

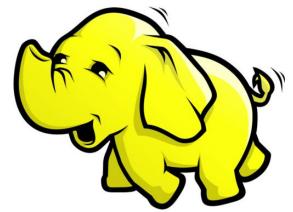
IT Systems Engineering | Universität Potsdam

Übung Datenbanksysteme II

Web-Scale Data Management

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Folien basierend auf Thorsten Papenbrock



Introduction



2

- MapReduce ...
 - is a paradigm derived from functional programming.
 - is implemented as framework.
 - operates primarily data-parallel (not task-parallel).
 - scales-out on multiple nodes of a cluster.
 - uses the Hadoop distributed filesystem.
 - is designed for Big Data Analytics:
 - Log-files
 - Weather-statistics
 - Sensor-data
 - ...
- "Competitors":



Stratosphere

3

MapReduce:

Introduction

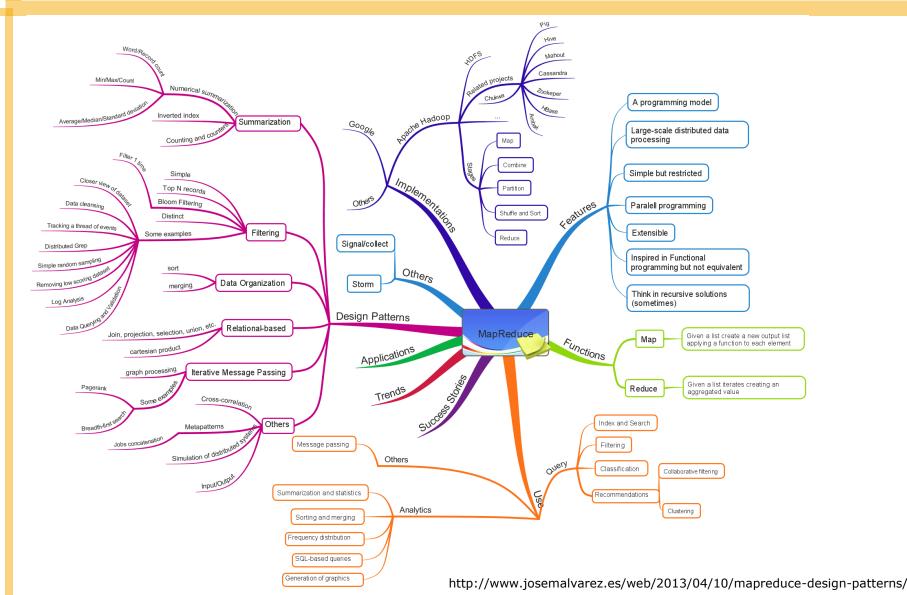


Who is using Hadoop?

- Yahoo!
 - Biggest cluster: 2000 nodes, used to support research for Ad Systems and Web Search.
- Amazon
 - Process millions of sessions daily for analytics, using both the Java and streaming APIs. Clusters vary from 1 to 100 nodes.
- Facebook
 - Use Hadoop to store copies of internal log and dimension data sources and use it as a source for reporting/analytics.
 600 machine cluster.
- •

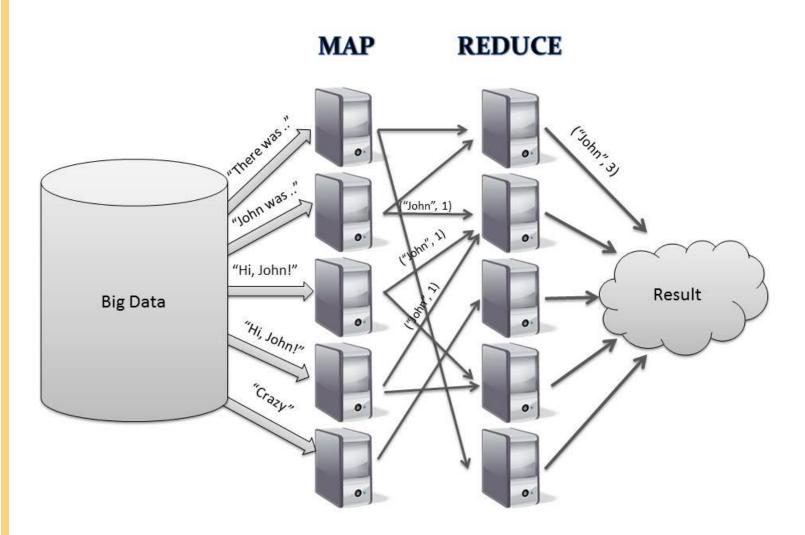
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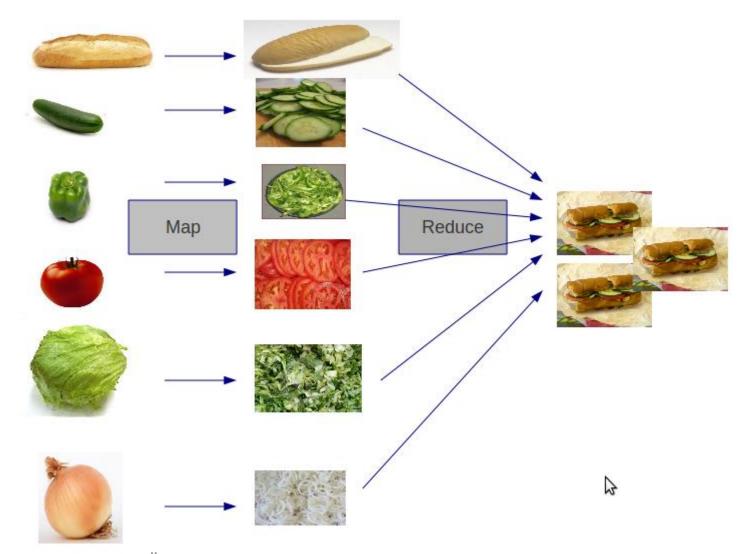
Introduction





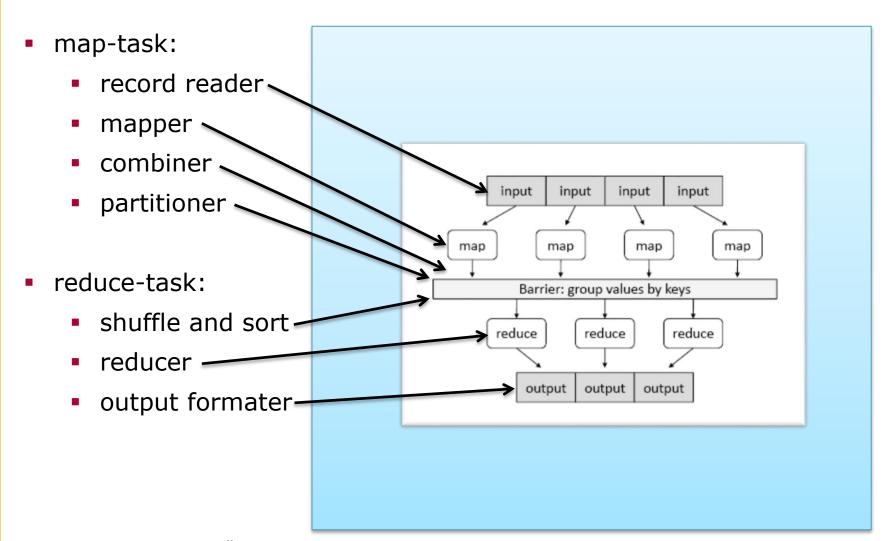
Introduction





Phases





Phases



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input: <data entry> (row/split/item)
- Output: <key, record>
- "key" is usually positional information
- "record" represents a raw data record
- Translates a given input into records
- Parses data into records but not the records itself

Phases



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input: <key, record>
- Output: <key*, value>
- "key*" is a problem-specific key
 - e.g. the word for the word-count-task
- "value" is a problem-specific value
 - e.g. "1" for the occurence of a word
- Executes user defined code that starts solving the given task
- Defines the grouping of the data
- A single mapper can emit multiple
 <key*, value> output pairs for a single
 <key, record> input pair

Phases



- map-task:
 - record reader
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 - reducer
 - output formater

- Input: <key*, values>
- Output: <key*, value>
- "key*" is a problem-specific key
 - e.g. the word for the word-count-task
- "value" is a problem-specific value
 - e.g. "1" for the occurence of a word
- Executes user defined code that merges a set of values
- Pre-aggregates values to reduce network traffic
- Is an optional, localized reducer

Phases



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input: <key*, value>
- Output: <key*, value> + reducer
- "reducer" is the reducer number that should handle this key/value pair; reducer might be located on other compute nodes
- Distributes the keyspace randomly to the reducers
- Calculates the reducer by e.g. key*.hashCode() % (number of reducers)

Phases



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input: <key*, value> + reducer
- Output: <key*, value> + reducer
- Downloads the <key*, value> data to the local machines that run the corresponding reducers

Phases



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input: <key*, values>
- Output: <key*, result>
- "result" is the solution/answer for the given "key*"
- Executes user defined code that merges a set of values
- Calculates the final solution/answer to the problem statement for the given key

Phases



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input: <key*, result>
- Output: <key*, result>
- Writes the key/result pairs to disk
- Formates the final result and writes it record-wise to disk

Phases



15

map-task: useful to increase record reader the performance mapper combiner helpful to build a partitioner sorting algorithm reduce-task: basic building blocks shuffle and sort with user defined code reducer

output formater

Example 1: Distinct



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input:
 - A relational table instance Car(name, vendor, color, speed, price)
- Output:
 - A distinct list of all vendors

```
map (key, record) {
    emit (record.vendor, null);
}

reduce (key, values) {
    write (key);
}
```

Example 2: Index-Generation



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input:
 - A relational table instance Car(name, vendor, color, speed, price)
- Output:
 - An index on Car.vendor

```
map (key, record) {
    emit (record.vendor, key);
}

reduce (key, values) {
    String refs = concat(values);
    write (key, refs);
}
```

Example 3: Join



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

- Input:
 - Two relational table instances
 Car(name, vendor, color, speed, price)
 Plane(id, weight, length, speed, seats)
- Output:
 - All pairs of cars and planes with the same speed

Example 3: Join



- map-task:
 - record reader
 - mapper
 - combiner
 - partitioner
- reduce-task:
 - shuffle and sort
 - reducer
 - output formater

```
Car(name, vendor, color, speed, price)
Plane(id, weight, length, speed, seats)
map (key, record) {
   emit (record.speed,
      {'table' => table(record),
        'record' => record});
}
reduce (key, values) {
   cars = valuesWhere('table', 'car');
   planes = valuesWhere('table', 'plane');
   for (car : cars)
      for (plane : planes)
         write (car.record, plane.record);
```

