## Conceptualizing the Processing Model for Apache Spark Structured Streaming

#### GETTING STARTED WITH STRUCTURED STREAMING



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#### Overview

Streaming and its place in the Apache Spark stack

**RDDs and DataFrames** 

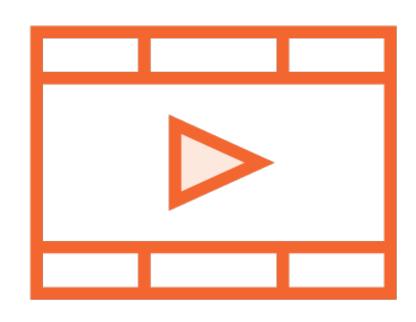
Structured Streaming in Spark 2.x

Batch as a prefix of stream

ClusterManager, SparkSession, and Executors

## Prerequisites and Course Outline

## Prerequisites

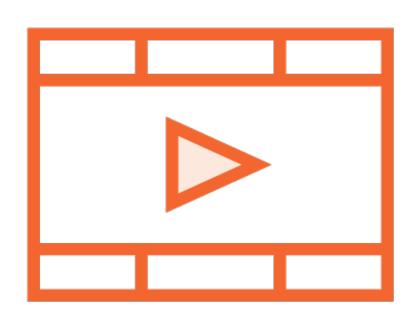


Comfortable programming in Python

Some exposure to Apache Spark and DataFrames

Exposure to streaming data would be helpful, but is not required

## Prerequisite Courses



Apache Spark Fundamentals
Getting Started with Spark 2

#### Course Outline



Getting Started with Structured Streaming

**Executing Streaming Queries** 

Understanding Scheduling and Checkpointing

**Configuring Processing Models** 

**Understanding Query Planning** 

## Introducing Apache Spark

#### Hadoop

HDFS MapReduce YARN

A file system to manage the storage of data

A framework to define a data processing task

A framework to run the data processing task

## Co-ordination Between Hadoop Blocks

MapReduce

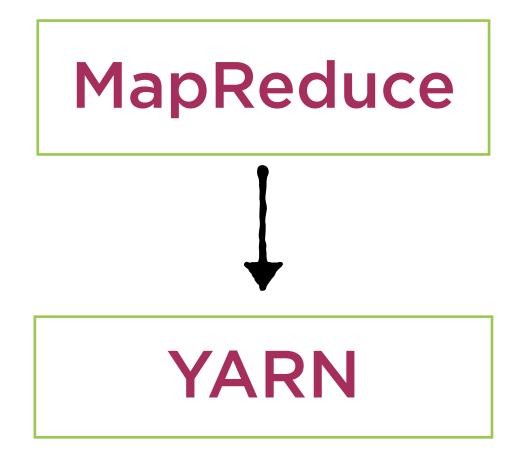


User defines map and reduce tasks using the MapReduce API

**YARN** 

**HDFS** 

#### Co-ordination Between Hadoop Blocks

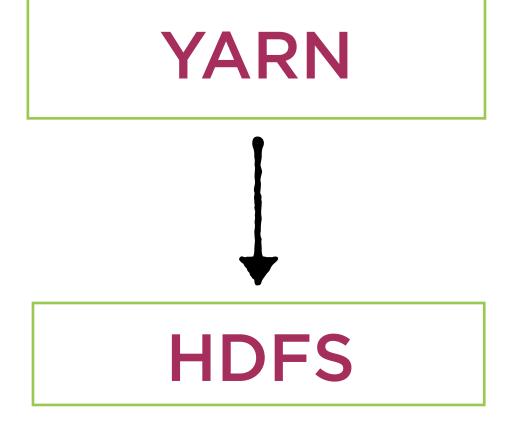


A job is triggered on the cluster

**HDFS** 

#### Co-ordination Between Hadoop Blocks

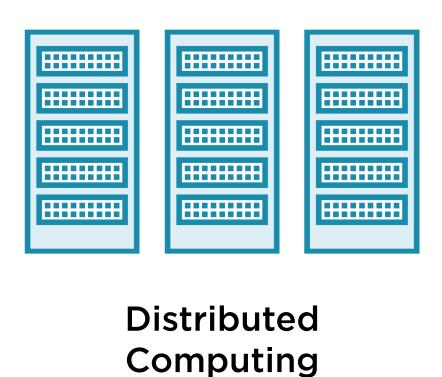
MapReduce



YARN figures out where and how to run the job, and stores the result in HDFS







An engine for data processing and analysis



**Analytics and ML on Big Data** 

Extremely powerful and popular Big Data technology

Distributed computing framework for general-purpose computing

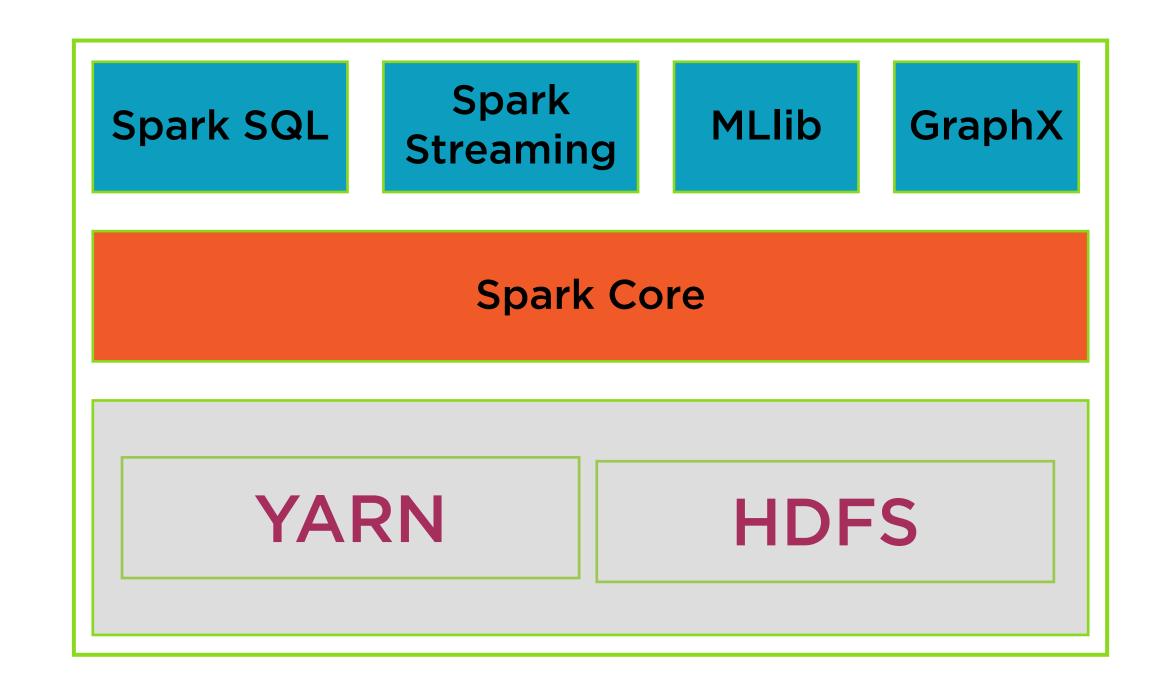
Open-source from Apache

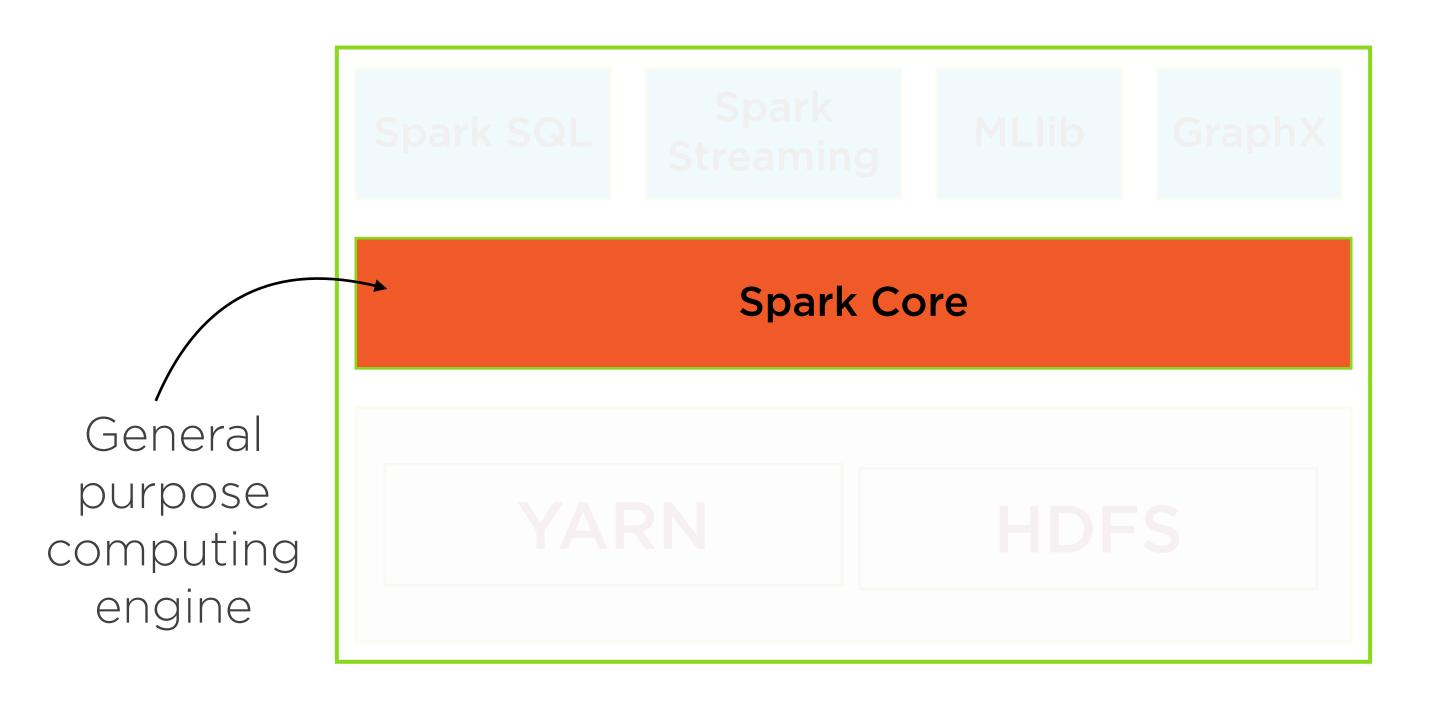
Written in Scala

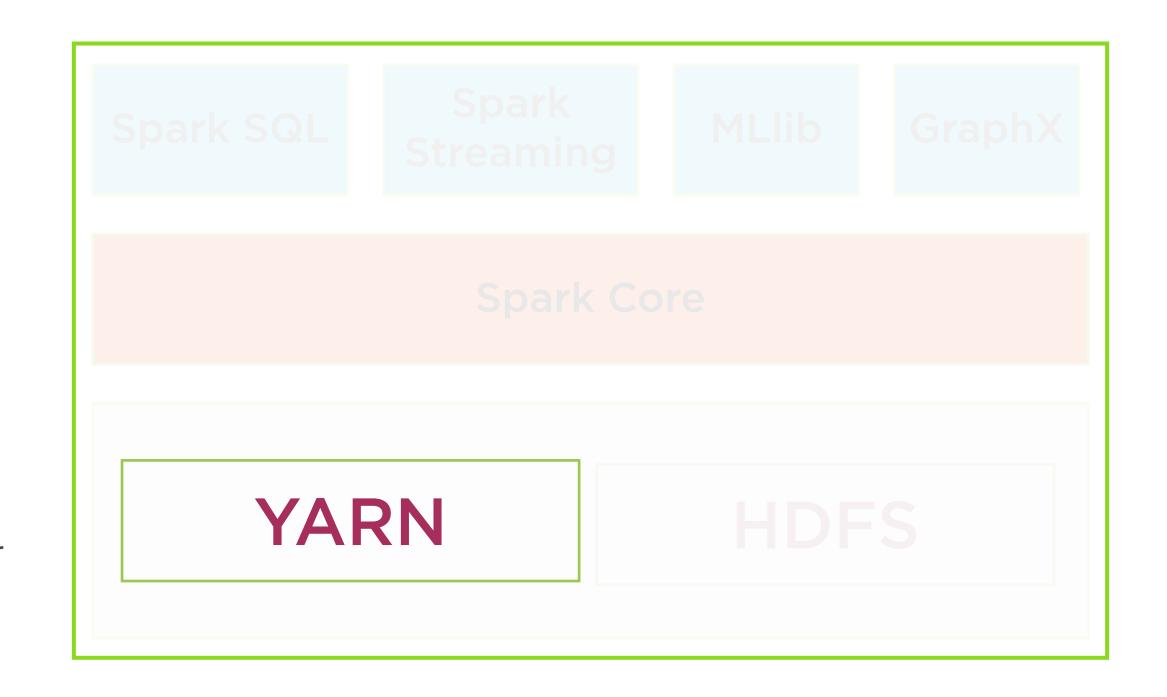


Real-time as well as batch
Interactive REPL environment
Read-Evaluate-Print-Loop
Support for multiple programming languages

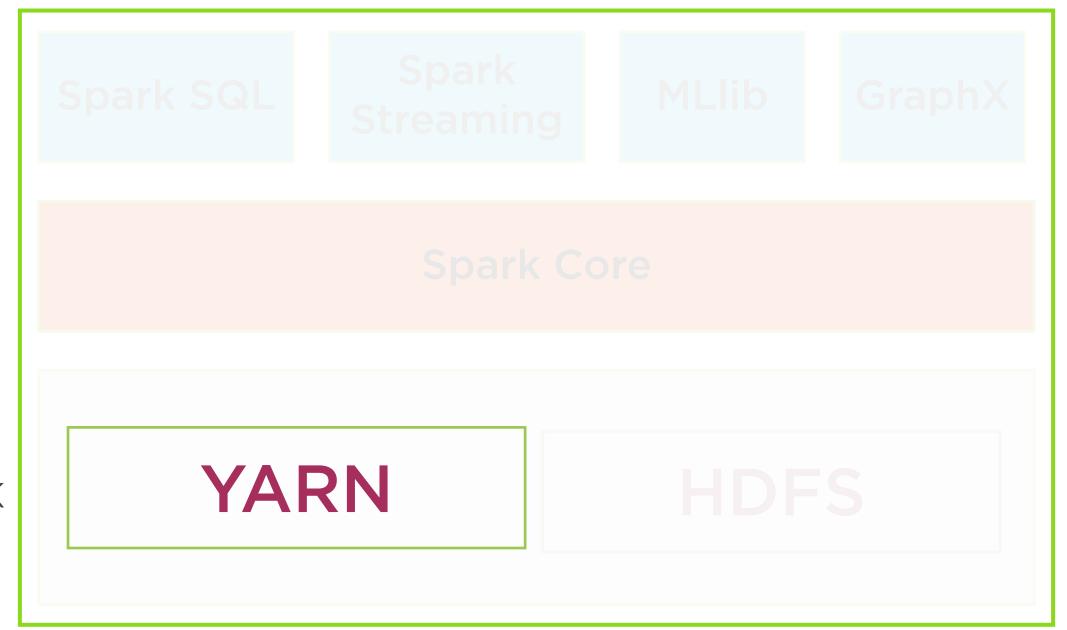
- Python, Scala, R, ...



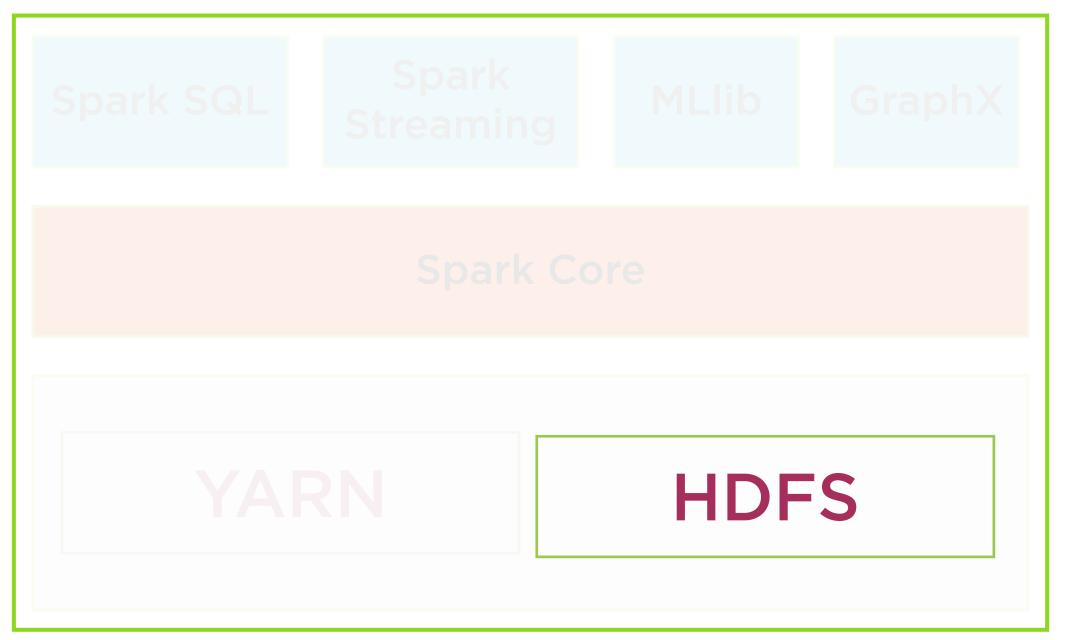




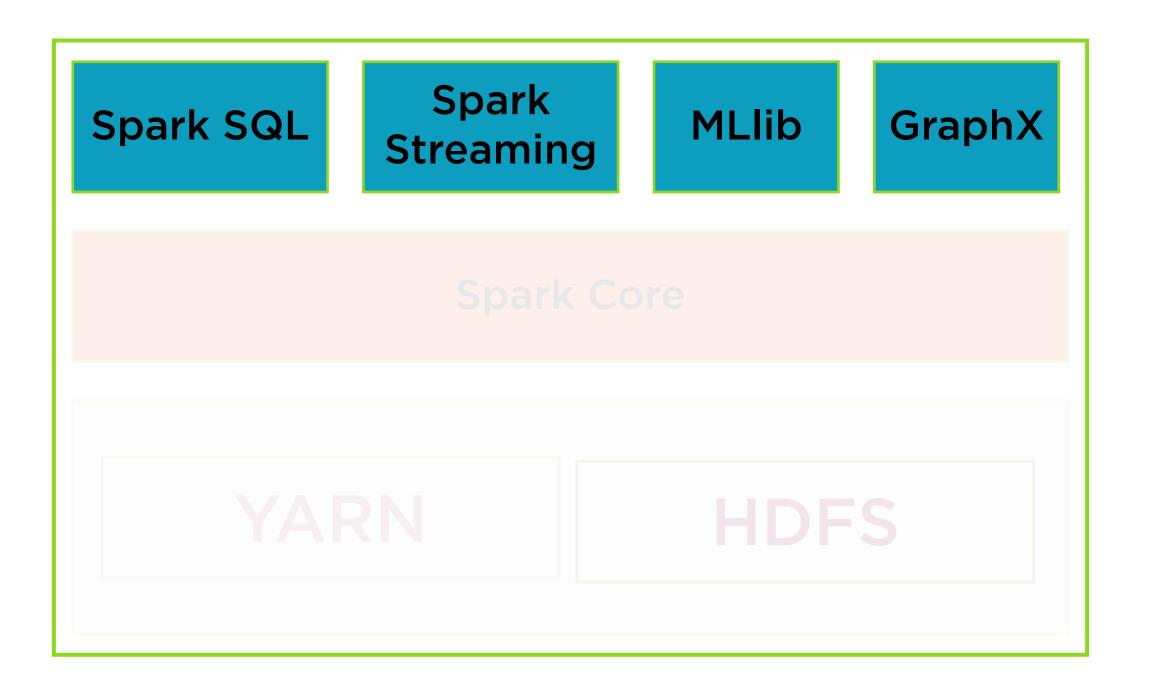
Cluster manager



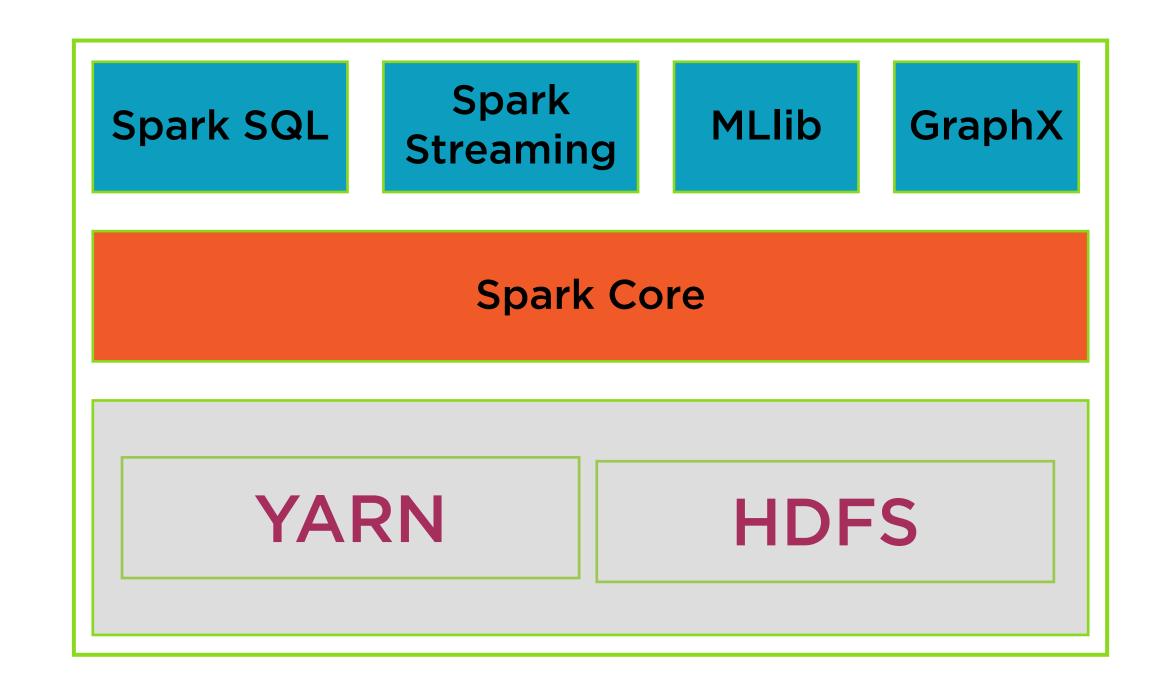
Alternatives: Mesos, Spark Standalone, Kubernetes



Distributed Storage system

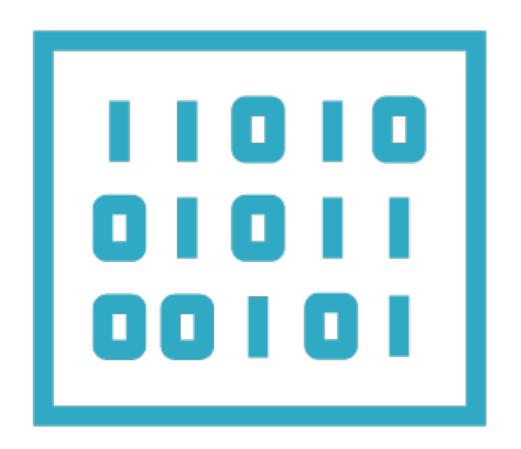


Spark libraries



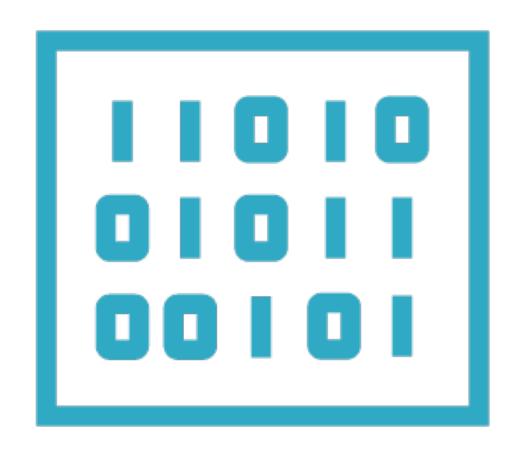
#### RDDs and DataFrames

#### Resilient Distributed Datasets



All operations in Spark are performed on in-memory objects

#### Resilient Distributed Datasets



An RDD is a collection of entities - rows, records, an RDD is the basic data structure used in Spark 1.x

## Why is this relevant in Spark 2?

## RDDs are still the fundamental building blocks of Spark

#### Characteristics of RDDs

#### **Partitioned**

RDDs are split across nodes in a cluster

#### Immutable

RDDs, once created, cannot be changed

#### Resilient

Can be reconstructed on node crashes

#### RDDs Support Two Operations

#### **Transformation**

Transform input RDDs into another RDD

#### Action

Request a result, to a file, to console window

#### Lazy Evaluation

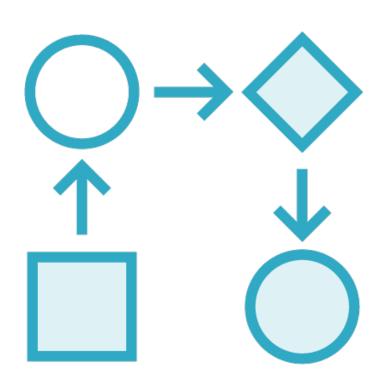


Spark keeps a record of the series of transformations

Transformations are not performed when defined

Transformations are materialized only when the user requests a result

#### Lineage

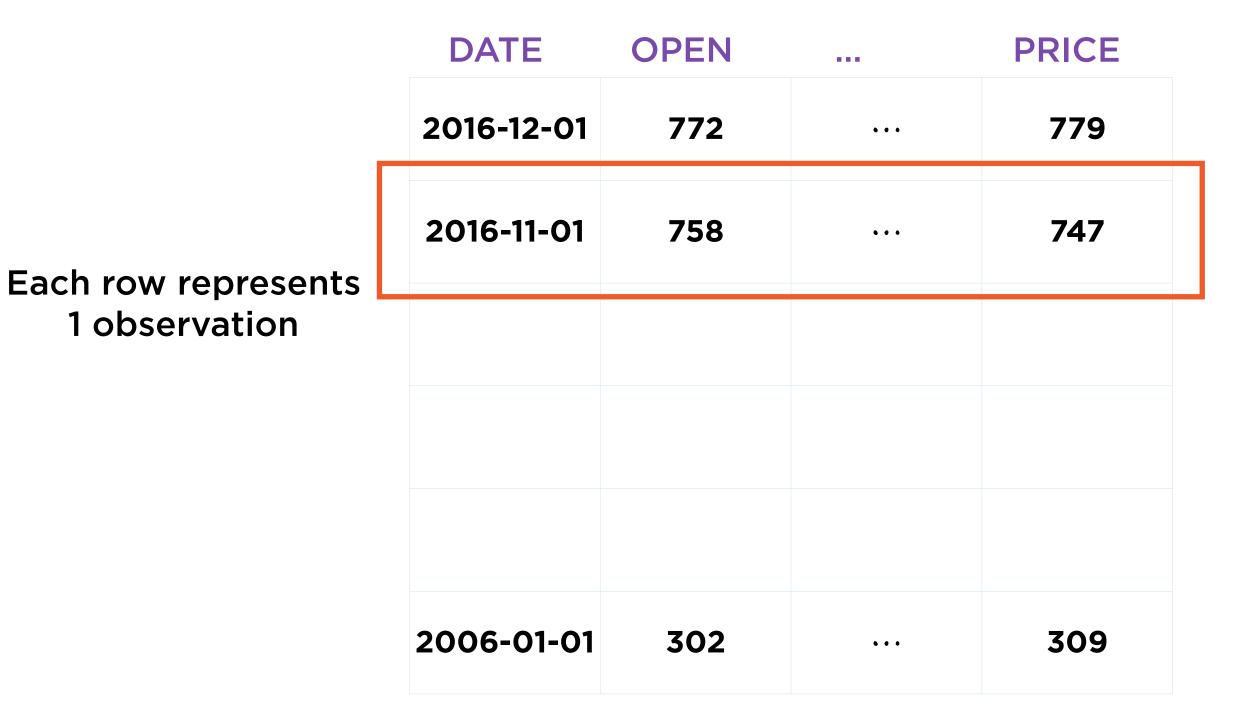


The record of transformations is called lineage

Allows RDDs to be reconstructed in case of node crashes

# The basic data structure for records in Spark 2.x is the DataFrame

#### DataFrame: Data in Rows and Columns

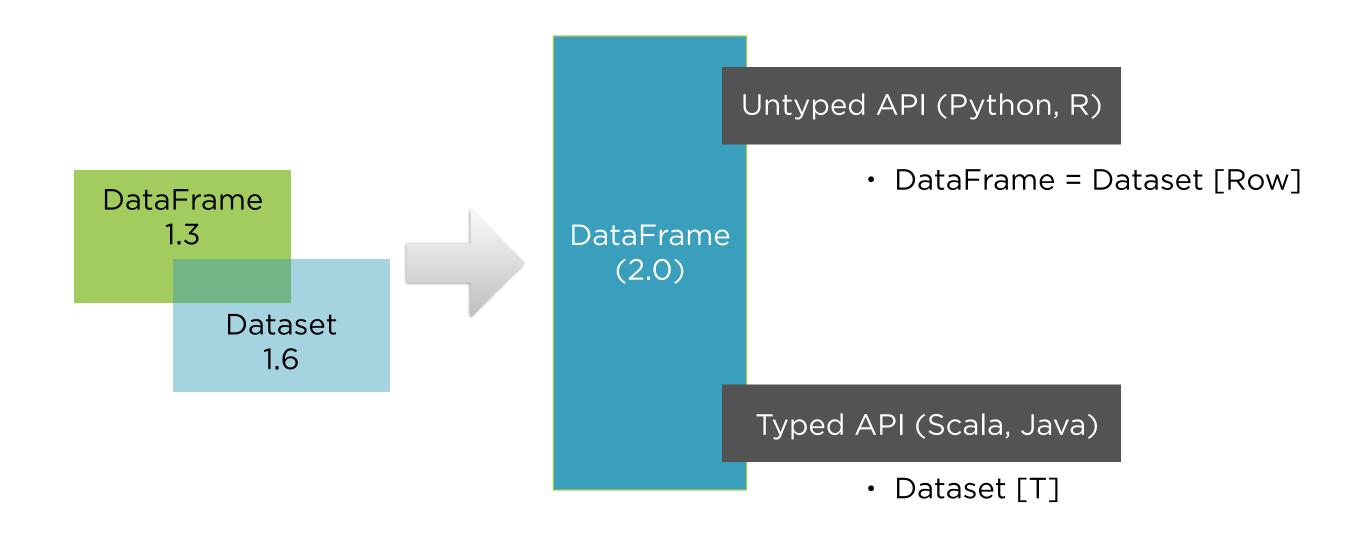


#### DataFrame: Data in Rows and Columns

DATE	OPEN		PRICE
2016-12-01	772	•••	779
2016-11-01	758	•••	747
2006-01-01	302	•••	309

Each column represents 1 variable (a list or vector)

#### Unified API for DataFrames



#### DataFrames Built on Top of RDDs

#### **Partitioned**

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## Streaming in Spark 1.x and Spark 2.x

#### Streaming Data Spark 1.x

```
2016-12-30 09:09:57,862 INFO
org.apache.hadoop.http.HttpServer2: Jetty bound to port
56745
2016-12-30 09:09:57,862 INFO org.mortbay.log: jetty-6.1.26
2016-12-30 09:09:58,037 INFO org.mortbay.log: Started
HttpServer2$SelectChannelConnectorWithSafeStartup@localhost:
56745
2016-12-30 09:09:58,124 INFO
org.apache.hadoop.hdfs.server.datanode.web.DatanodeHttpServe
r: Listening HTTP traffic on /0.0.0.0:50075
2016-12-30 09:09:58,239 INFO
```

2016 -12-30 09:0 9:58 ,239 INFO org.apache.hado op.hdfs.server. datanode.web.Da tanodeHttpServe r: Listening HTTP traffic on /0.0.0.0:50075

HttpServer 2\$SelectCh annelConne ctorWithSa feStartup@ localhost: 56745

2016-12-3 0 09:09:58, 037 INFO org.mortb ay.log: Started 2016-12-30 09:09:57,8 62 INFO org.mortba y.log: jetty-6.1. 26

2016-12-30 09:09:57,862 INFO org.apache.hadoo p.http.HttpServe r2: Jetty bound to port 56745

Each log message is one entity in this stream

2016 -12-30 09:0 9:58 ,239 INFO

org.apache.hado
op.hdfs.server.
datanode.web.Da
tanodeHttpServe
r: Listening
HTTP traffic on
/0.0.0.0:50075

HttpServer 2\$SelectCh annelConne ctorWithSa feStartup@ localhost: 56745

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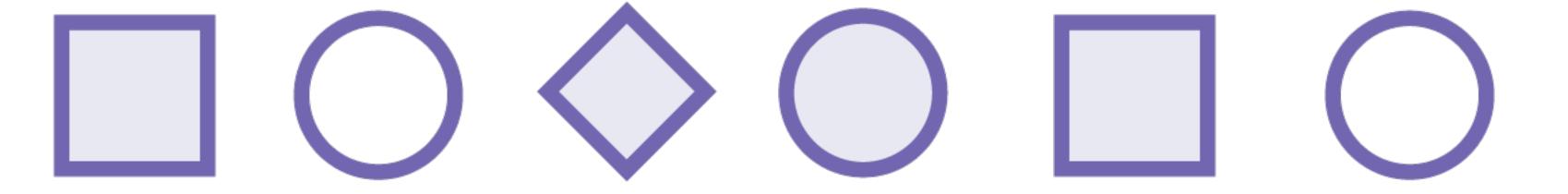
# Spark works with stream data using the same batch RDD abstraction

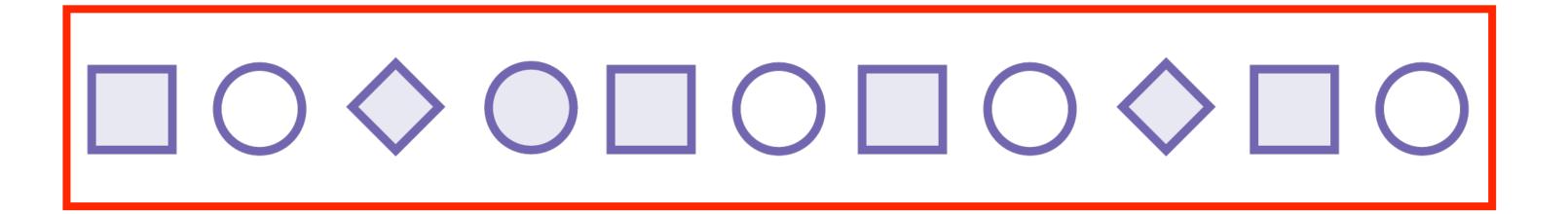
2016 -12-30 09:0 9:58 ,239 INFO

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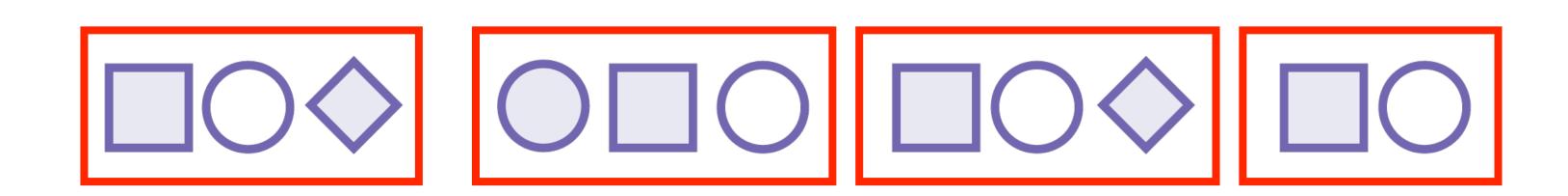
This stream of entities is represented as a discretized stream or DStream

# DStream = Sequence of RDDs



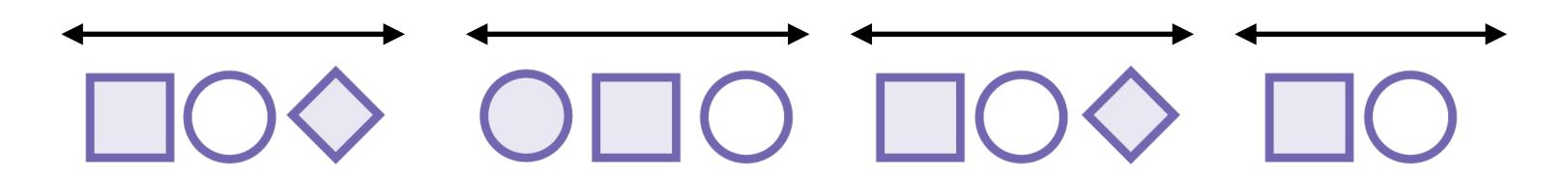
Entities => RDDs => DStream



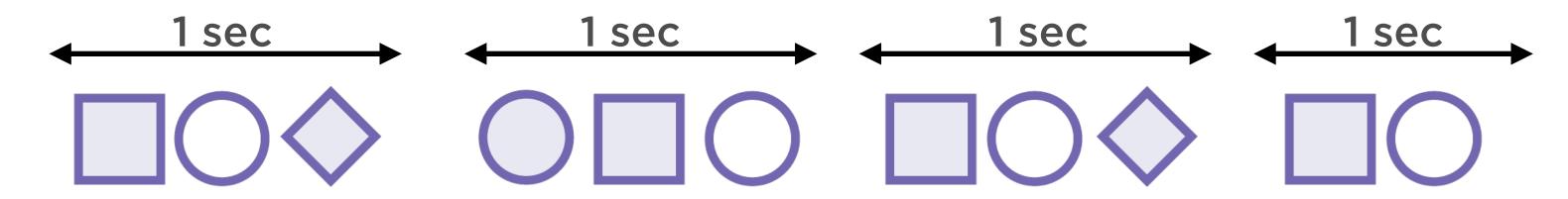


Entities in a stream are grouped into batches

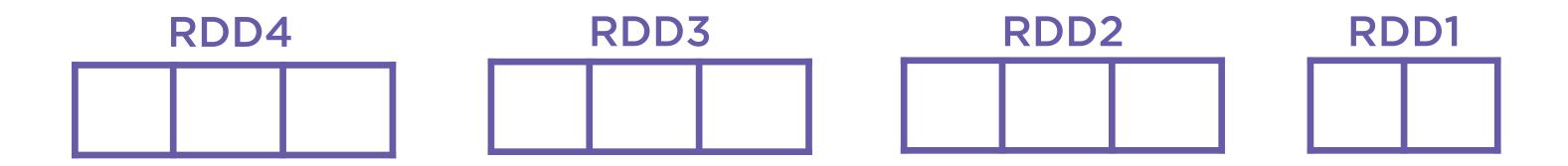
Each batch = 1 RDD



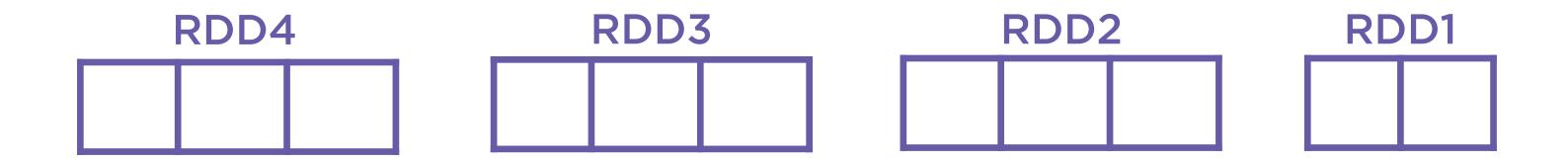
Batches are formed based on a batch interval



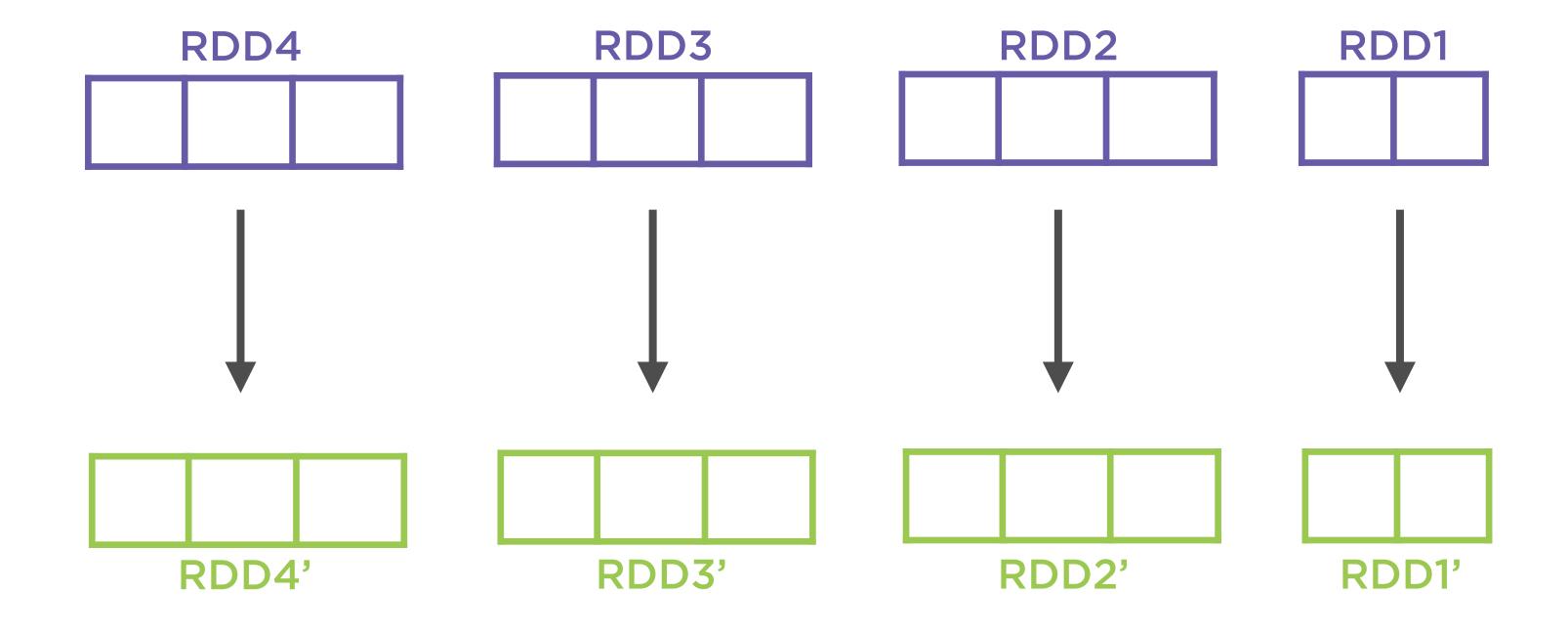
All entities received within the batch interval make one RDD

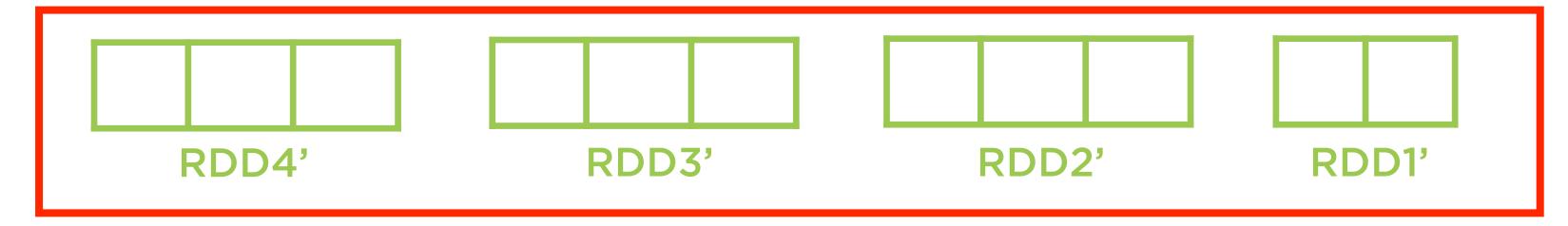


All entities received within the batch interval make one RDD



Within a DStream, Spark still performs operations on individual RDDs





**Transformed DStream** 

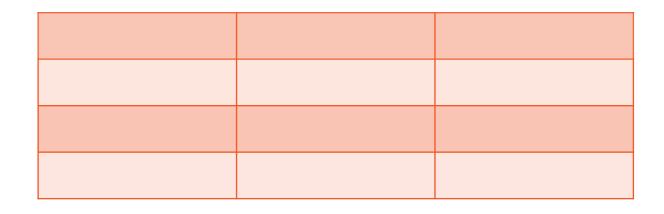
# Within DStreams, Spark 1.x does batch processing on individual RDDs

# The basic data structure for records in Spark 2.x is the DataFrame

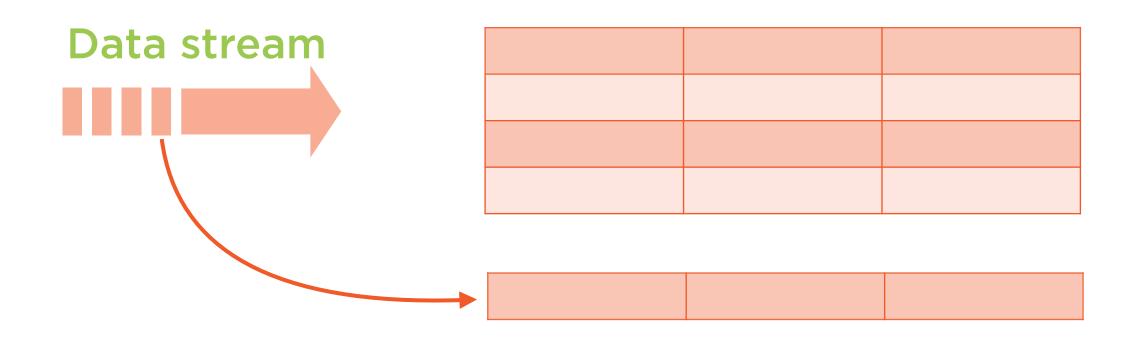
#### DataFrame: Data in Rows and Columns

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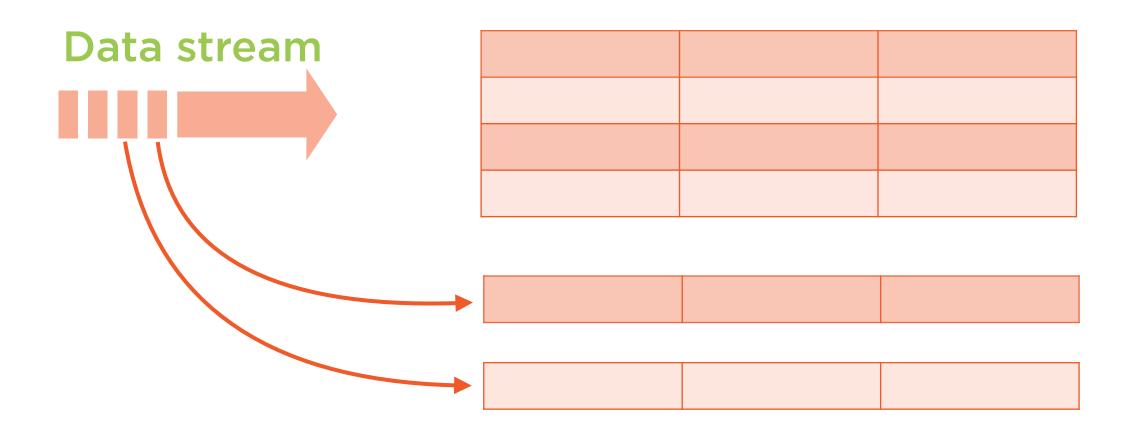




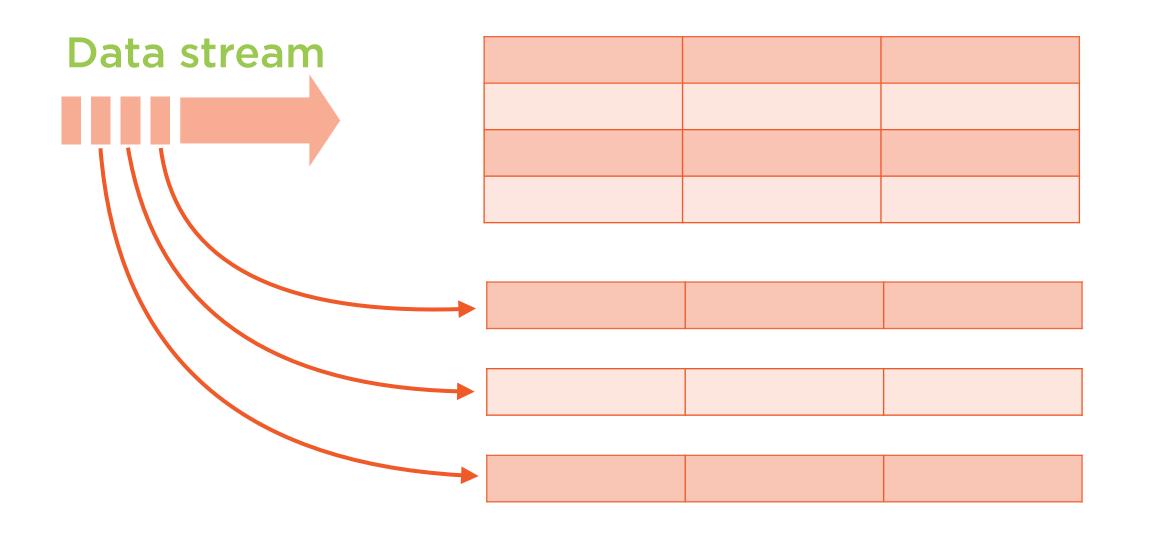
Every data item that is arriving on the stream is like a new row being appended to the input table



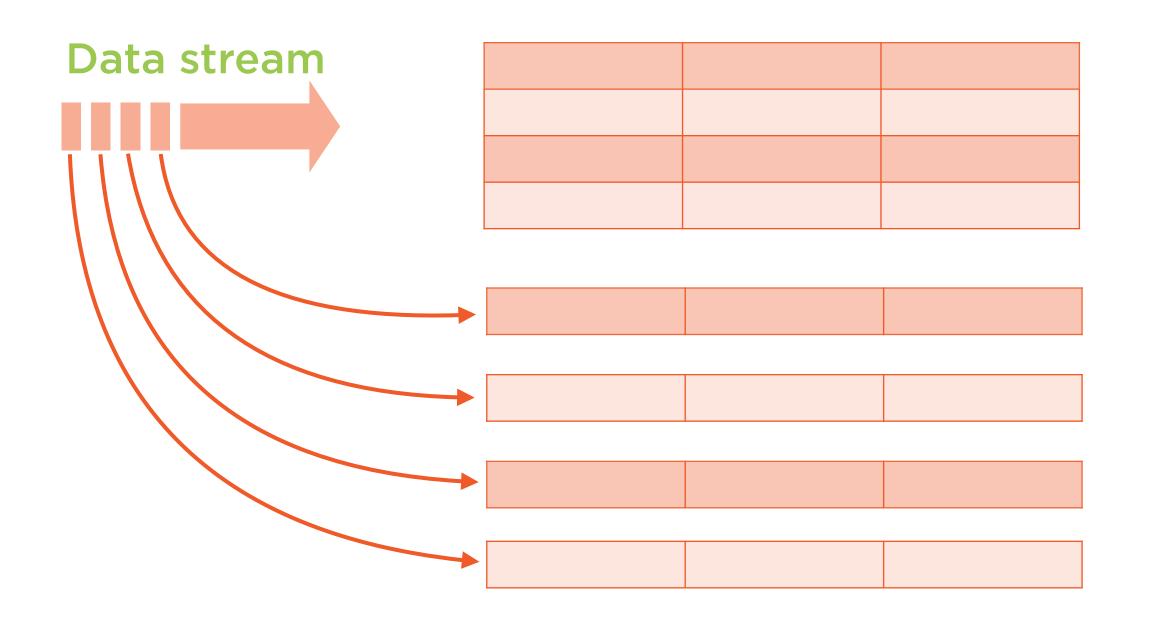
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Every data item that is arriving on the stream is like a new row being appended to the input table

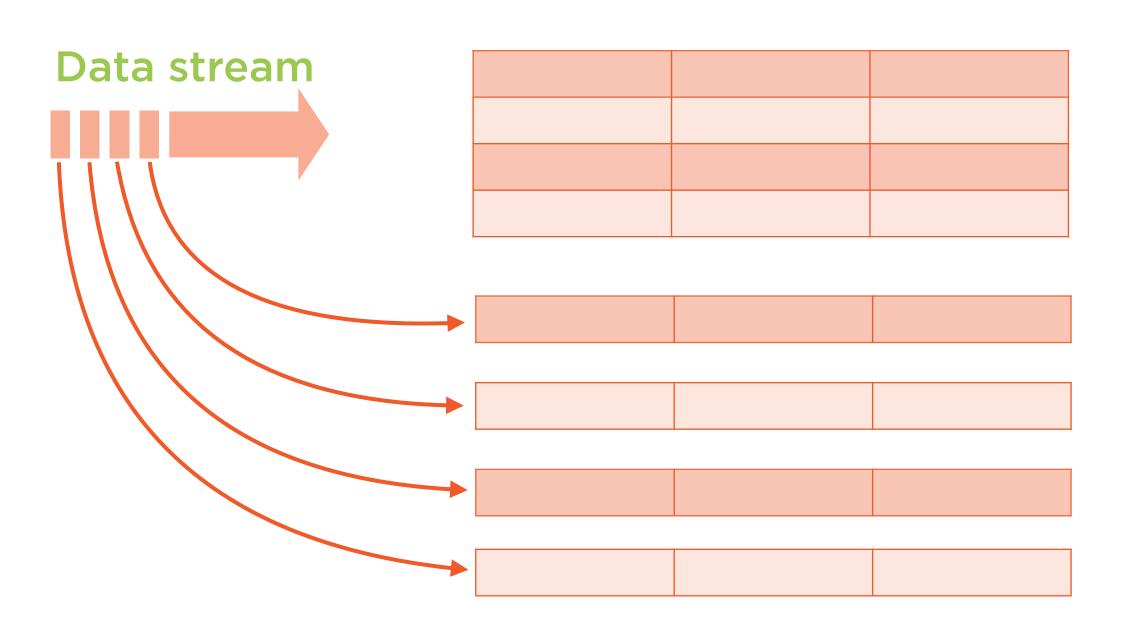


Every data item that is arriving on the stream is like a new row being appended to the input table



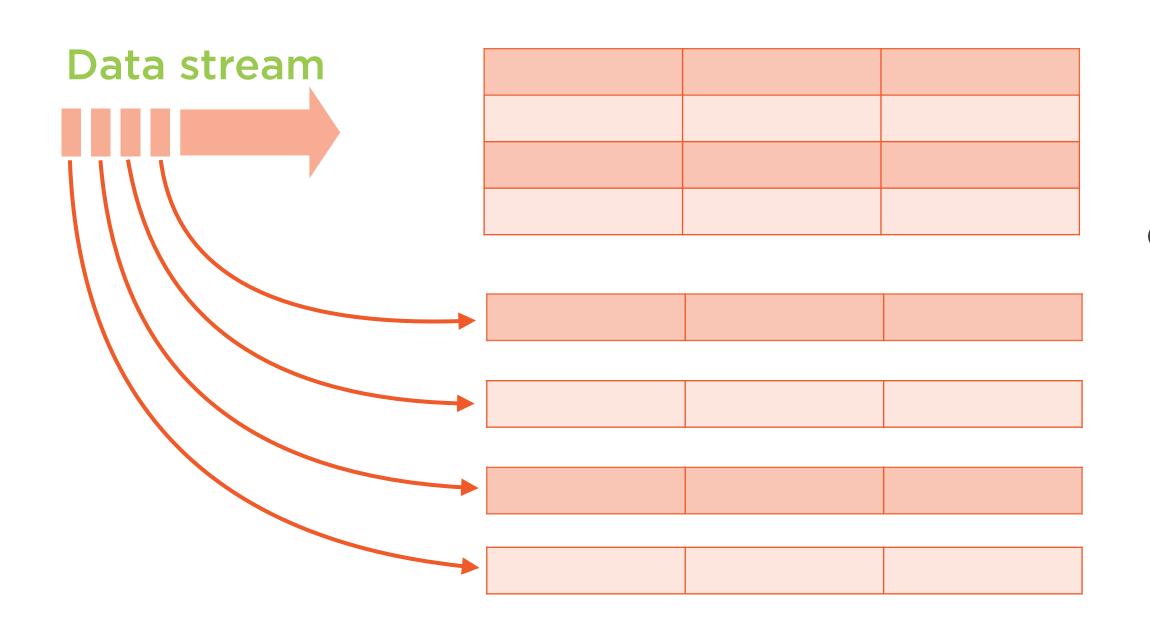
Every data item that is arriving on the stream is like a new row being appended to the input table

#### Batch is Simply A Prefix of Stream



In other words, the input table (batch) is simply a prefix of the stream

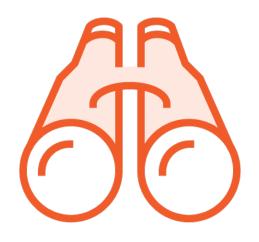
#### Batch is Simply A Prefix of Stream

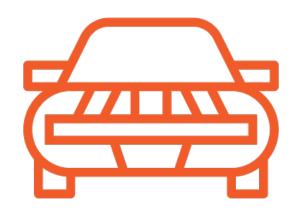


All operations that can be performed on data frames can be performed on the stream

# Structured Streaming treats a live data stream as a table that is being continuously appended

#### Structured Streaming







#### What

A high-level API that takes burden off user

#### How

Micro-batch processing with exactly-once fault-tolerance

#### Why

Code virtually identical for batch and streaming

## Structured Streaming

High-level API in Apache Spark 2.0 that supports continuous applications and replaces Spark Streaming

https://databricks.com/blog/2016/07/28/continuous-applications-evolving-streaming-in-apache-spark-2-0.html

#### Streaming and Structured Streaming

Streaming

**Structured Streaming** 

Older

Newer

**RDDs** 

**DataFrames** 

No optimizations

**Optimizations on DataFrames** 

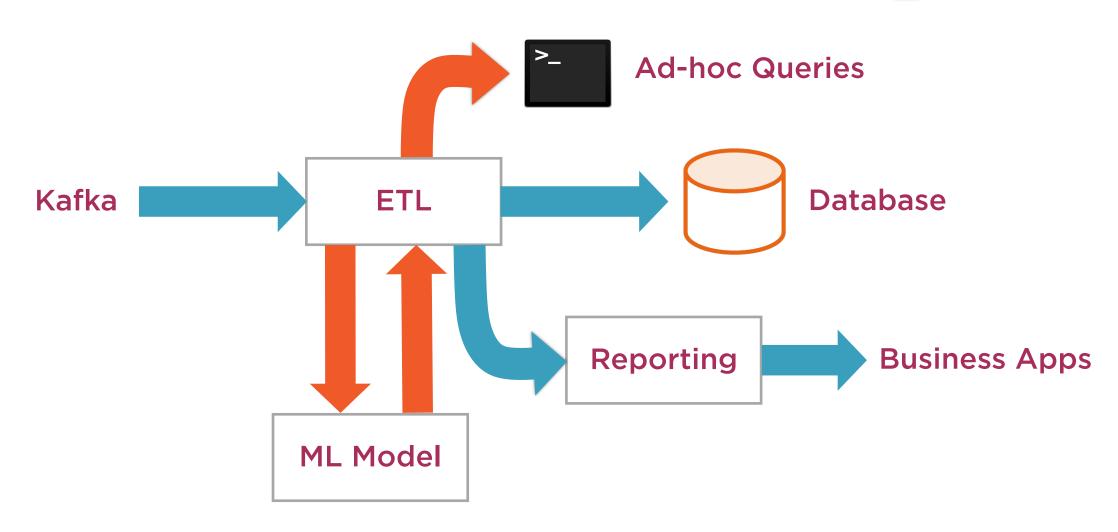
Batch and streaming support not unified

Unified support for batch and streaming

## Structured Streaming

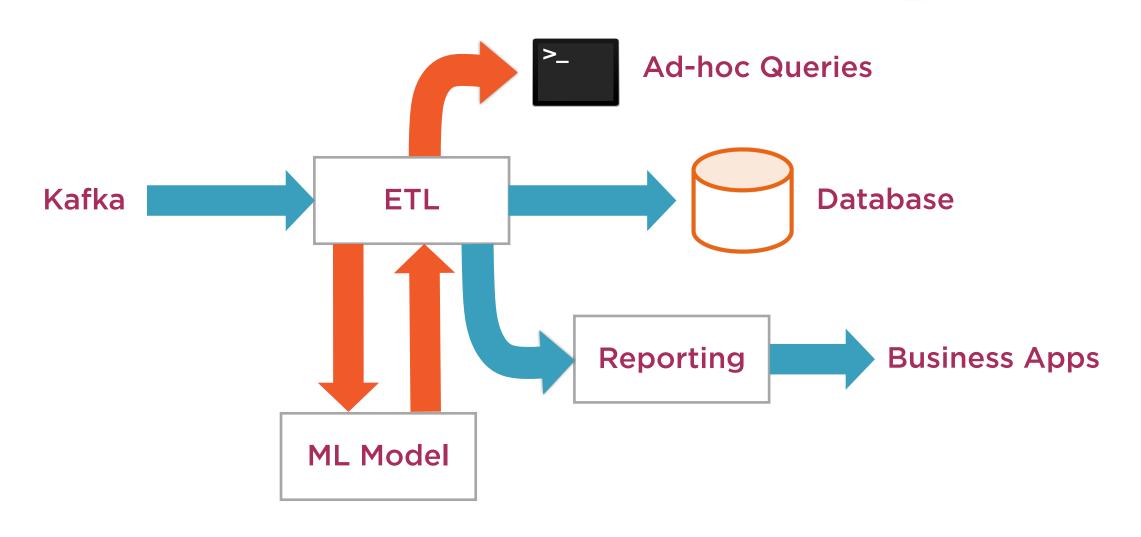
#### Spark 2.x: Continuous Applications

Traditional streamingDataset (bounded or unbounded)

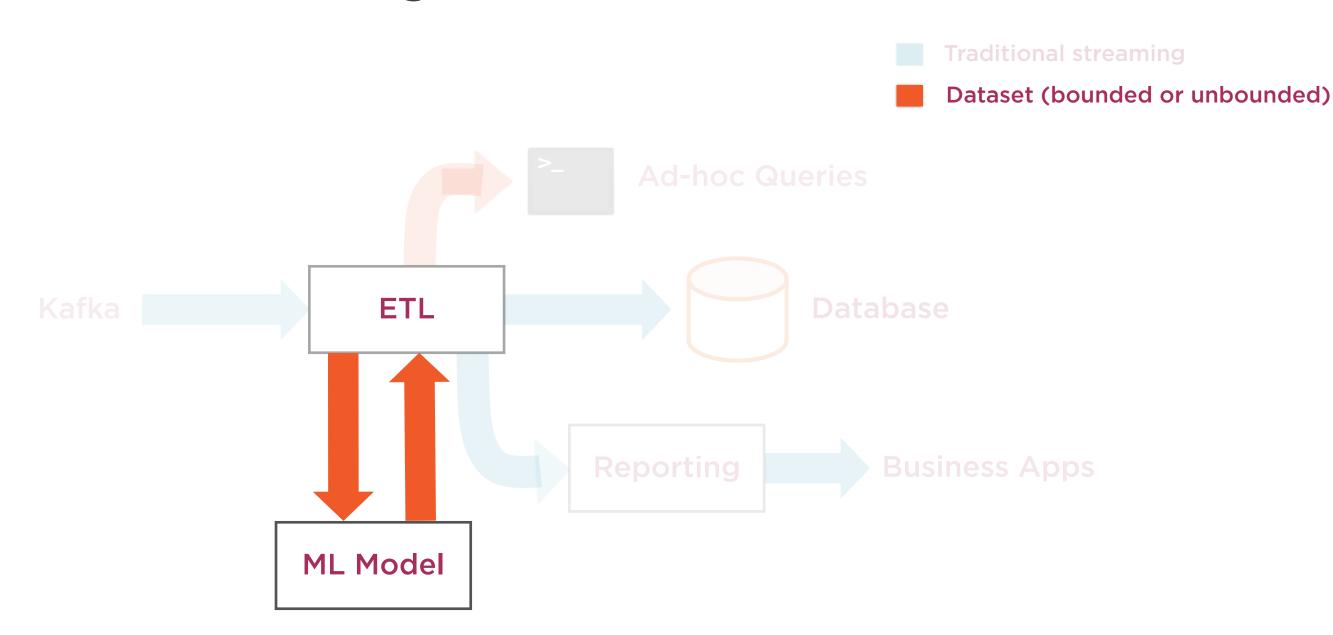


### Structured Streaming

Traditional streamingDataset (bounded or unbounded)

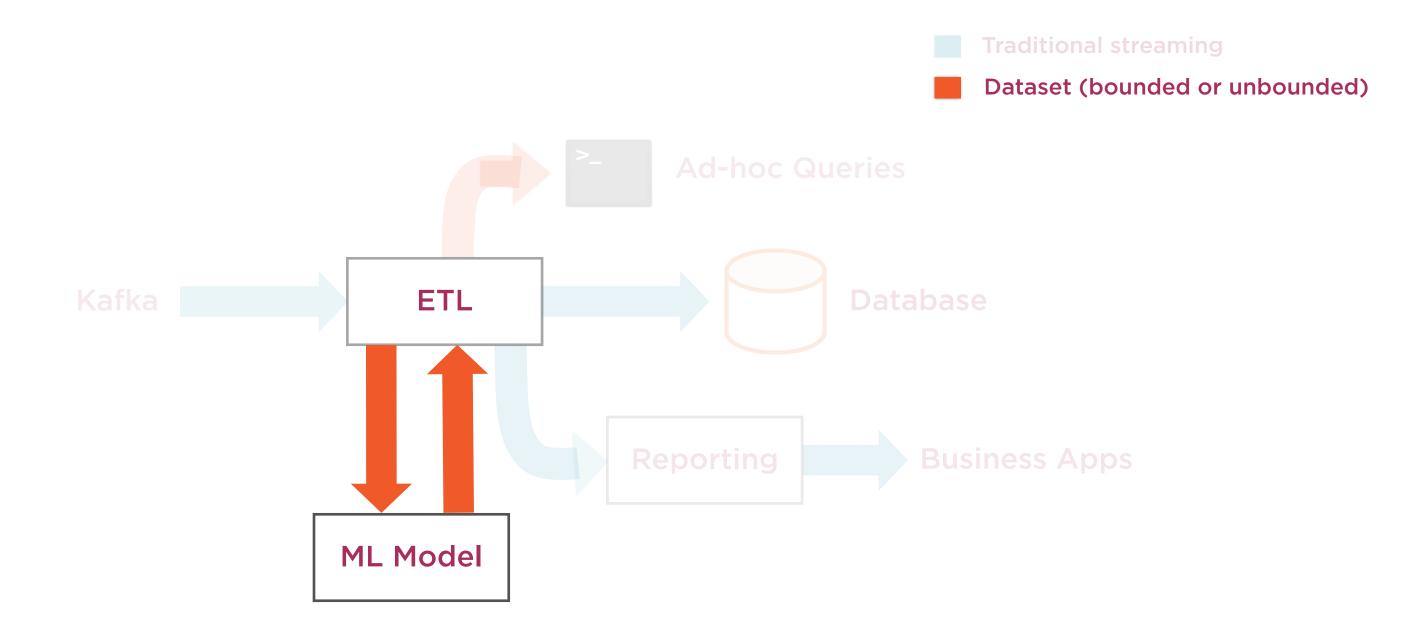


#### High-level User API



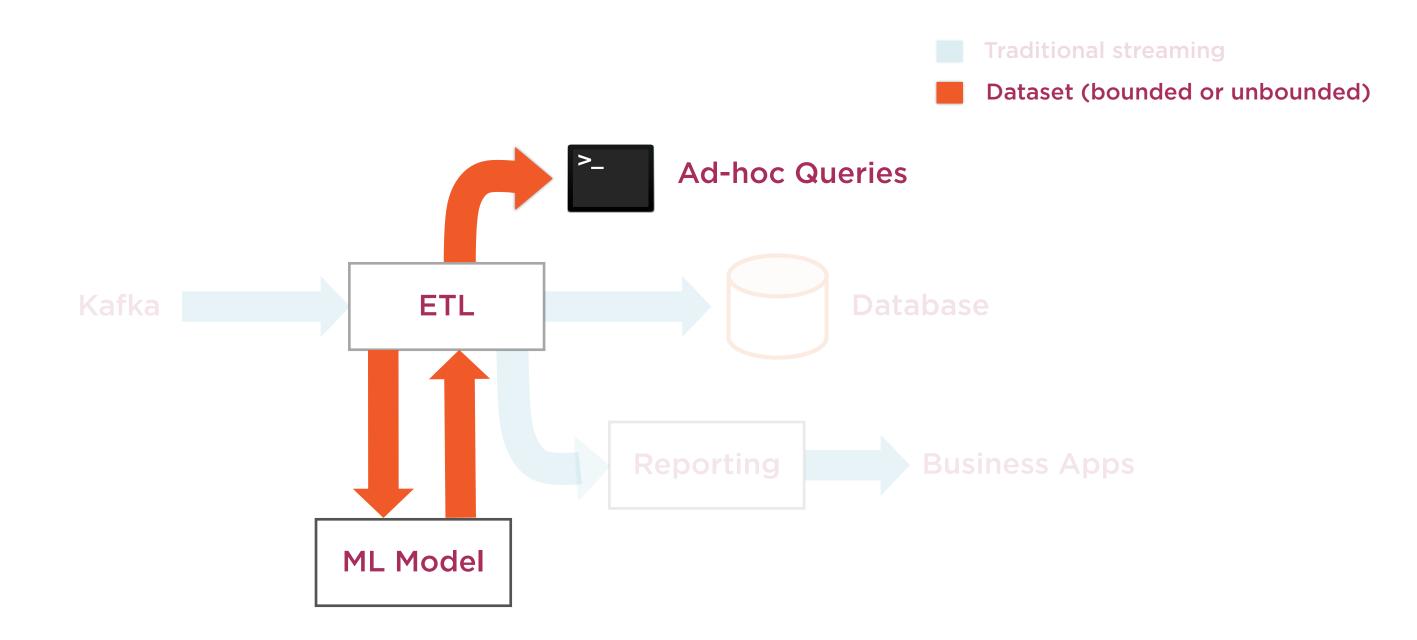
User implements batch computation using DataFrame/Dataset API

### Automatic Support for Continuous Apps



Spark automatically incrementalizes the batch computation

#### Automatic Support for Continuous Apps



i.e. Spark automatically converts the job from batch to streaming

#### Micro-batch Processing



Default stream processing mode in Spark

Data streams processed as a series of batch jobs

End-to-end latencies as low as 100ms

Exactly-once fault-tolerance guarantees

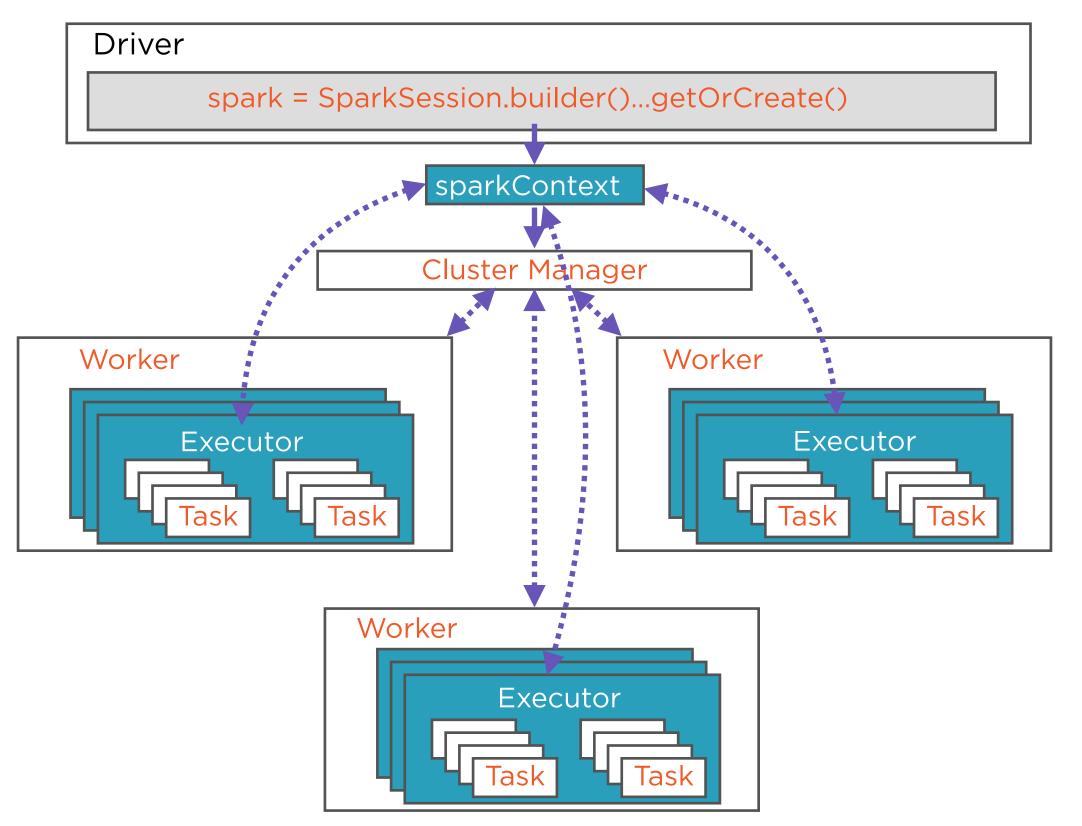
#### Micro-batch Processing

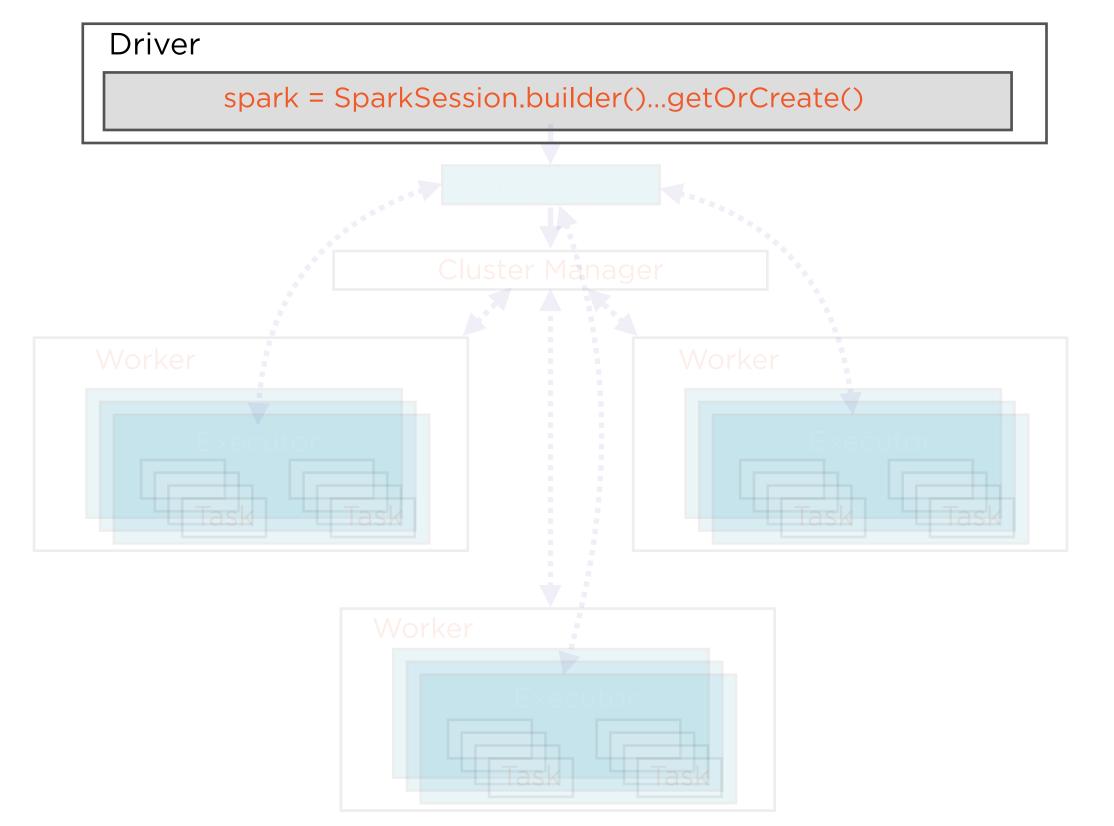


Since Spark 2.3 there exists a new continuous processing mode

Currently in the experimental stage

## Spark Architecture





#### Driver



Separate process (JVM)

The master node in a Spark application

Launches tasks

**Hosts SparkContext** 

#### Driver



# Several groups of services run inside the driver

- SparkEnv
- DAGScheduler
- Task Scheduler
- SparkUI
- -

# Spark Application



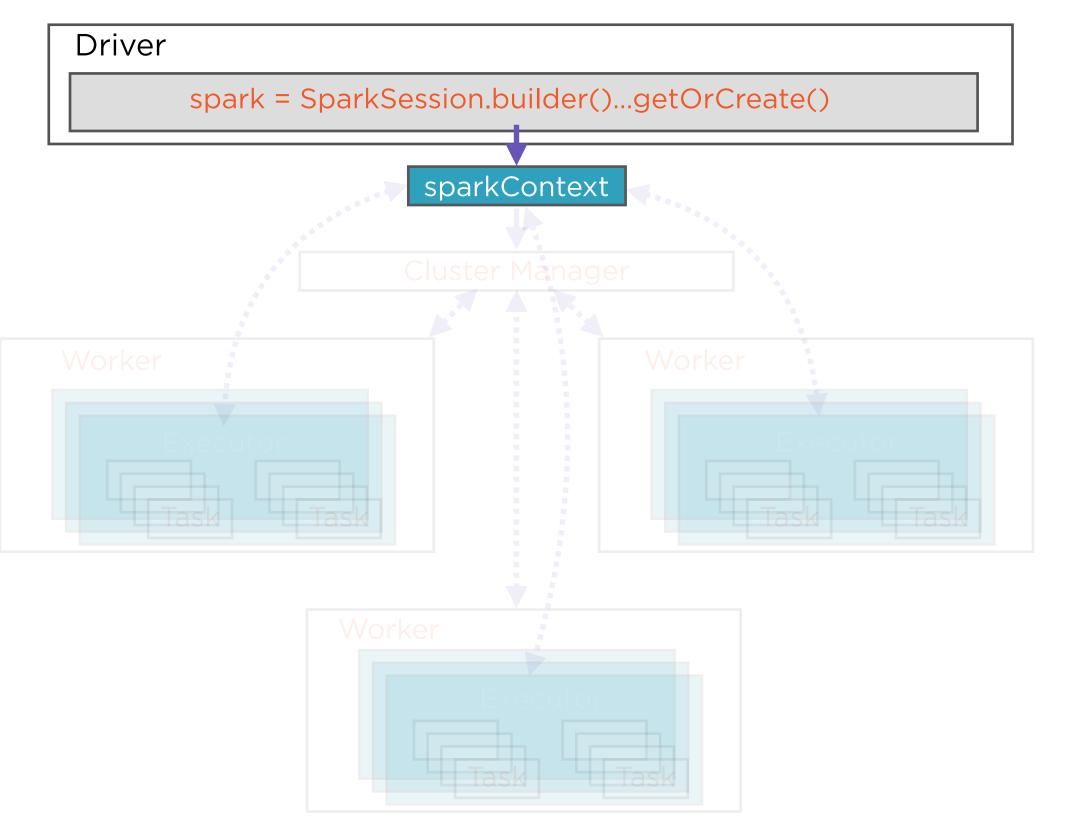
Uses SparkContext as entry point

Creates RDD Directed Acyclic Graph

DataFrames run on top of RDDs

Internally, Spark creates *Stages* (physical execution plan)

Each stage is split into operations on RDD partitions called *Tasks* 



### SparkContext (Spark 1.x)



Familiar code entry point to Spark

sc = new sparkContext(...)

Create RDDs, accumulators...

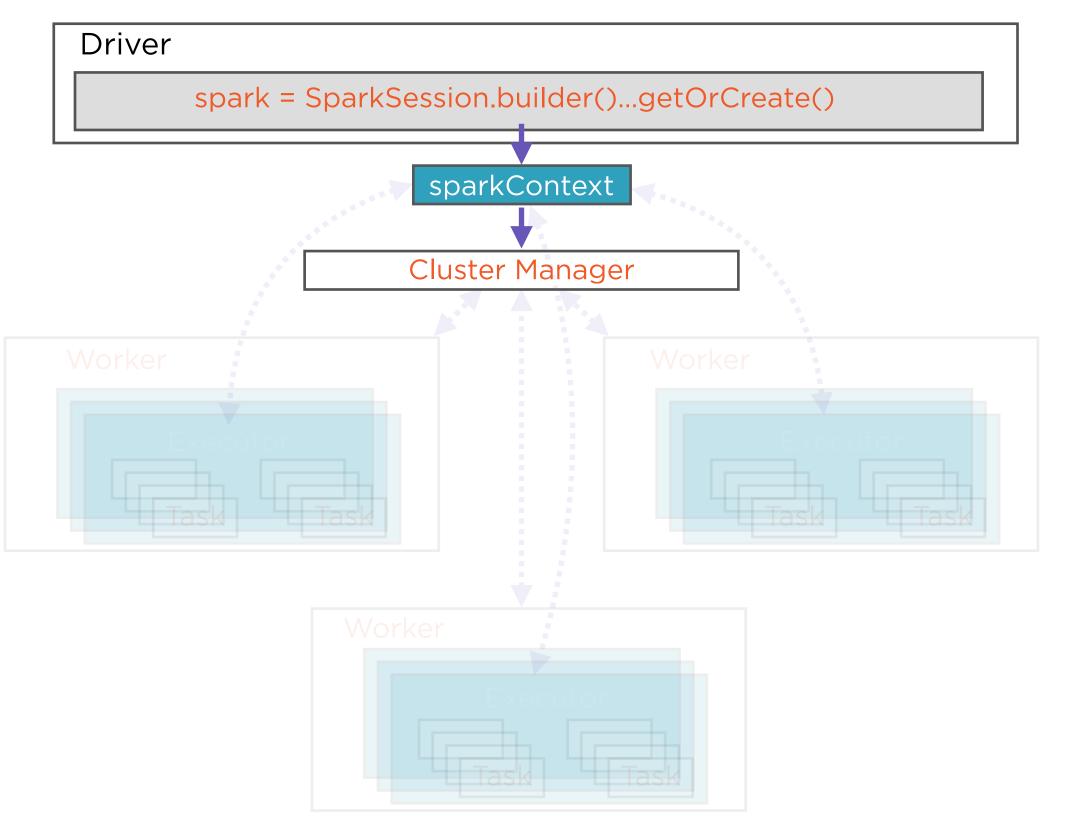
Run jobs

# SparkSession (Spark 2.x)

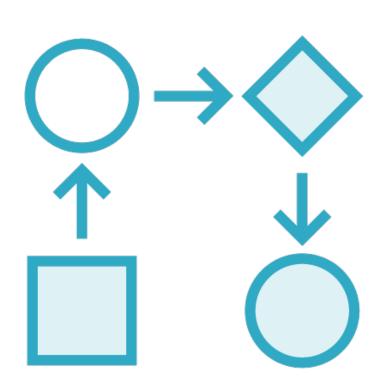


In Spark 2.x, SparkContext is wrapped in SparkSession

**Encapsulates SparkContext, SQLContext, HiveContext...** 



## Cluster Manager



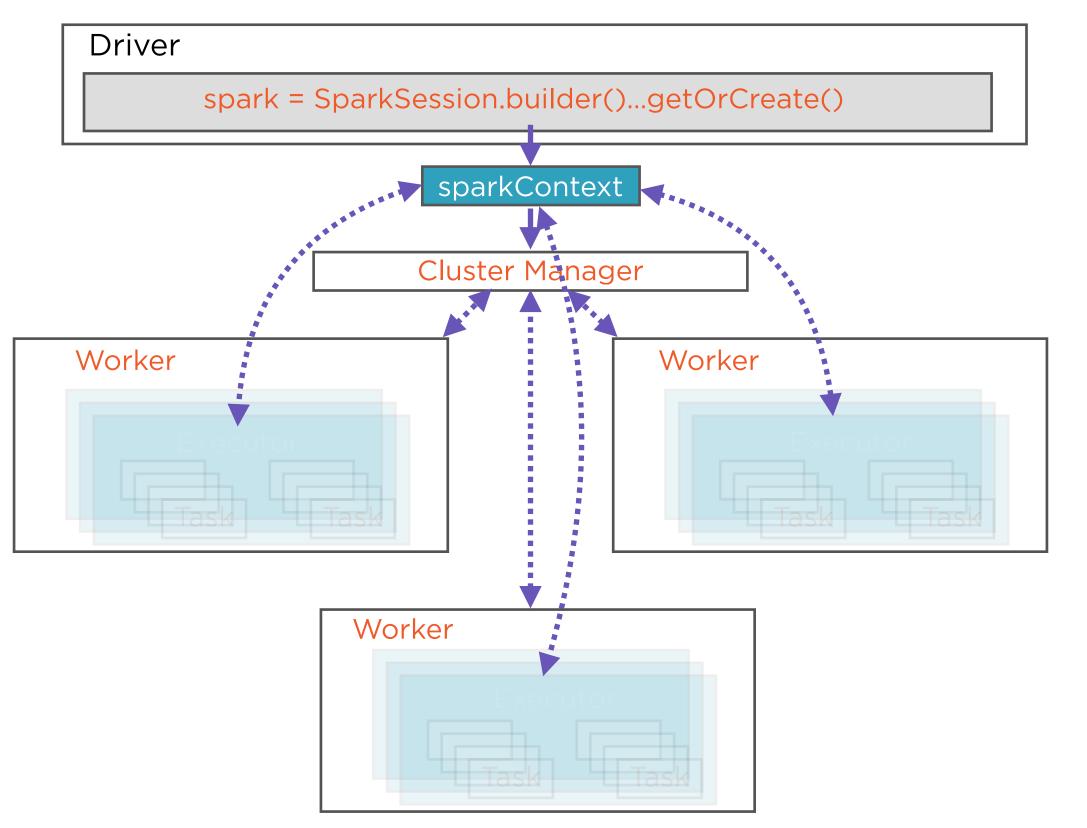
Hadoop's YARN

**Apache Mesos** 

Kubernetes

**Spark Standalone** 

**Orchestrates execution** 



#### Worker

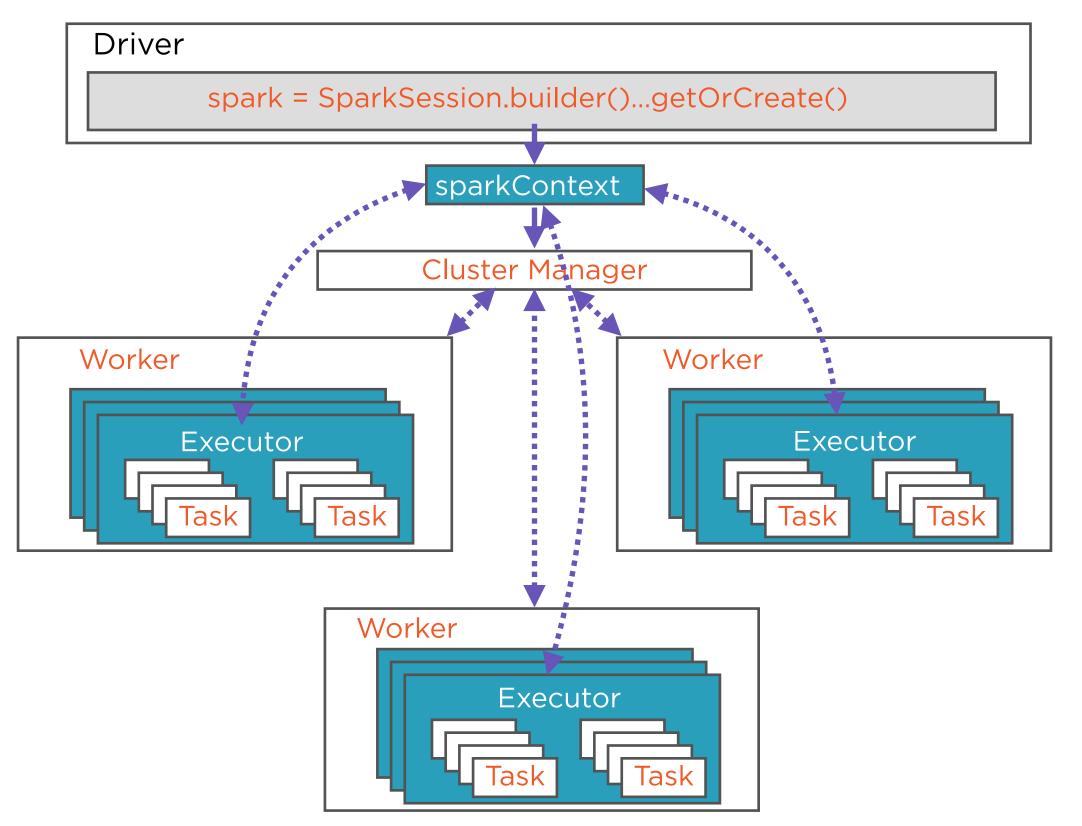


Compute nodes in cluster

Runs the Spark application code

When SparkContext created...

...Each worker starts executors



#### Executors



Distributed agents that execute tasks

Tasks are basic units of execution

Tasks belong inside *stages* 

Stages are physical units of execution

Install and set up Apache Spark on MacOS

Install and set up Apache Spark on Windows

Streaming word count with a socket source and a console sink

Aggregations on streaming data with a socket source and console sink

#### Summary

Streaming and its place in the Apache Spark stack

**RDDs and DataFrames** 

Structured Streaming in Spark 2.x

Batch as a prefix of stream

ClusterManager, SparkSession, and Executors

#### Up Next:

Executing Streaming Queries