

# Join Mechanics



# Objective

Recap essential join concepts

Understand why joins are (usually) slow and expensive

Learn how partitioning plays a critical role in joins



# Joins

## Combine the data in multiple DataFrames/RDDs

- rows are combined
- join condition: only the rows passing the condition are kept

## Join types

- inner: combine only the rows passing the condition
- left\_outer: inner + all rows in the left "table", with nulls in the corresponding fields of the other "table"
- right\_outer: same, for the right "table"
- full\_outer = same, for both "tables"

## More join types (DFs only)

- left\_semi = all the rows in the left DF for which there is a row in the right DF passing the condition
- left\_anti = all the rows in the left DF for which there is NO row in the right DF passing the condition

\* "table" = DataFrame or RDD

# Why Are Joins Slow?

If DFs or RDDs don't have a known partitioner, a shuffle is needed

- data transfer overhead
- potential OOMs
- limited parallelism

## Co-located RDDs

- have the same partitioner
- reside in the same physical location in memory (on the same executor)
- can be joined without any network transfer

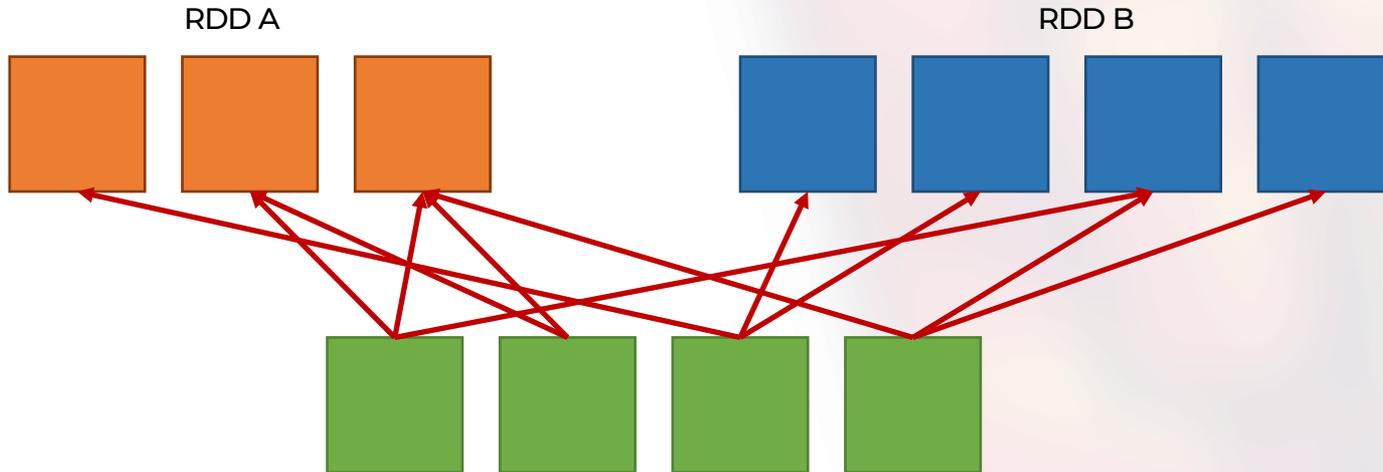
## Co-partitioned RDDs

- have the same partitioner
- may be on different executors
- will (in general) be joined with network traffic
  - although much less than without the partitioning information

# Shuffled Join

No partitioner is known

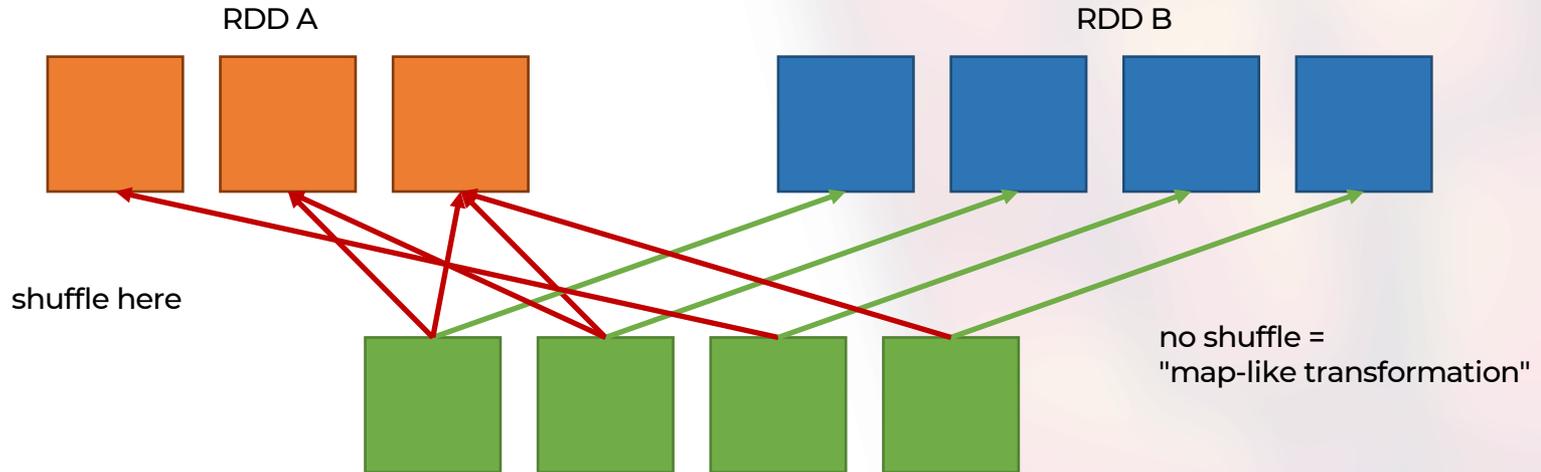
- Rows with the same key must be on the same partition
- Spark needs to shuffle both RDDs
- VERY expensive



# Optimized Join

One RDD doesn't have a known partitioner, or partitioners are different

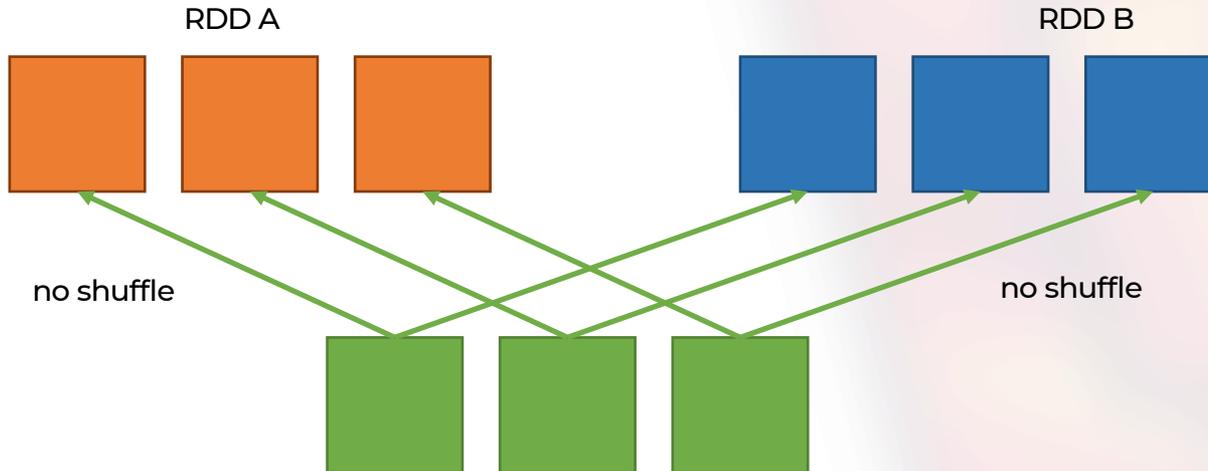
- we can force the other RDD to obey the same partitioner
- one shuffle instead of two



# Optimized Join +

Both RDDs have the same partitioner (co-partitioned)

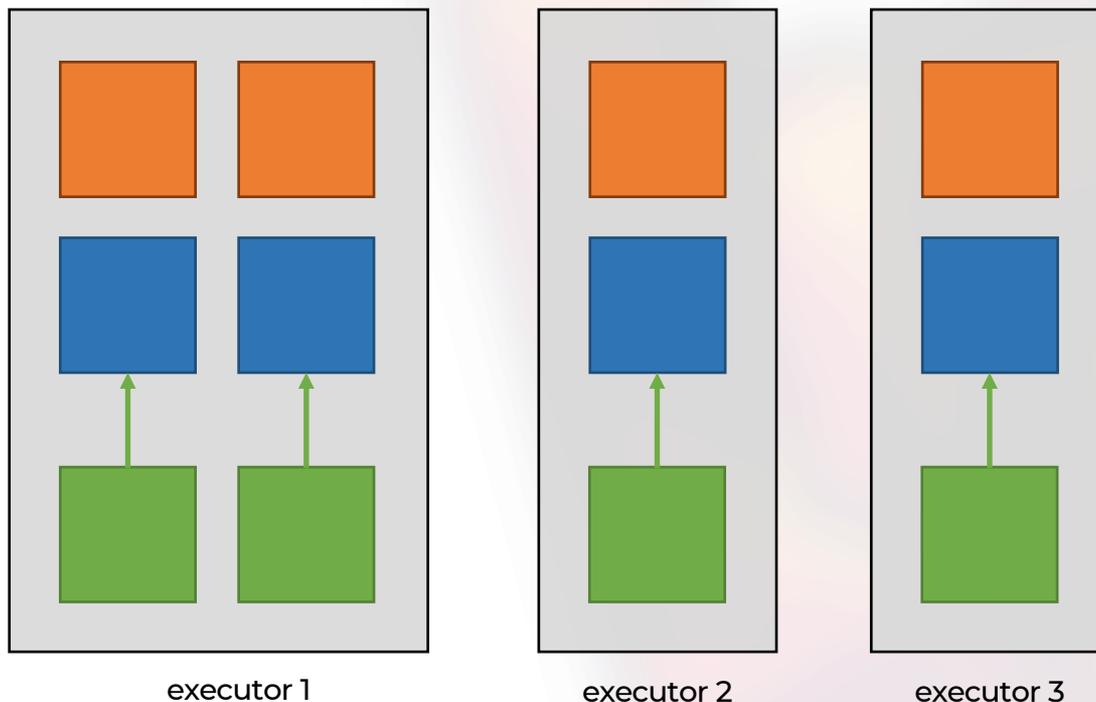
- just fetch the existing partitions and do the join
- no shuffles
- narrow dependency



# Optimized Join ++

Same partitioners, partitions loaded in memory (co-located)

- no partition fetching
- no shuffle
- no network transfer
- fastest joins possible



# Join Mechanics

Shuffling, colocation & copartitioning apply to RDDs and DFs

This chapter: joins on DFs

Techniques apply to grouping as well

Next chapter: joins on RDDs

**Spark rocks**

