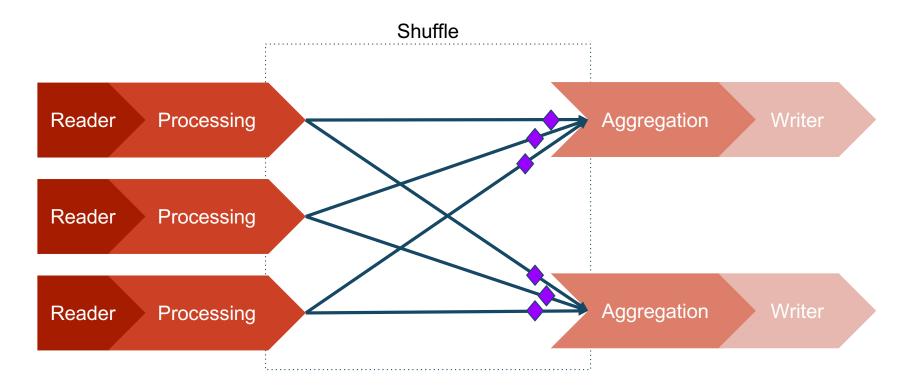




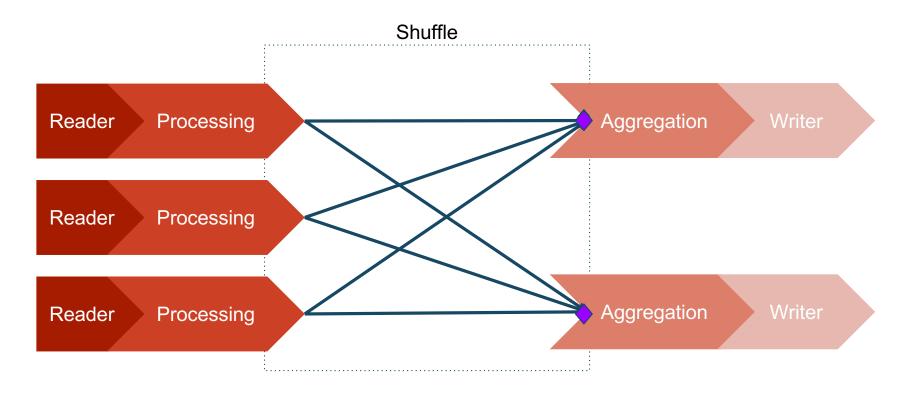




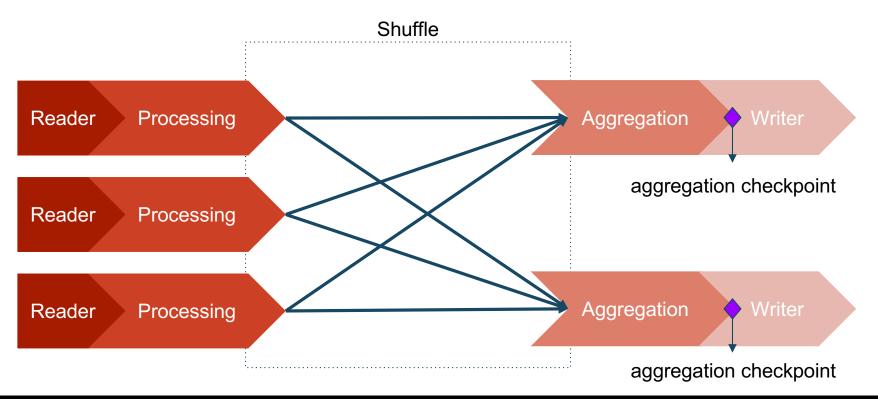




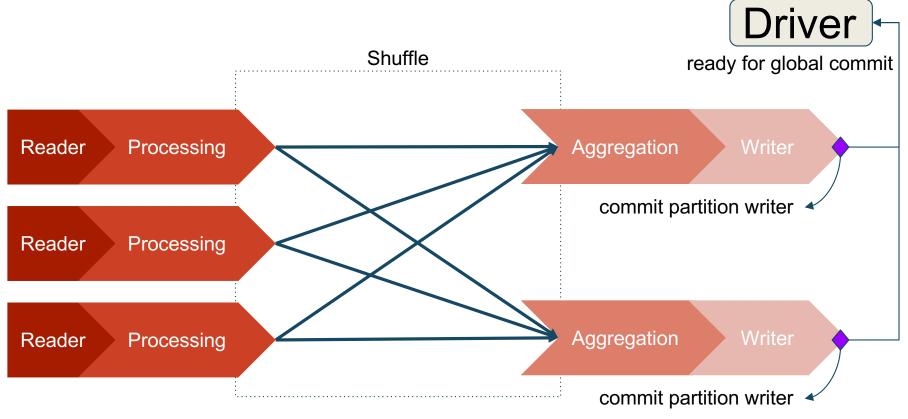














# **Continuous Processing API**

- It's just Structured Streaming
- Run the same queries in continuous mode



# **Continuous Processing in 2.3**

- Initial experimental release
- Supports ETL use cases



#### **Ongoing And Future Work**

- Shuffles (SPARK-24036)
- Event time (SPARK-24459)
- Metrics (SPARK-23887)
- Exactly-once semantics mode (SPARK-24460)
- Performance testing (TBD)
- Additional data sources (TBD)



# Q&A