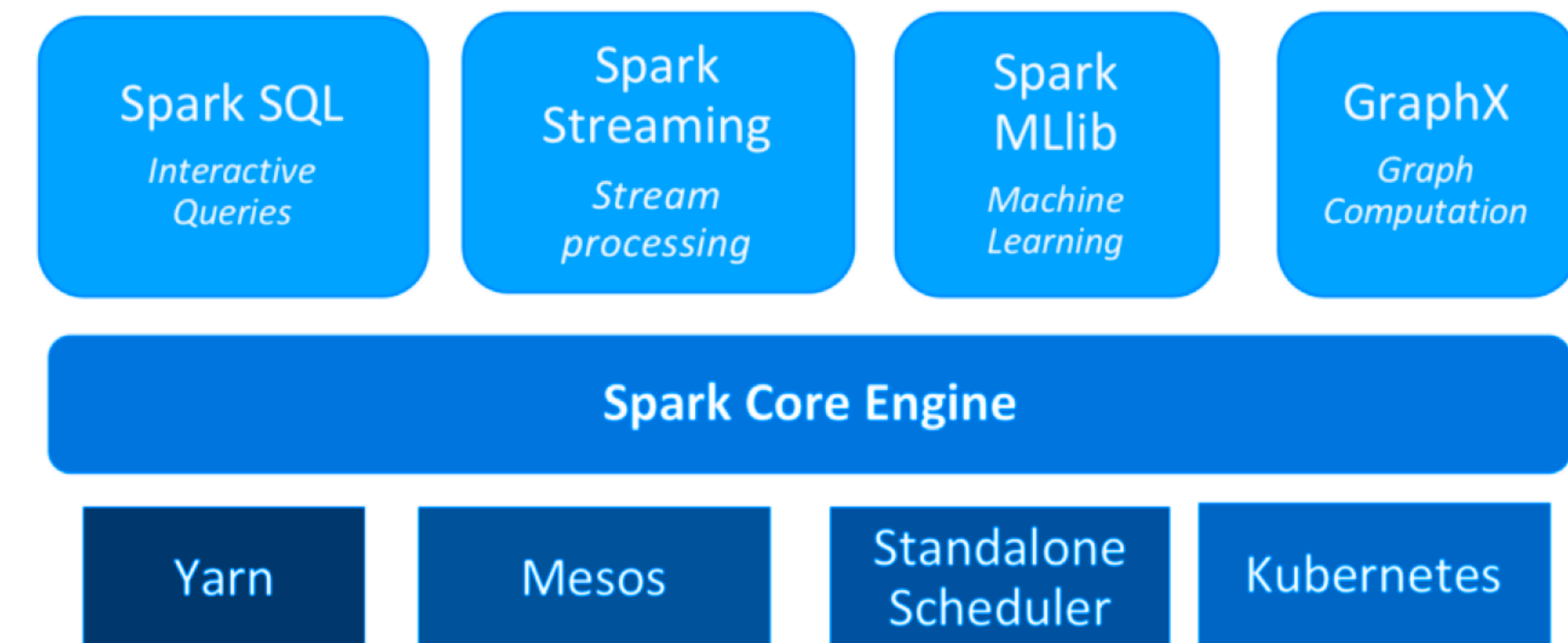


Spark & Scala

Developing highly distributed applications

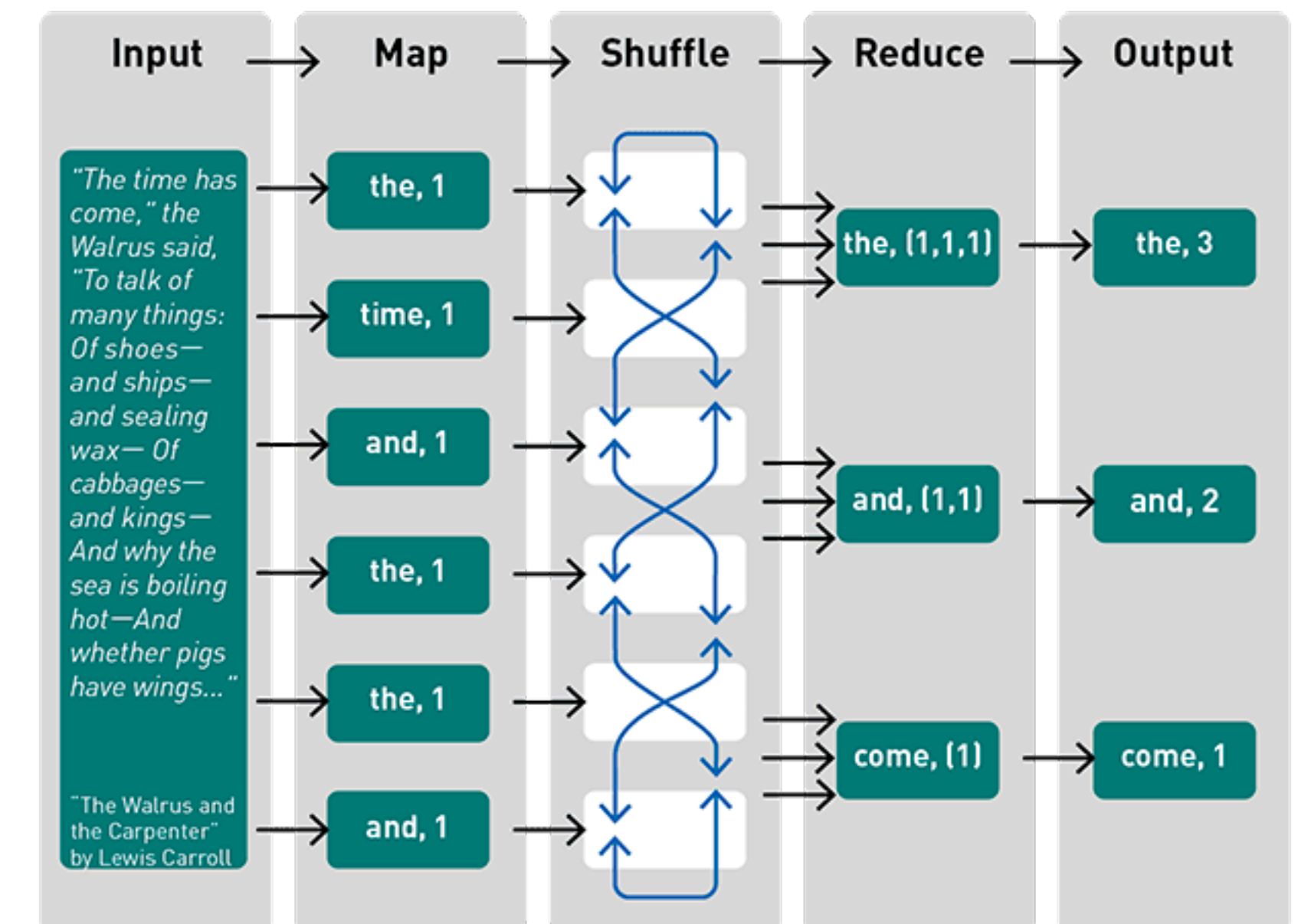
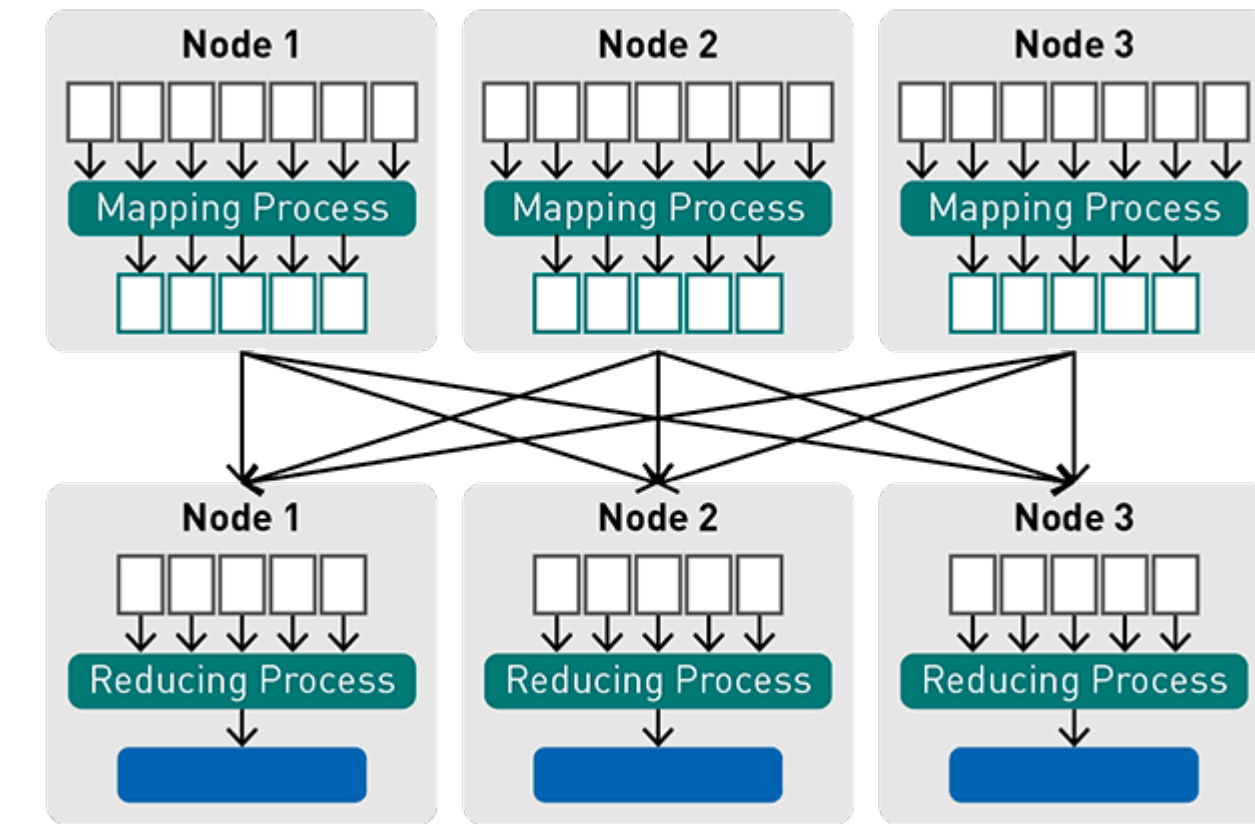
What is Spark?

- unified analytics engine
- large-scale data processing
- fault tolerant
- highly distributed
- 'easy to use'
- runs 'everywhere'



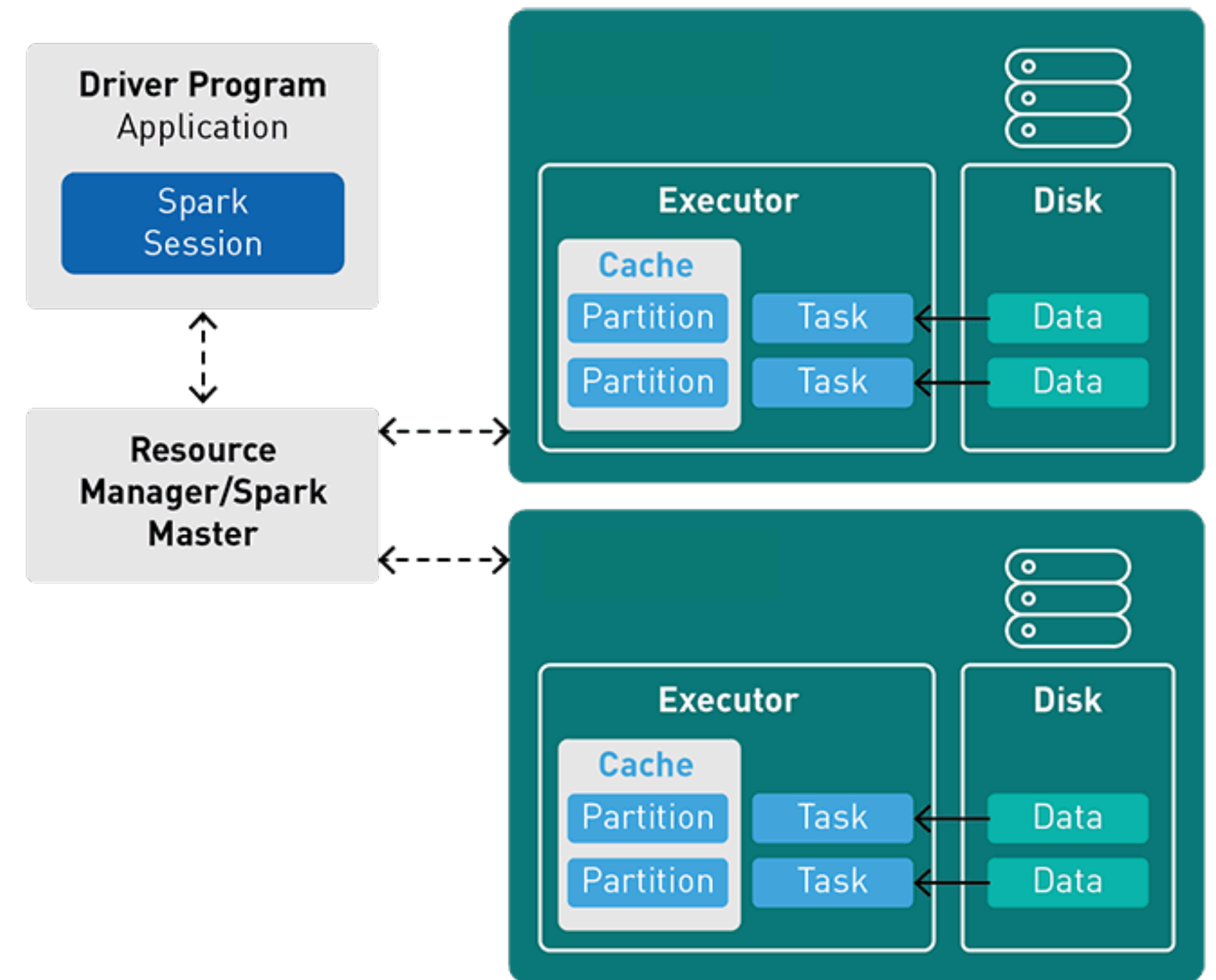
MapReduce

- Chunk data in store
- Distribute compute
 - Map nodes to data chunks
 - Reduce the results
- Fault tolerant



SparkContext

- Independent process coordinated by **SparkSession** in **Driver**
- **Resource Manager** assigns **Tasks** to **Executors** on **Worker** nodes
- Each **Task** is assigned to a **Partition** of data
- A **Task** applies a unit of work to the **Partition** and outputs a new **Partition**
- Results are sent back to the **Driver**



Spark Jobs

- **Job** is an individual action
- **Job** is split in **Stages**, based on reading, caching and joining steps or shuffling events
- **Tasks** are the minimum execution unit

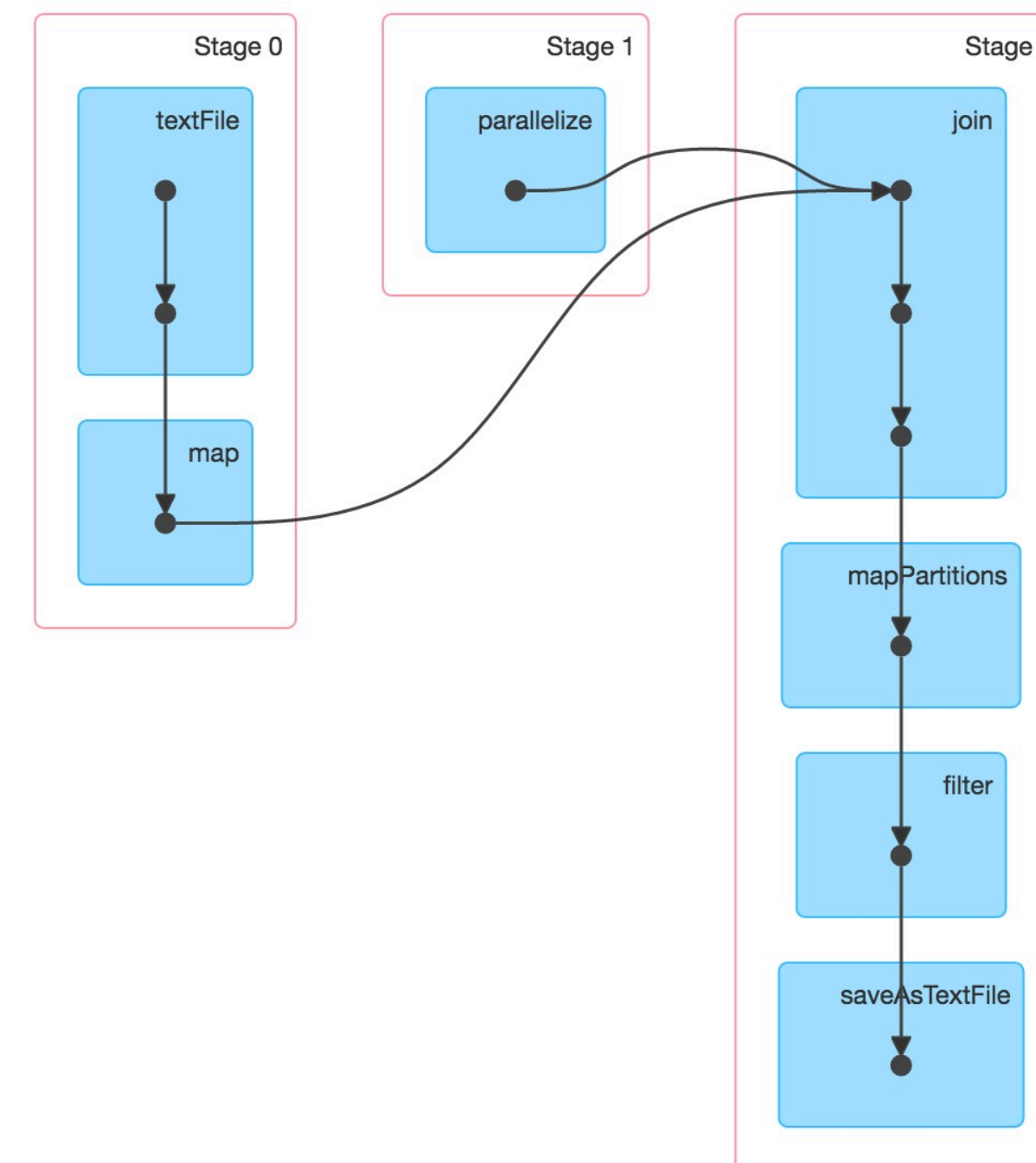
Details for Job 0

Status: SUCCEEDED

Completed Stages: 3

► Event Timeline

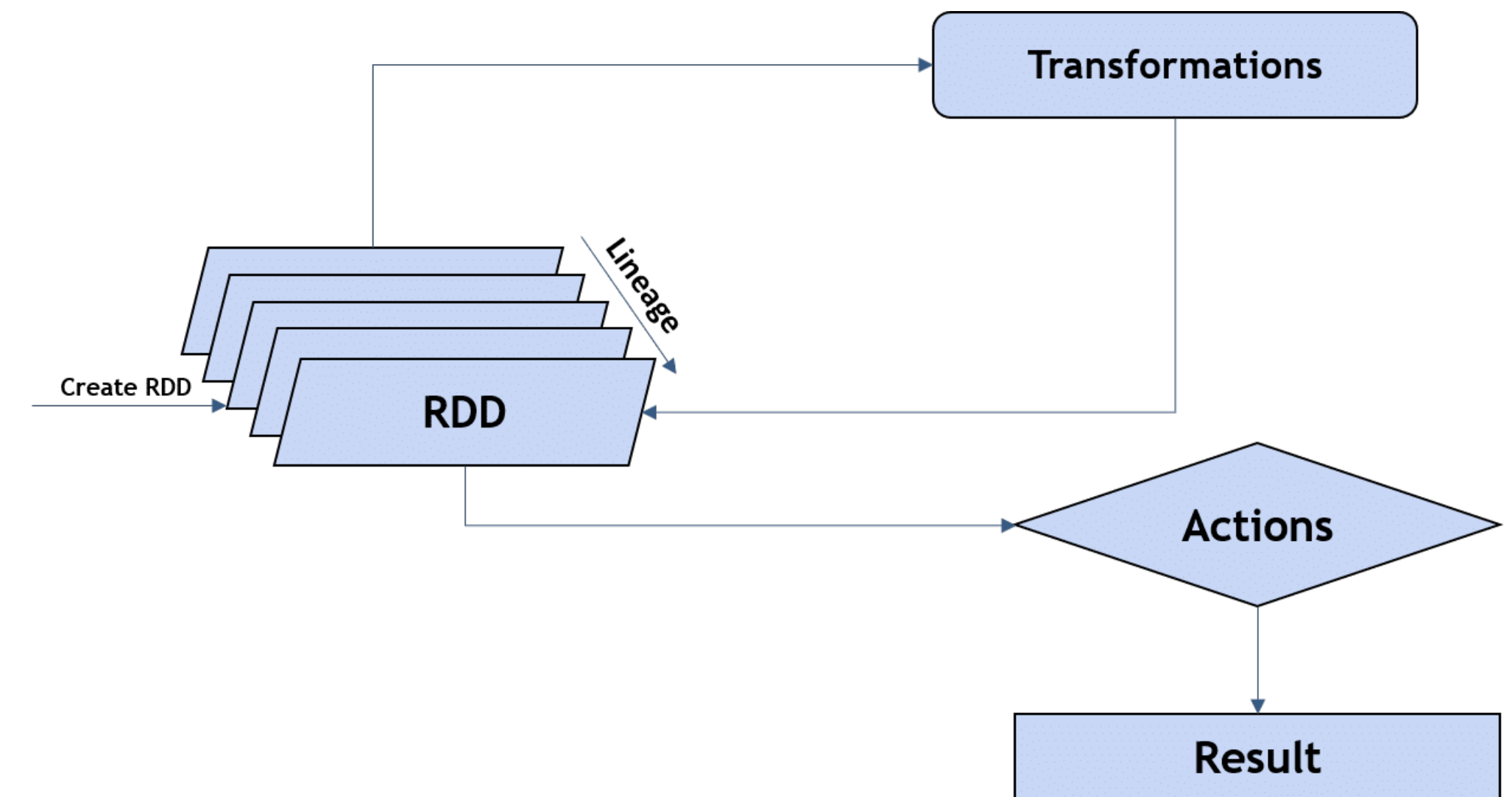
▼ DAG Visualization



demo JobsContext & Spark UI

RDD

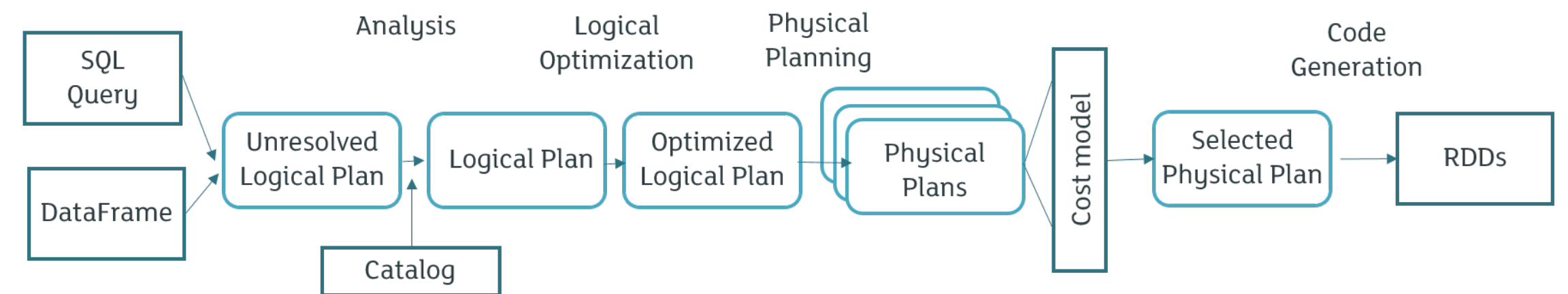
- **R**esilient **D**istributed **D**ataset
- Spark Core
- Huge List split in chunks across machines
- Lazy evaluated
- Composed of
 - Partitions
 - Functions
 - Dependencies other RDD



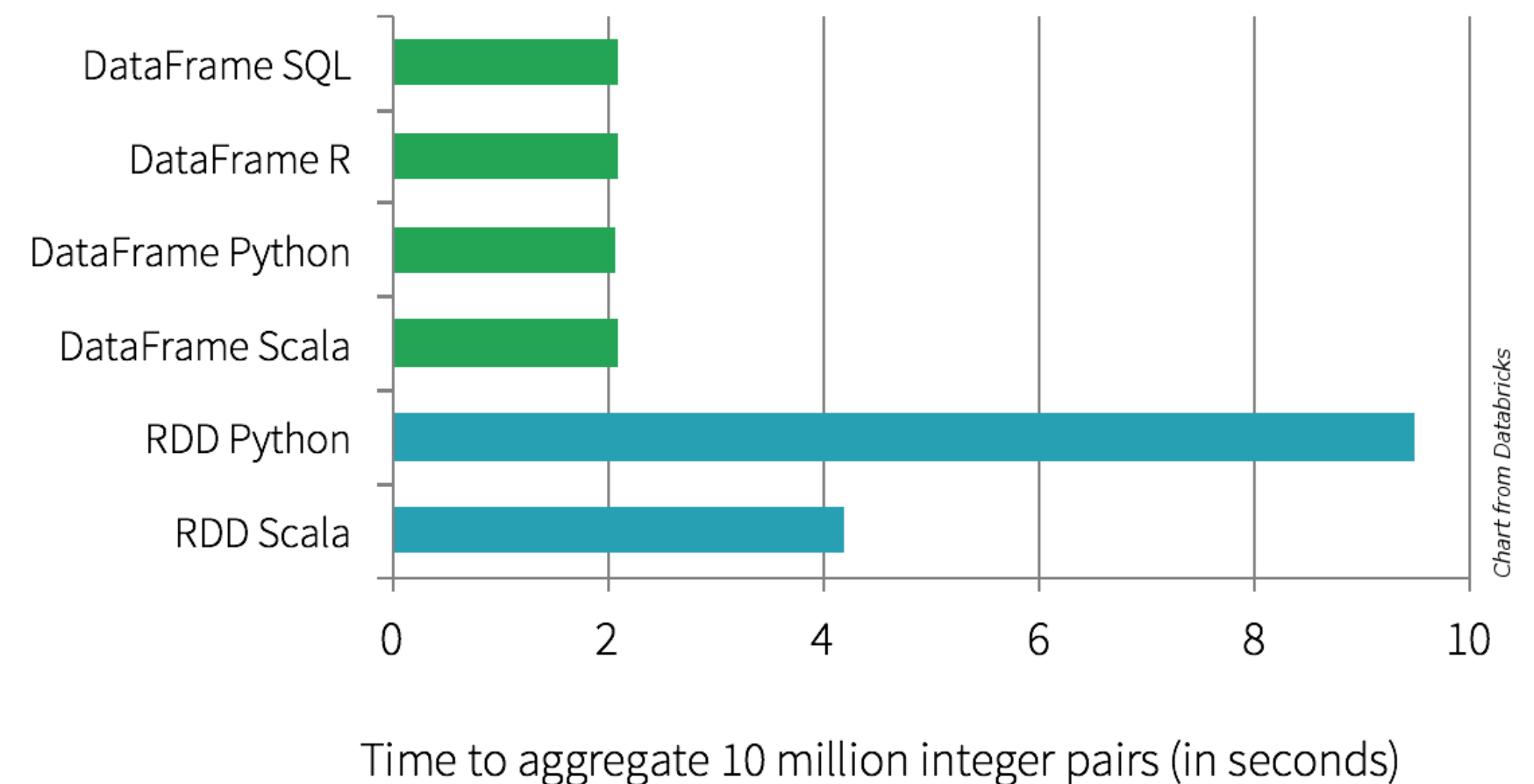
demo RDD & PairRDD

SparkSQL

- Data sources unified
 - Json / Parquet / Avro / CSV / Text / ...
 - Postgres / MongoDB / Kafka / ElasticSearch / ...

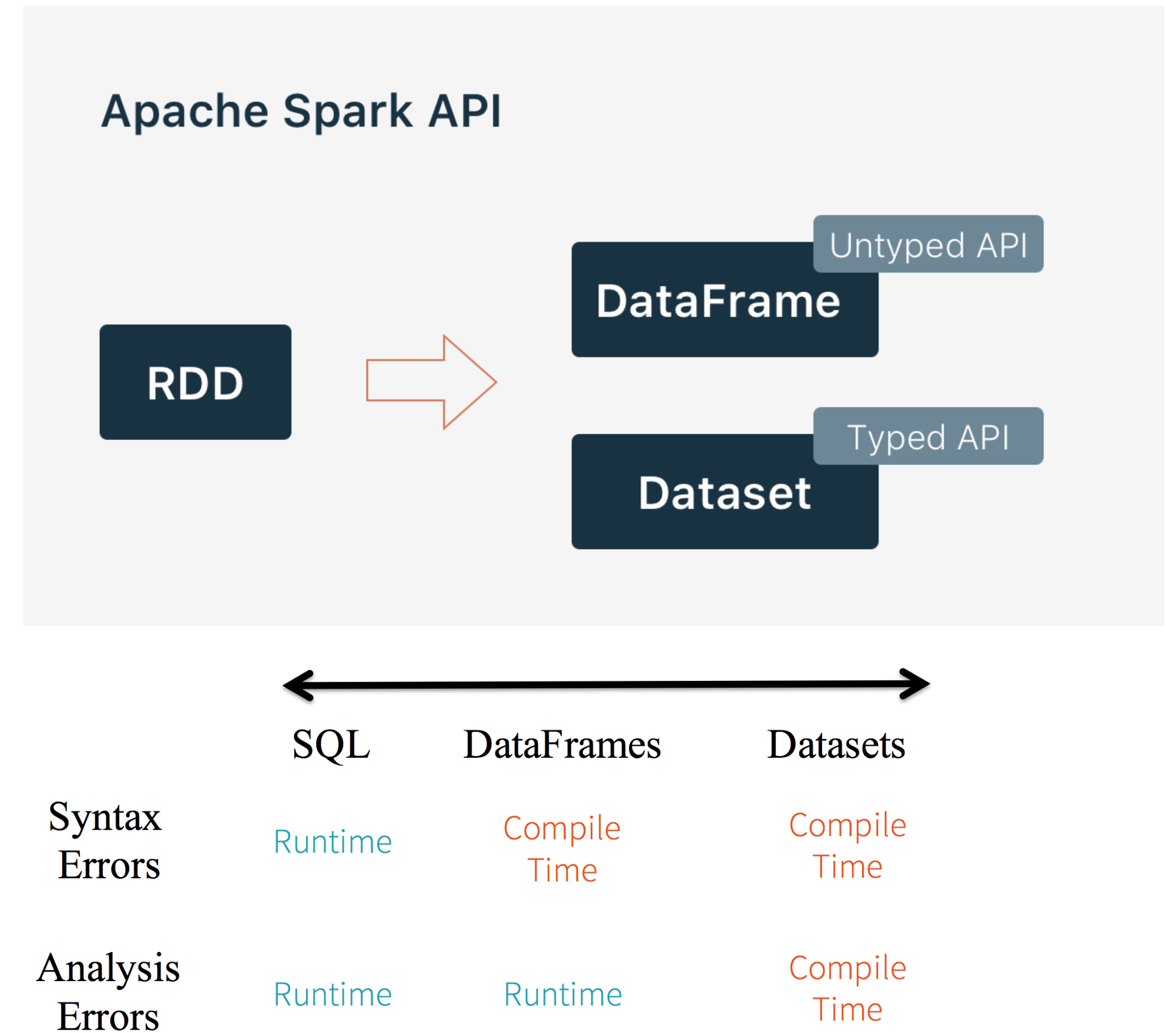


- **DataFrame**
 - DataFrame = RDD[Row]
 - Row holds column & types
 - Good for analysis of messy data
- **Tungsten** Custom Memory Management (off heap binary)
- **Catalyst Engine** (Optimised Plans)



SparkSQL: Dataset

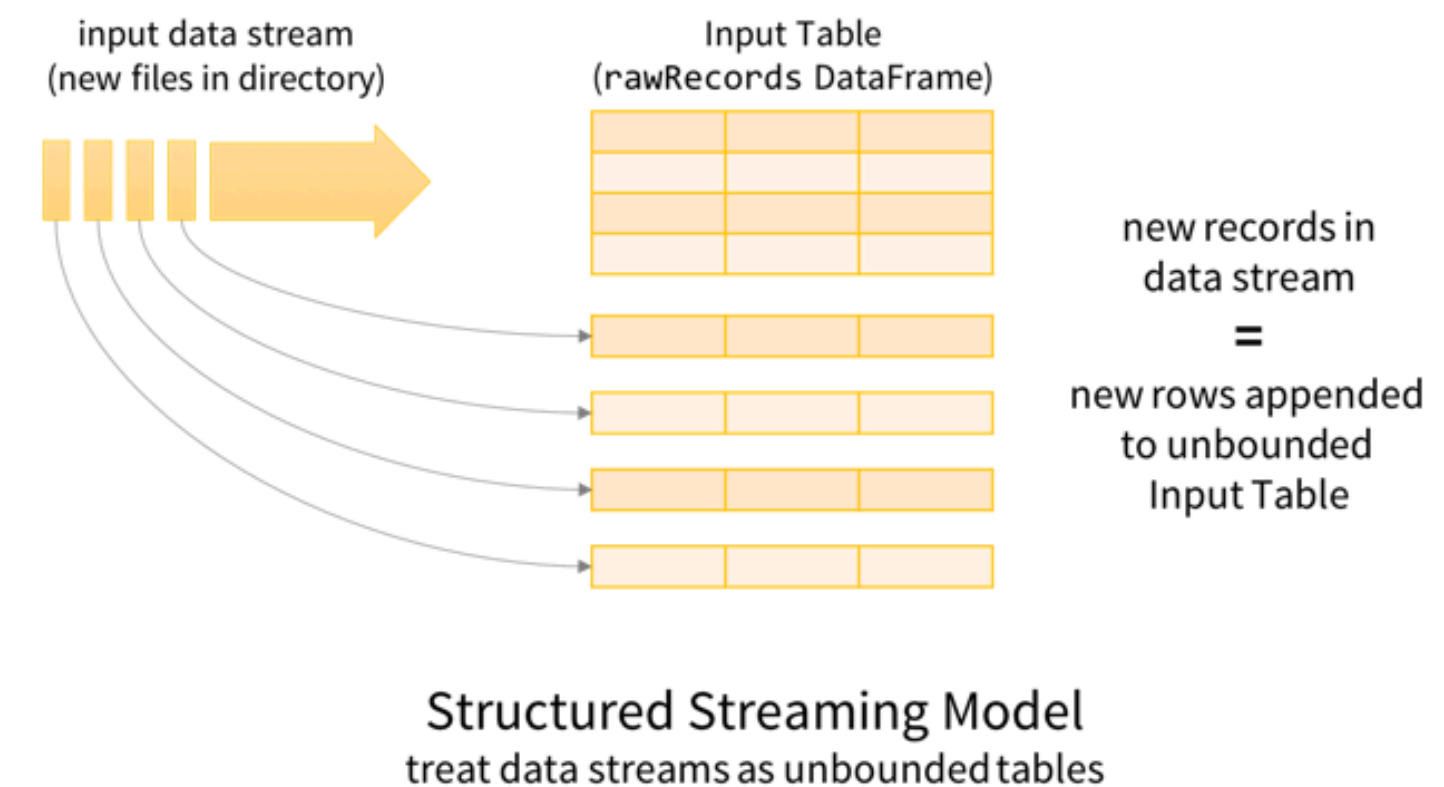
- Dataset[T]
 - All benefits of DataFrame
 - DataFrame = Dataset[Row]
- Typed



demo DataFrame & Dataset

Structured Streaming

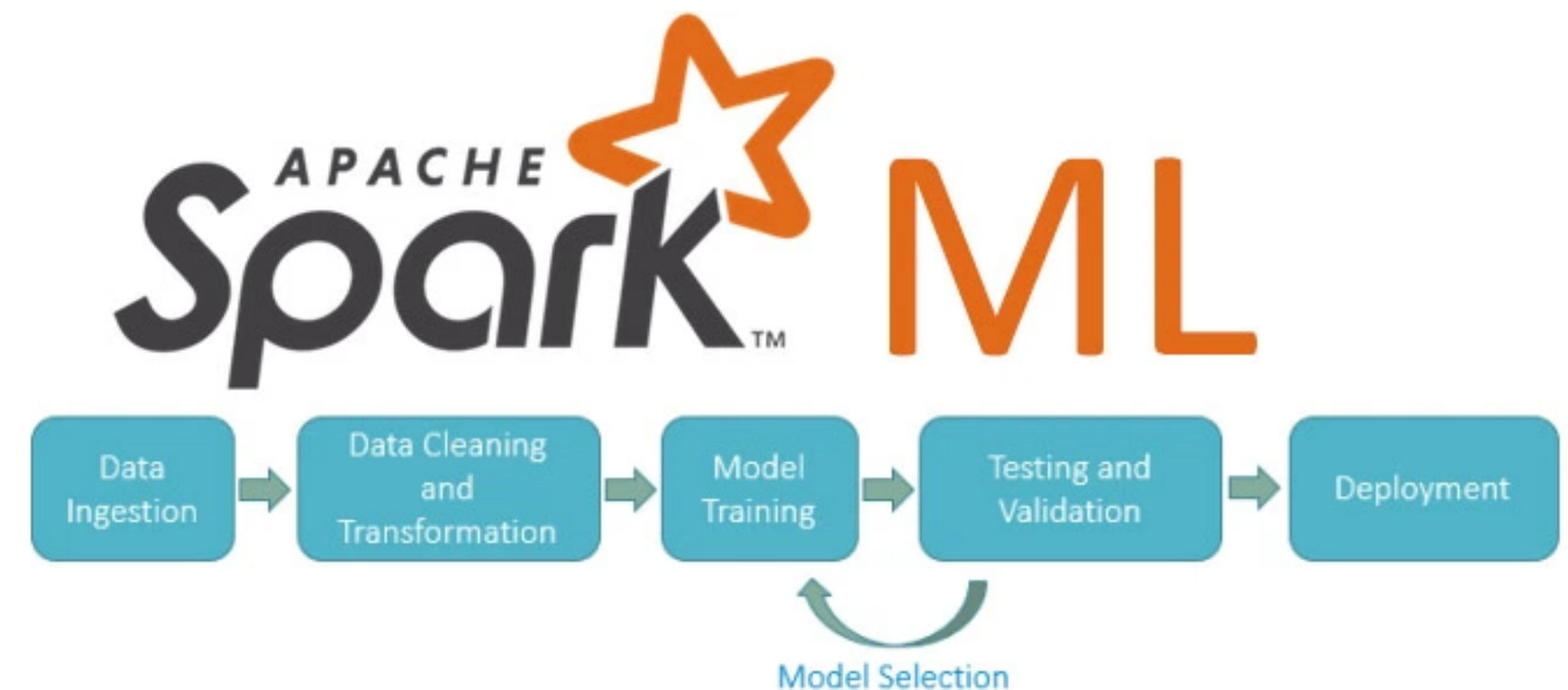
- Similar to Dataset / DataFrame processing
- end-to-end exactly once
 - checkpointing
 - write ahead logs
- < 100ms latency
 - or <1 ms in continuous mode



demo Structured Streaming

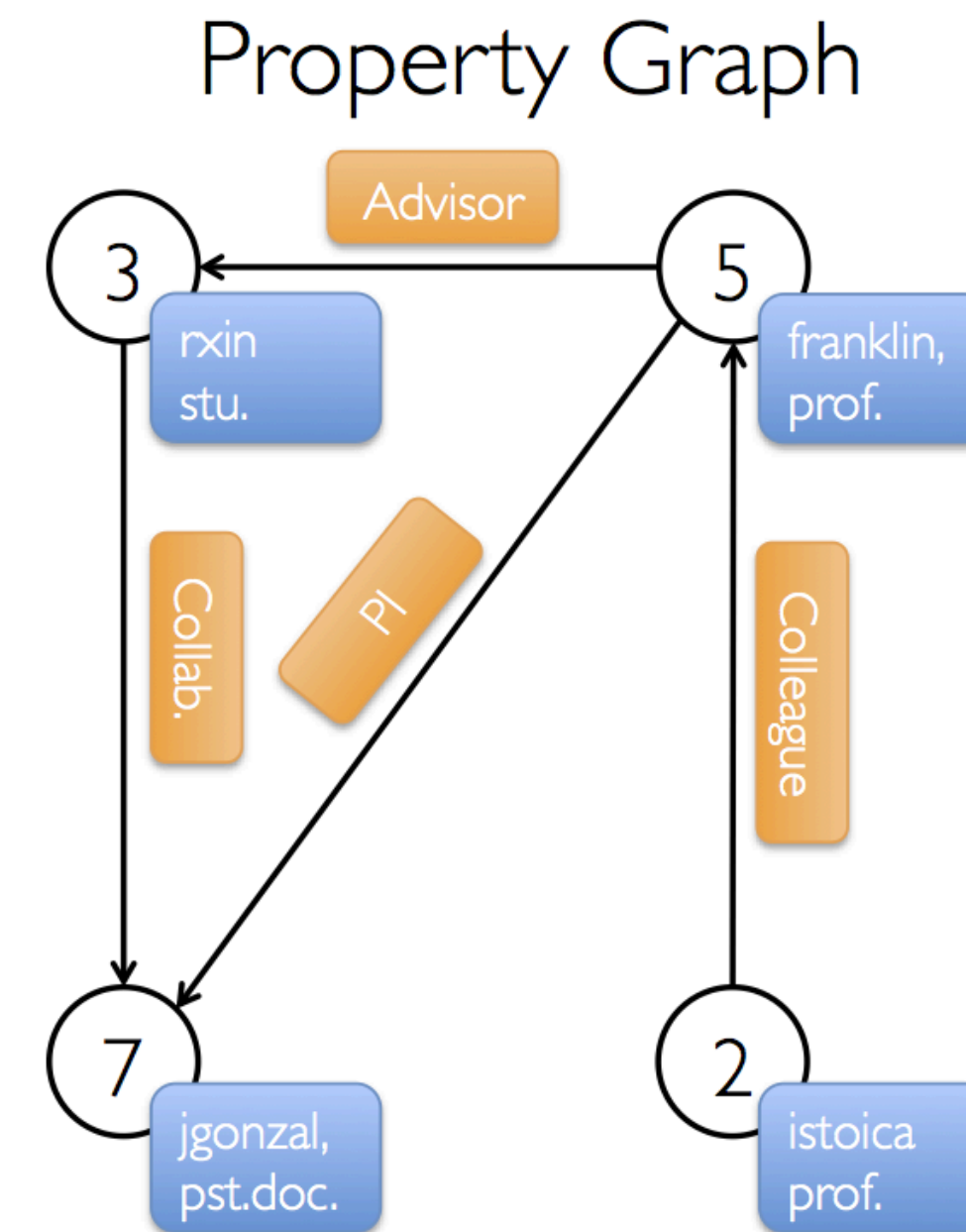
MLlib

- ML Algorithms
 - classification, regression, clustering, and collaborative filtering
- Featurization
 - feature extraction, transformation, dimensionality reduction, and selection
- Pipelines
- Persistence
- Utilities
 - linear algebra, statistics, data handling



GraphX

- Abstracts over RDD
- Directed multigraph (Vertex / Edge)
- Powerful distributed functions
 - Connected Components
 - Pagerank
 - Triangle Counting
 - Pregel API



Vertex Table

| Id | Property (V) |
|----|-----------------------|
| 3 | (rxin, student) |
| 7 | (jgonzal, postdoc) |
| 5 | (franklin, professor) |
| 2 | (istoica, professor) |

Edge Table

| SrcId | DstId | Property (E) |
|-------|-------|--------------|
| 3 | 7 | Collaborator |
| 5 | 3 | Advisor |
| 2 | 5 | Colleague |
| 5 | 7 | PI |

Deployment

Local

- `master=local[n]`
- Spark driver is launched in same process
- Workers are launched in same process

Standalone

- `master=spark://host`
- Spark driver is launched on a cluster machine
- Workers are deployed as separate machines
- submit jar to nodes

demo MLlib & deployment

Done

You can now build your own
streaming predictor of stocks :-)

