

#### **Cloud Computing**

#### **MapReduce Programming Model**

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# Programming Platforms for Big Data

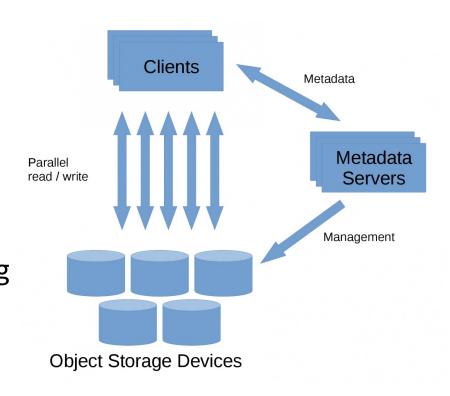
- ➤ Map-Reduce: **Apache Hadoop**, Aneka
- >Stream Processing: Heron, Apache Storm, Apache Spark
- Graph Processing: Pregel, Apache Giraph



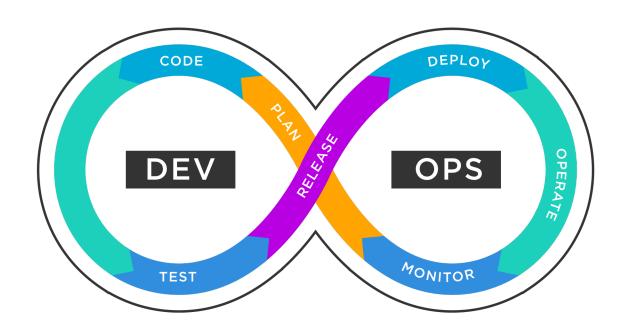
- ➤ Introduced by Google for processing large quantities of data.
- It expresses the computational logic of an application in two simple functions:
  - map
  - reduce



- Data transfer and management are completely handled by the Distributed Storage Infrastructure.
  - E.g., the Google File System
- ➤ Distributed Storage is in charge of providing access to data, replicating files, and eventually moving them where needed.



- Developers no longer have to handle storage issues.
- Developers are provided with an interface that presents data at a higher level: as a collection of key-value pairs.



The computation of MapReduce applications is then organized into a workflow of **map** and **reduce** operations that is entirely controlled by the runtime system.

Developers need only specify how the map and reduce functions operate on the key-value pairs.

- ➤ MapReduce model is expressed in the form of the two functions:
  - map( $k_1, v_1$ )  $\rightarrow$  list( $k_2, v_2$ )
  - Reduce( $k_2$ , list( $v_2$ ))  $\rightarrow$  list( $v_3$ )

➤ Map takes an input pair and produces a set of intermediate key/value pairs.

The MapReduce library groups together all intermediate values associated with the same intermediate key / and passes them to the Reduce function.

The Reduce function accepts an intermediate key *I* and a set of values for that key.

- ➤ It **merges** together these values to form a possibly smaller set of values.
  - Typically just zero or one output value is produced per Reduce invocation.

#### ➤ The framework signature is:

• computation: list(k1, v1)  $\rightarrow$  list(k2, v2)

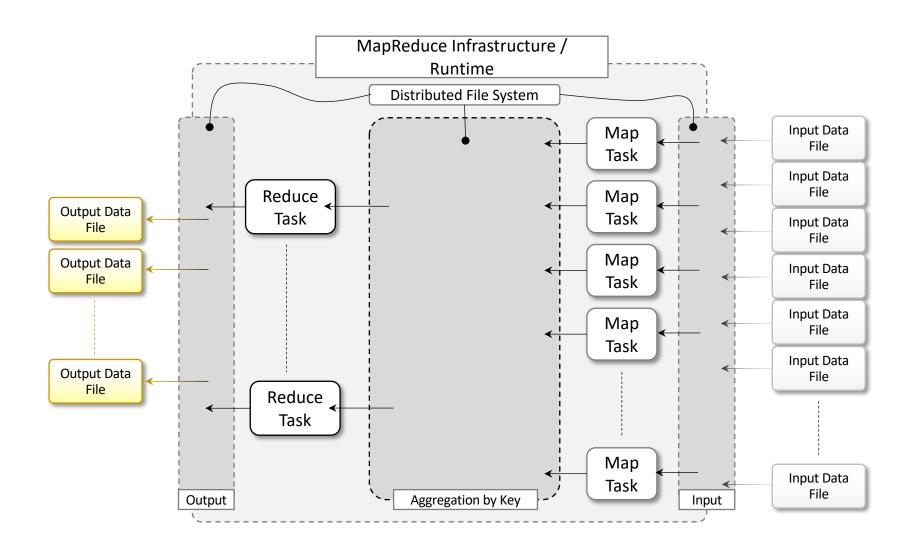
#### ➤ Map Phase

• MapPhase:  $list(k1, v1) \rightarrow list(k2, v2)$ 

#### **≻** Reduce Phase

- ReducePhase = preparation; reduction
- Preparation:  $list(k2, v2) \rightarrow list(k2, list(v2))$
- Reduction: list(k2, list(v2)) → list(k2, v3)

# MapReduce Framework



### MapReduce Example

Problem: counting the number of occurrences of each word in a large collection of documents

```
map(String key, String value):
    // key: document name
    // value: document
contents
    for each word w in value:
        EmitIntermediate(w, "1")
```

```
reduce(String key, Iterator values):
   // key: a word
   // values: a list of counts
   int result = 0;
   for each v in values:
      result += ParseInt(v);
   Emit(AsString(result));
```

### More MapReduce Examples (Cont.)

#### **➤** Distributed Grep:

- Looking for a string in several files
- Print name of files that contains a string

#### Distributed Grep (Solution)

- The map function emits a line if it matches a supplied pattern.
- The reduce function is an identity function that just copies the supplied intermediate data to the output.

### More MapReduce Examples (Cont.)

#### **►** Count of URL Access Frequency:

How many time a URL is accessed given many web logs?

#### Count of URL Access Frequency (solution)

- The map function processes logs of web page requests and outputs <URL,1>.
- The reduce function adds together all values for the same URL and emits a <URL, total count> pair.

### More MapReduce Examples (Cont.)

#### **►** Inverted index:

 Given a set of documents, create the set of {word → list of documents containing the word}.

#### **►** Inverted Index (Solution)

- The map function parses each document, and emits a sequence of <word, document ID> pairs.
- The reduce function accepts all pairs for a given word, sorts the corresponding document IDs and emits a <word, list(document ID)> pair.
- The set of all output pairs forms a simple inverted index.

➤ In general, any computation that can be expressed in the form of two major stages can be represented in the terms of **MapReduce** computation:

- Analysis
- Aggregation

### Analysis

➤ Analysis operates directly on the data input file and corresponds to the operation performed by the map task.

➤ The computation at this stage is expected to be embarrassingly **parallel**, since map tasks are executed without any sequencing or ordering.

#### Aggregation

➤ Aggregation operates on the intermediate results and is characterized by operations that are aimed at aggregating, summing, and/or elaborating the data obtained at the previous stage to present the data in their final form.

> This is the task performed by the reduce function.