

CS 378 – Big Data Programming

Lecture 16

Join Patterns

Review

- Assignment 7 – Reduce-side join
 - User session and impression data
- Questions/issues?

Join Patterns

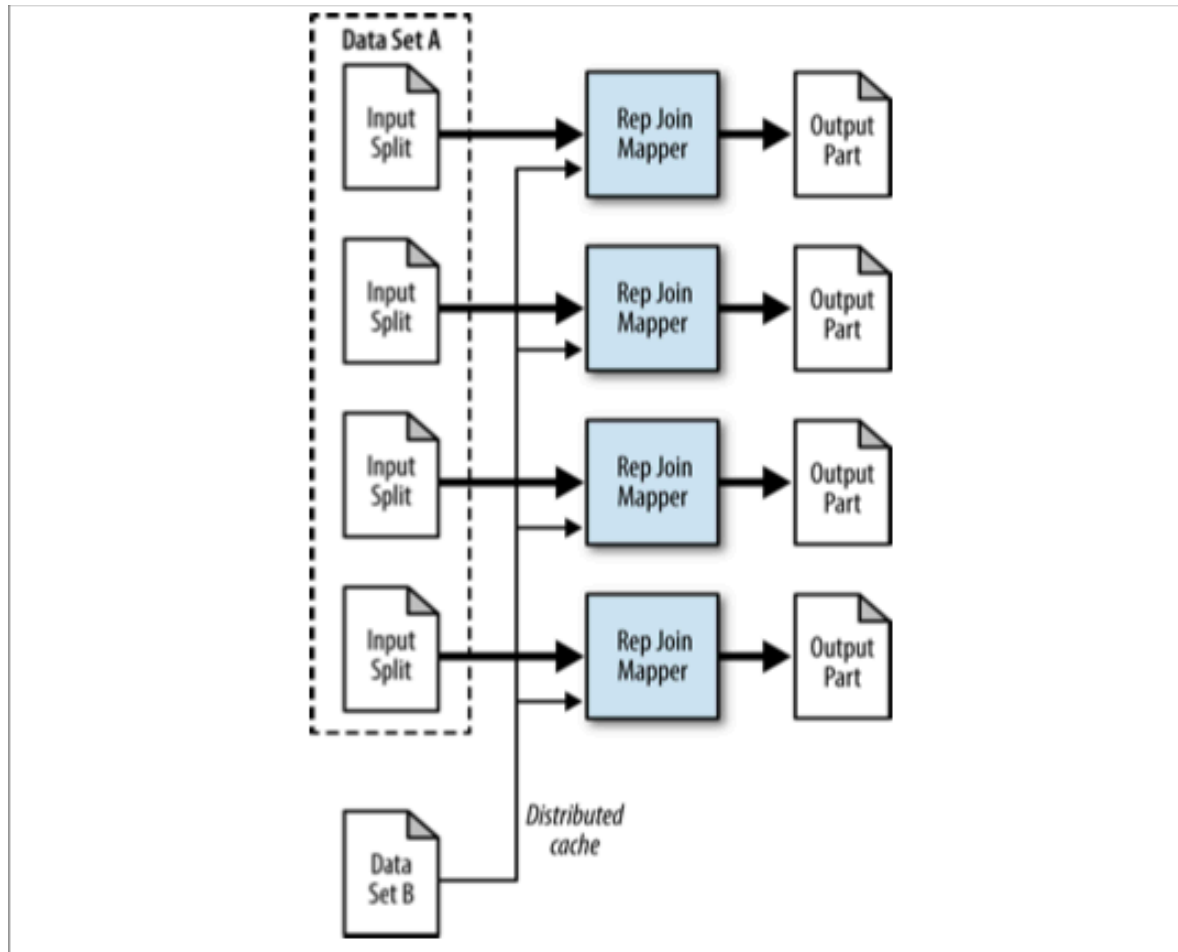
- Review: Suppose we want to join many sources, only one of which is large
 - User sessions (large)
 - Map from cities to DMA (demographic marketing area)
 - ...
- This is called a *replicated* join
 - All the small files will be replicated to all machines

Replicated Join

- Can be done completely in mappers
 - No need for sort, shuffle, or reduce
 - Files are replicated with `DistributedCache`
- Restrictions:
 - All but one of the inputs must fit in memory
 - Can only accomplish an inner join, or
 - A left outer join where the large data source is “left” part

Replicated Join - Data Flow

Figure 5-2 from MapReduce Design Patterns



Join Patterns

- OK, so replicated join was interesting, but more than one of my data sources is large.
- Is there a way to do a map-side join in this case?
- Or is reduce-side join my only option?

- If we organize the input data in a specific way,
- We can do this on the map-side.

Composite Join

- Hadoop class `CompositeInputFormat`
- Restricted to inner, or full outer join
- Input data sets must have the same # of partitions
 - Each input partition must be sorted by key
 - All records for a particular key must be in the same partition
- Seems pretty restrictive ...

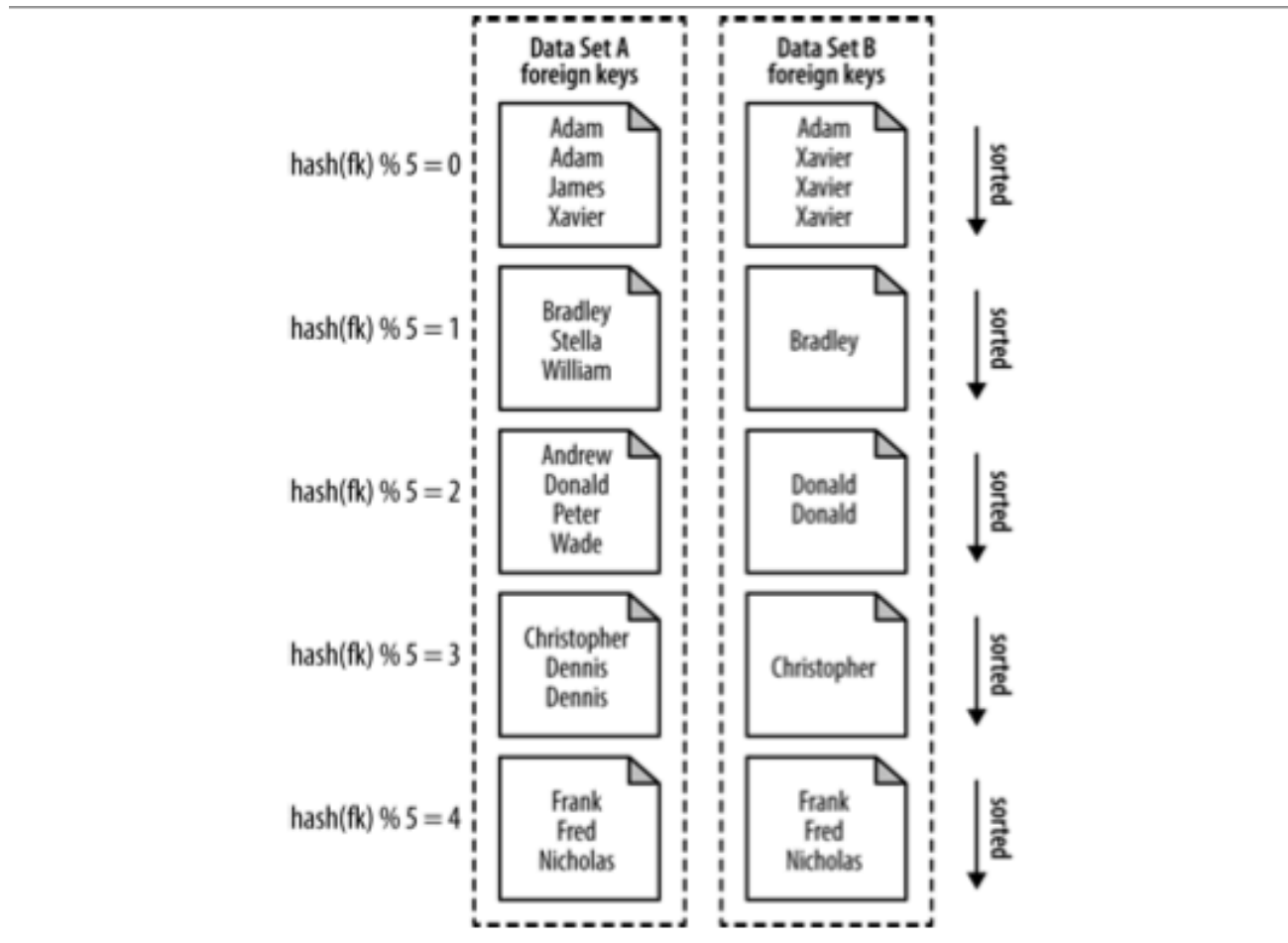
Composite Join

- These conditions might exist for data from other mapReduce jobs where:
 - The jobs had the same # of reducers
 - Recall that input data sets must be partitioned in same way
 - The jobs had the same foreign key
 - Output files aren't splittable

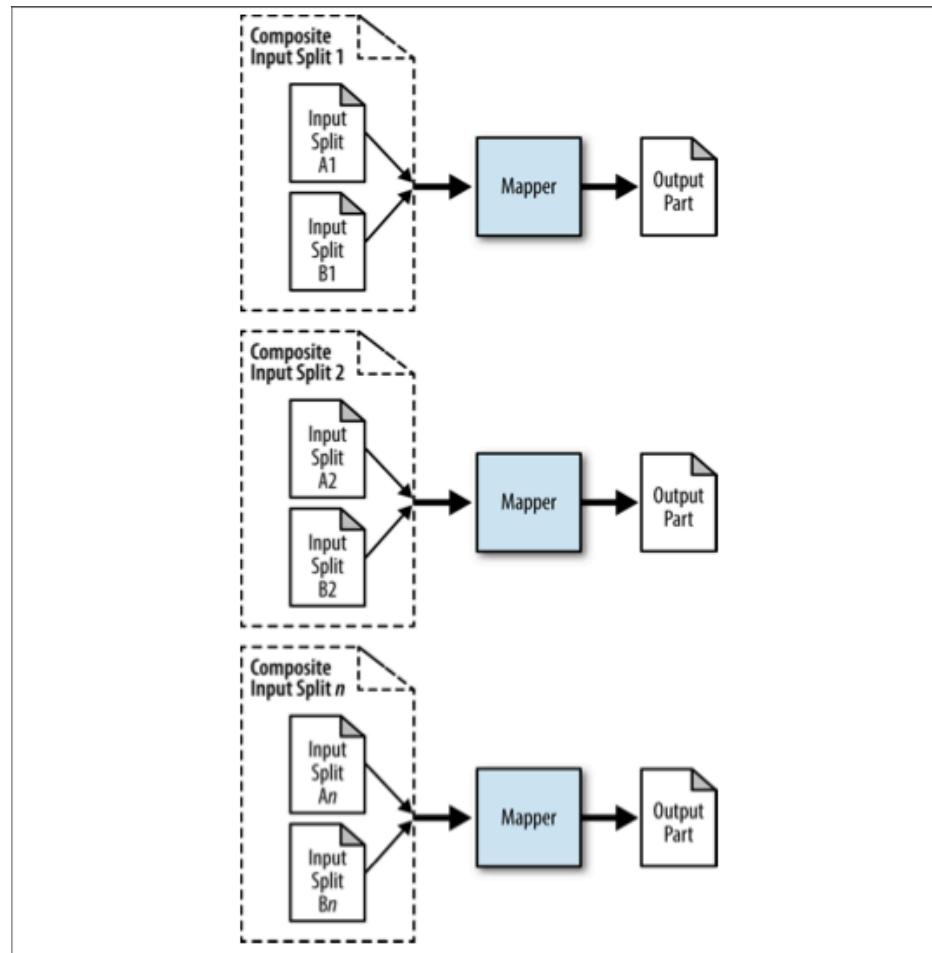
Composite Join

- If all those conditions are true, this join works
 - Map-side only, so it's efficient if we can use it.
- If you find that you are preparing and formatting the data only to be able to use composite join
- It's probably not worth it.
- Just use a reduce-side join.

Composite Join – Data



Composite Join – Data Flow



CompositeJoinInput

- In the driver code (`run()` method)
 - Get the file names from the command line
 - Specify the input format, join type, and files

```
conf.setInputFormat(CompostieInputFormat.class);
```

```
conf.set("mapred.join.expr",  
        CompositeInputFormat.compose("inner",  
        KeyValueTextInputFormat.class, file1, file2));
```

CompositeJoinInput

- How might this implement inner join?
- Outer join?
- Could we do any other join type?
 - Left outer? Anti-join?

One More Join Pattern

- Suppose we wanted to compare all cars currently available (for sale) to all other cars
 - To identify “similar” cars
 - Usage: “I like this car, show me others like it”
- This join is called “Cartesian Product”
 - Compare N items to M items requires $N \times M$ comparisons
 - Not straightforward to do with map-reduce

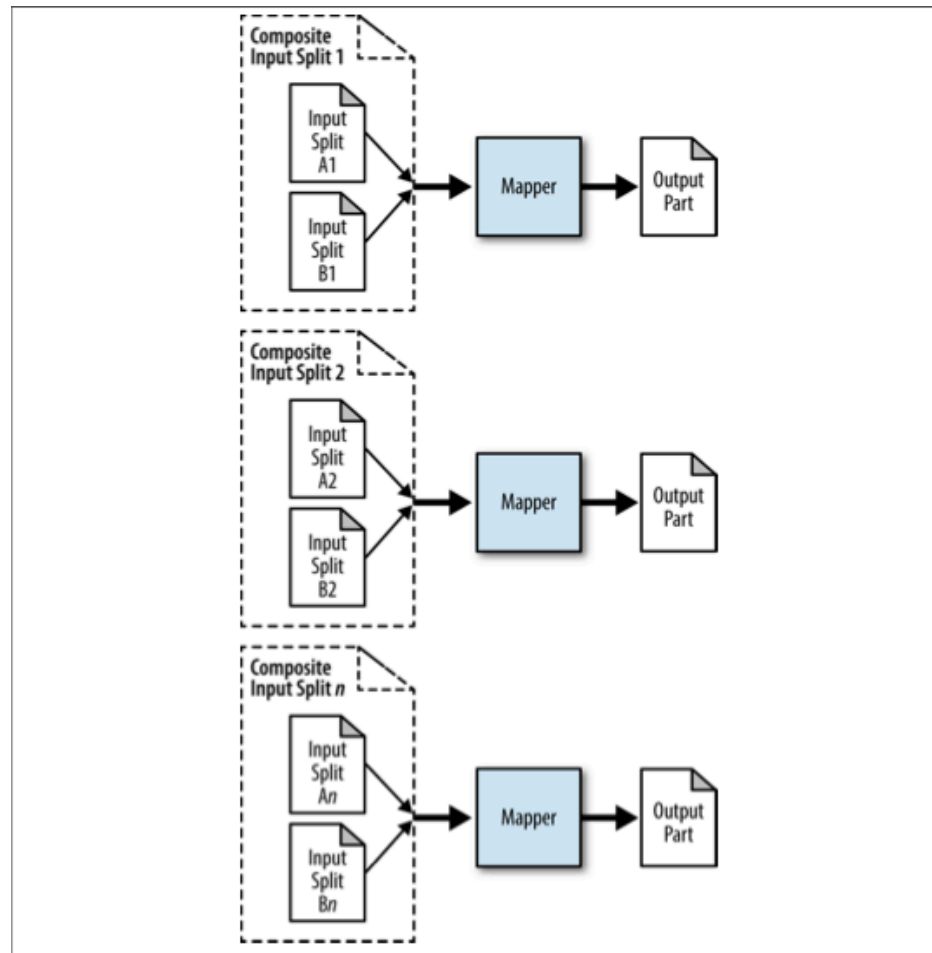
Cartesian Product

- Pairs every record with every other record
 - No keys needed
 - $N \times M$ results, for datasets of size N, M
- Map-only job
- But still expensive to compute
- Hadoop class: `CartesianInputFormat`

Cartesian Product

- To accomplish this join, we'll need to pair every record with every other record
- We can start with the approach for composite join
- For composite join, each mapper read two files
 - They had the same key set
 - The data was sorted by key
 - We don't care about the keys, just the 'two file input'

Composite Join – Data Flow



One Mapper, Two Inputs

- For composite join, the key order allowed us to:
 - Read each of the two files only once
 - Worked very much like merge sort
- For Cartesian product
 - For each record in data set 1
 - We'll read every record in data set 2
 - This pair of records is passed to the mapper
- We'd accomplish this with a custom input format
 - RecordReader resets data set 2 for each input of data set 1

Cartesian Product – Data Flow

