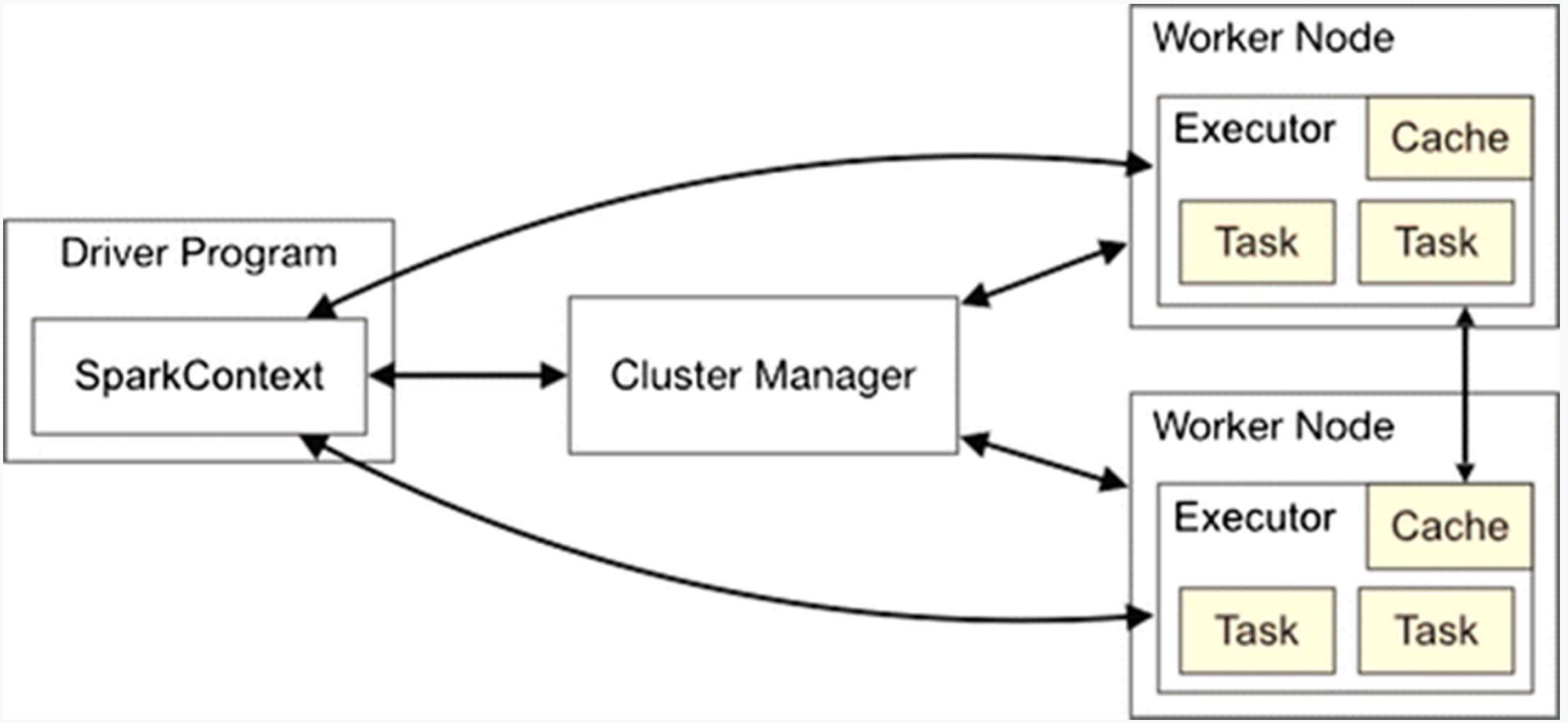
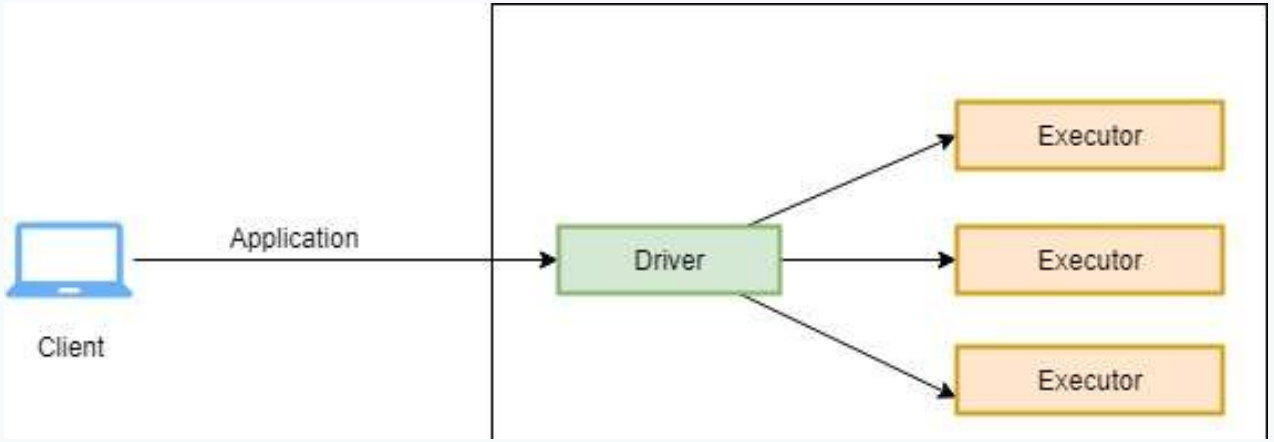
Apache Spark

Spark Architecture

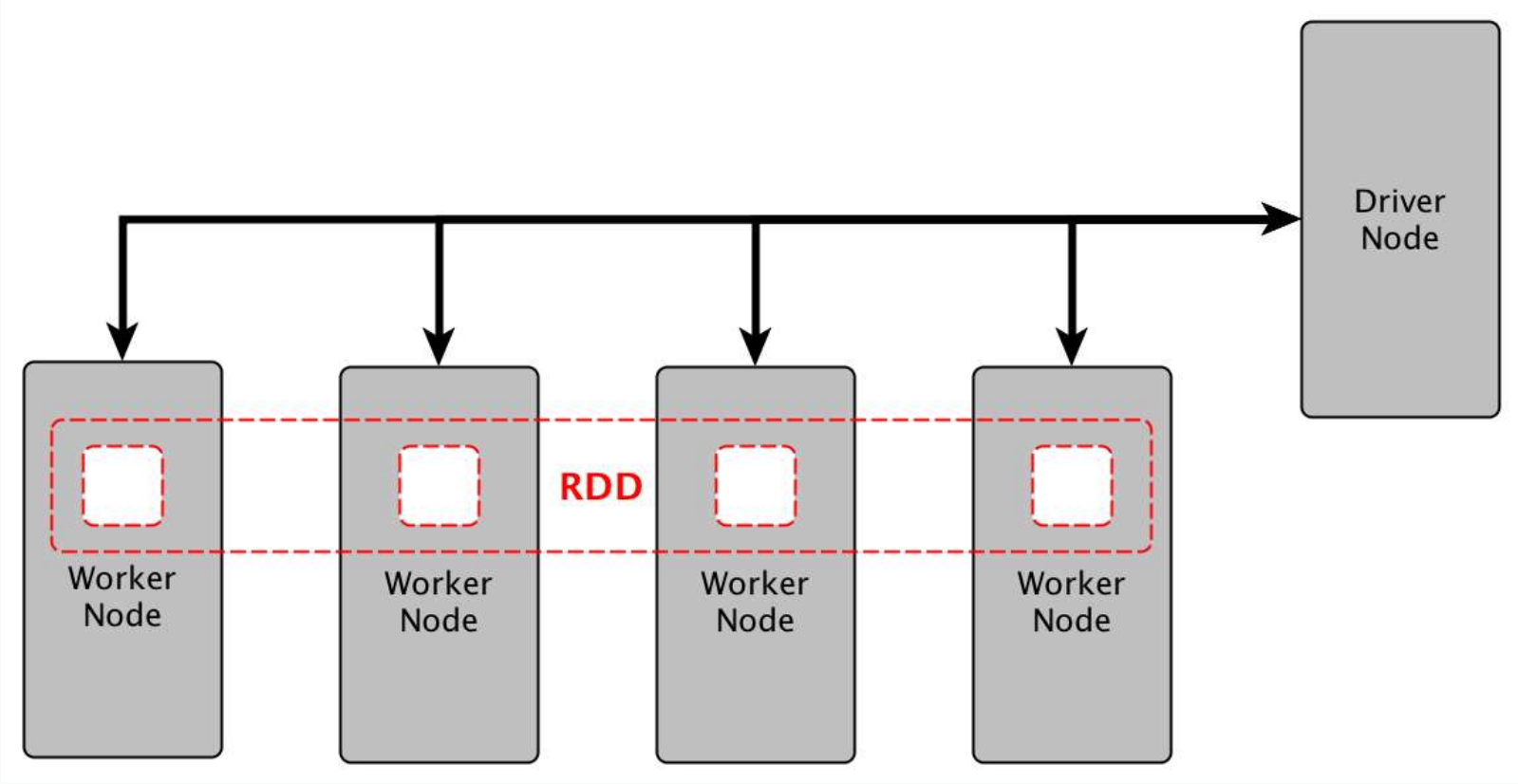


Apache Spark Execution

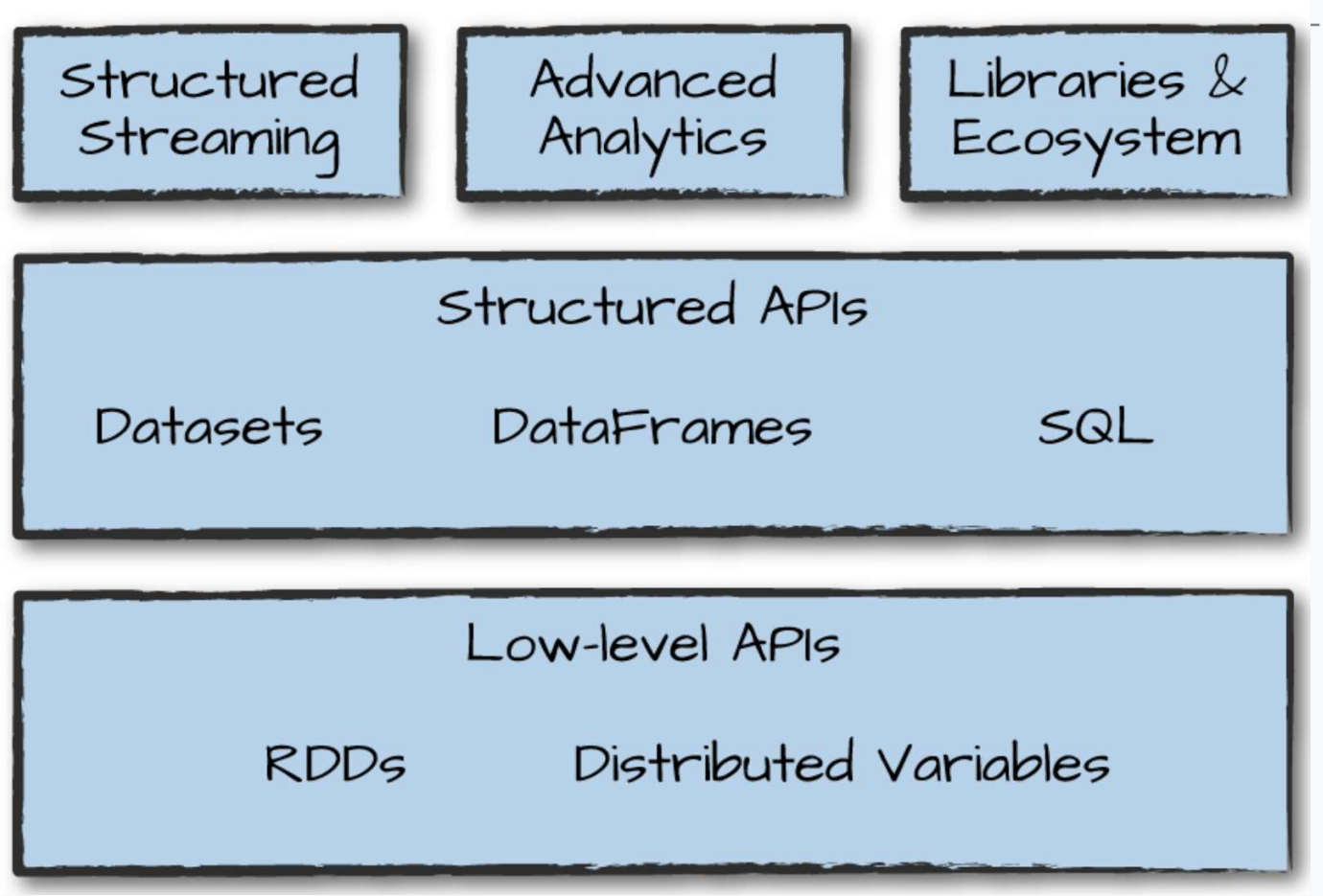
- For every application submitted on spark cluster, spark creates a dedicated Driver process and bunch of Executor processes.
- Driver process is responsible for analyzing, distributing, scheduling and monitoring of executor processes.
- Whereas the executor process is only responsible for running the task they were assigned by drivers and reporting the status back to the driver.



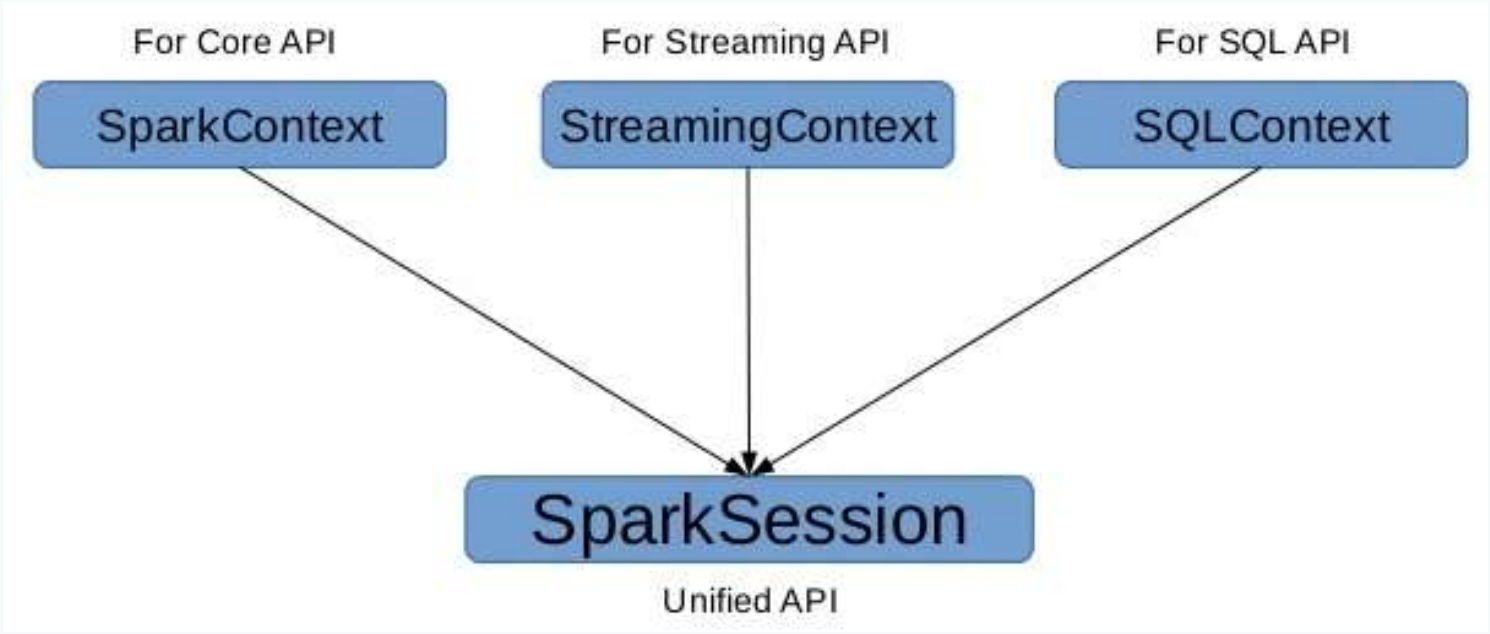
Apache Spark Execution



Spark's Language APIs



The SparkSession



DataFrames

- In Apache Spark, a DataFrame is a distributed collection of rows
- It has below characteristics:
  - Immutable in nature
    - We can create DataFrame RDD once but can’t change it.
  - Lazy Evaluations
    - Which means that a task is not executed until an action is performed.
  - Distributed

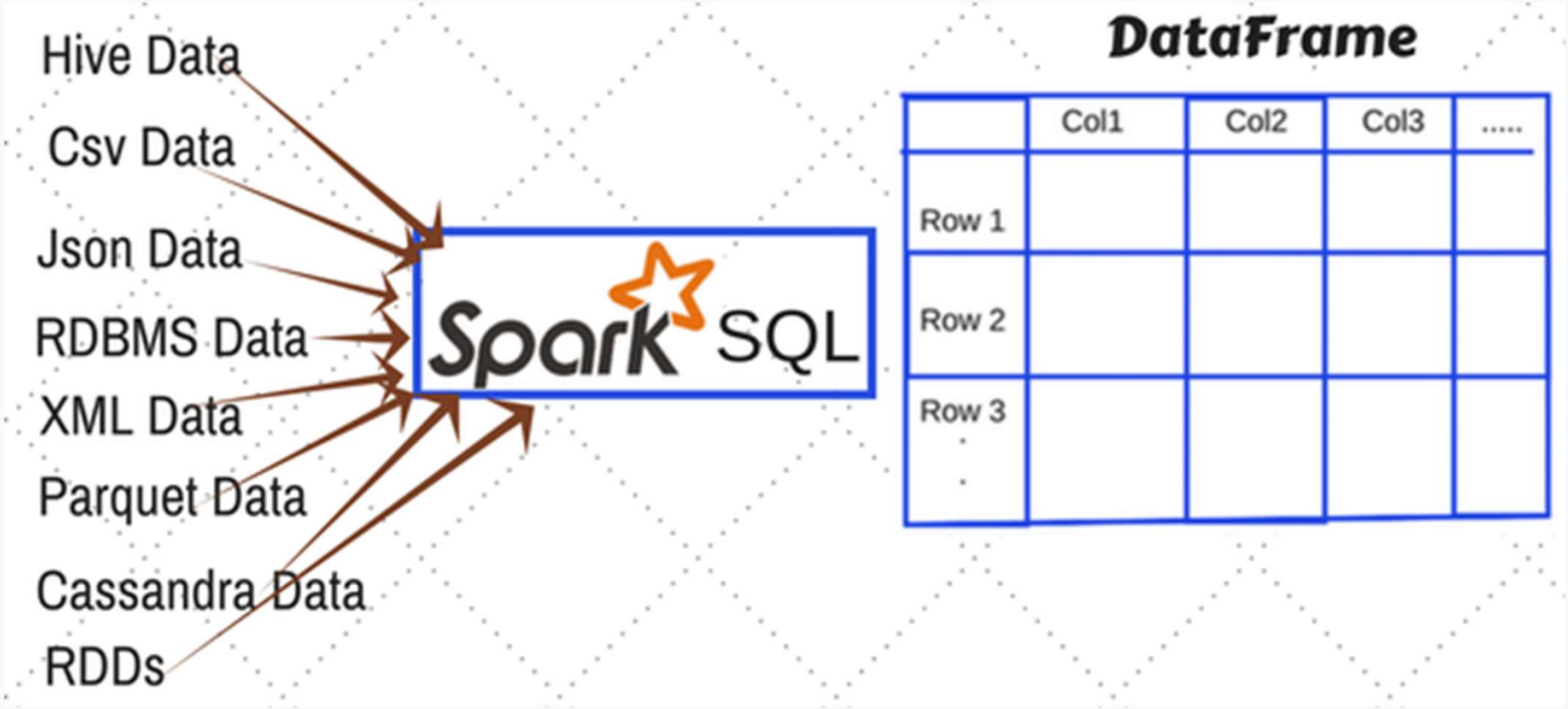
Why DataFrames are Useful?

- Designed for processing large collection of data.
- Has the ability to handle petabytes of data.
- Has API support for different languages like
  - Python,
  - R,
  - Scala,
  - Java.

Create a DataFrame

- Can be created using different data formats:
  - JSON
  - CSV
  - XML
  - Excel
- By loading data from Existing RDD
- By Programmatically specifying schema

Ways to create DataFrame in Spark

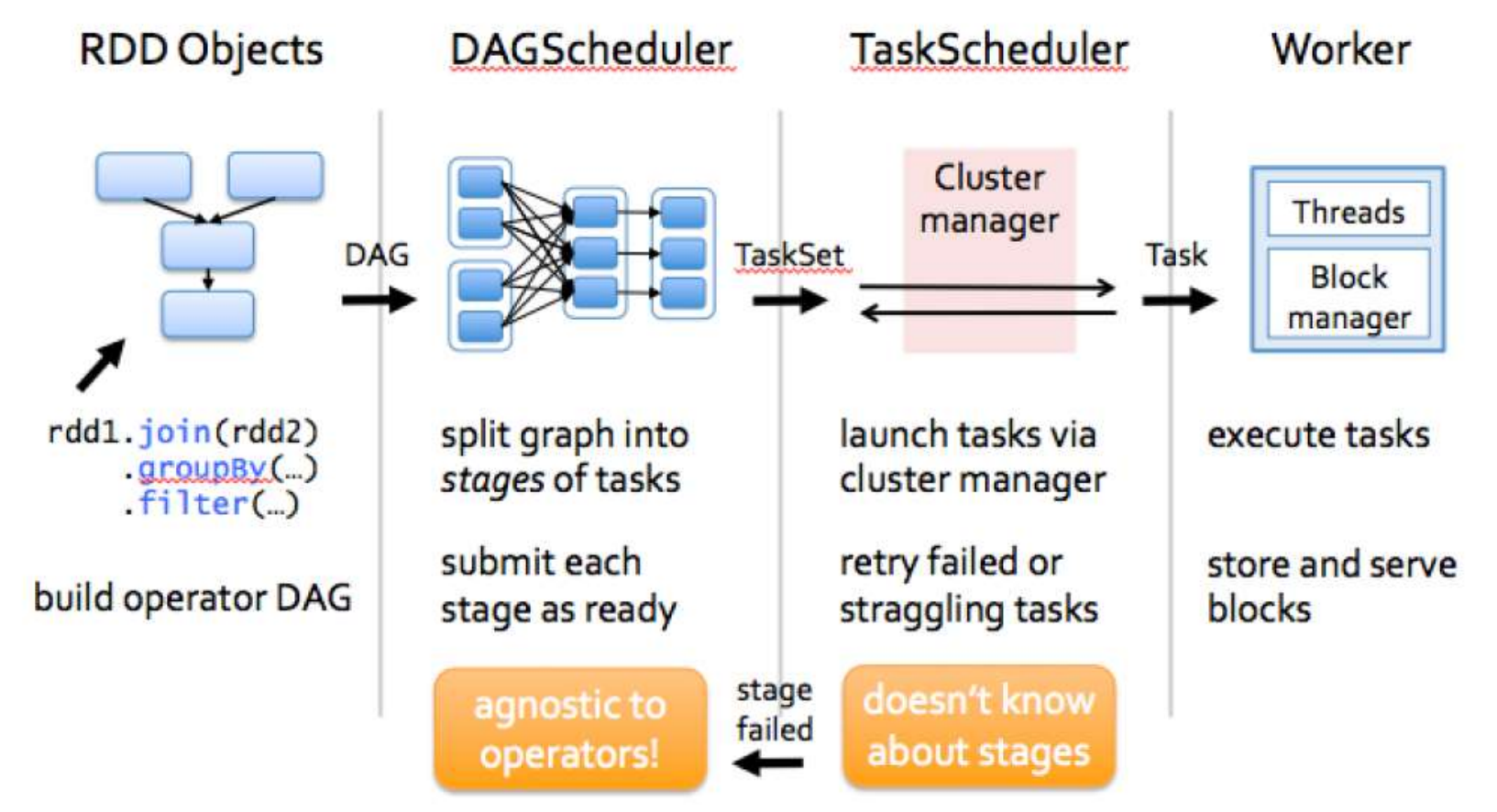


Creating DataFrame from RDD

- from pyspark.sql import Row
- l = [('Ankit',25),('Jalfaizy',22),('saurabh',20),('Bala',26)]
- rdd = sc.parallelize(l)
- people = rdd.map(lambda x: Row(name=x[0], age=int(x[1])))
- schemaPeople = sqlContext.createDataFrame(people)

- type(schemaPeople)
- #Output:
- #pyspark.sql.dataframe.DataFrame

How Sparks work?



Catalyst Optimizer

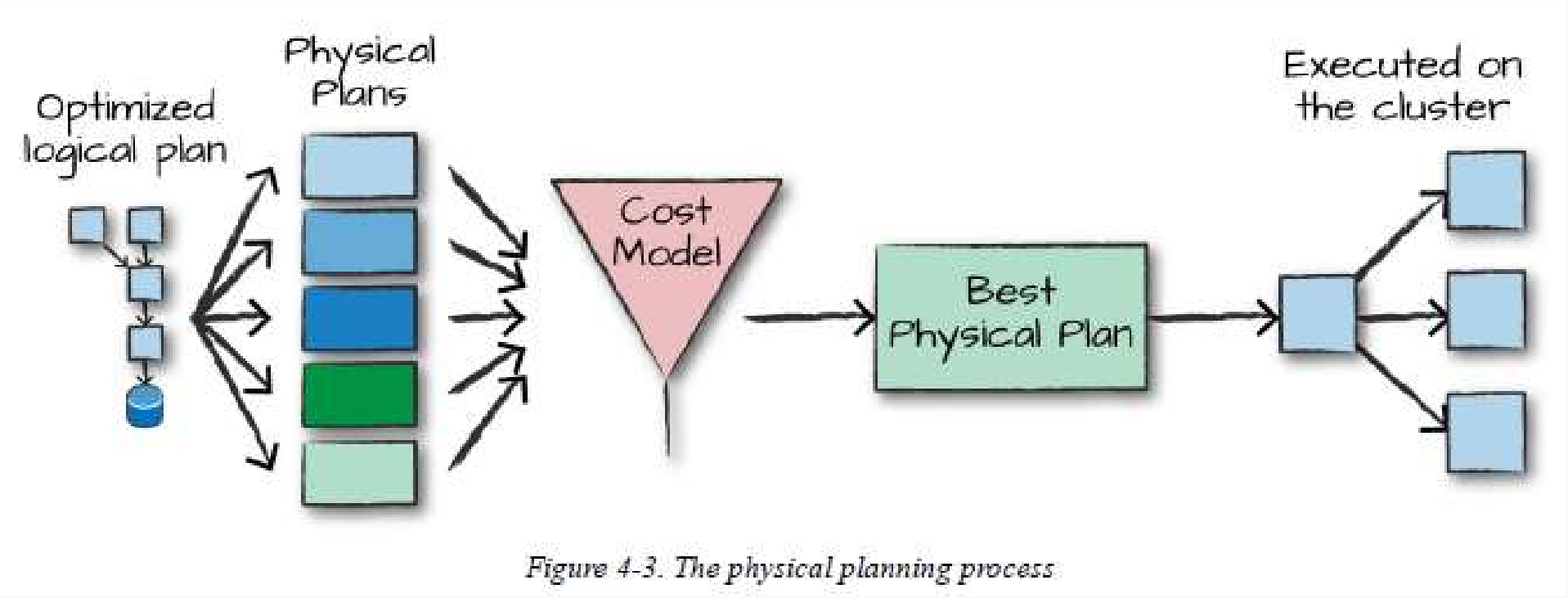


Figure 4-3. The physical planning process

Thank You