Map-Reduce

Scheduling and Data Flow

Map-Reduce: A diagram

MAP:

Read input and produces a set of

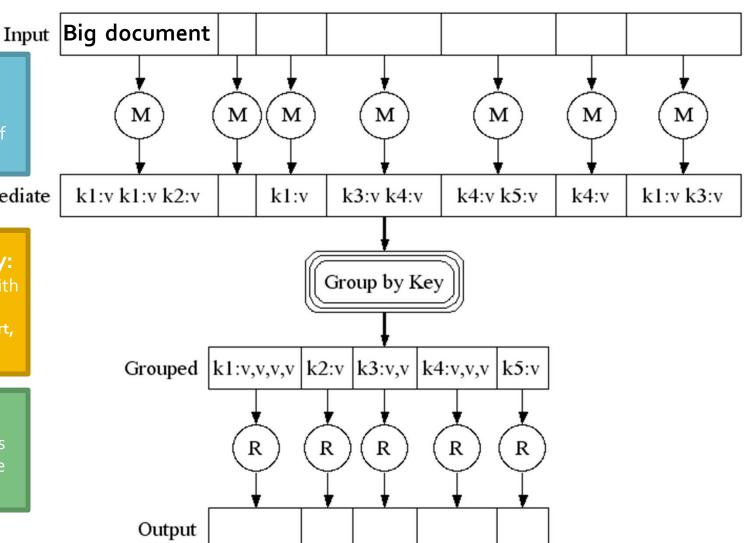
Intermediate

Group by key:

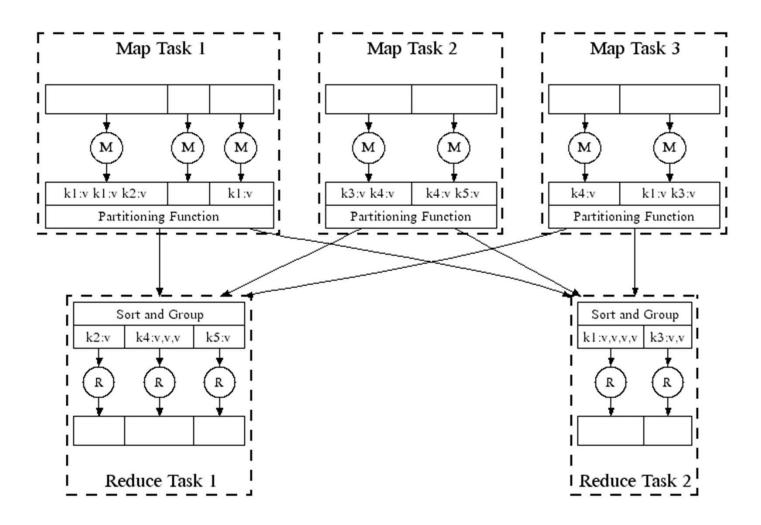
Collect all pairs with merge, Shuffle, Sort, Partition)

Reduce:

Collect all values belonging to the key and output



Map-Reduce: In Parallel



All phases are distributed with many tasks doing the work

Map-Reduce: Environment

Map-Reduce environment takes care of:

- Partitioning the input data
- Scheduling the program's execution across a set of machines
- Performing the group by key step
- Handling node failures
- Managing required inter-machine

communication

Data Flow

- Input and final output are stored on the distributed file system (DFS):
 - Scheduler tries to schedule map tasks "close" to physical storage location of input data
- Intermediate results are stored on local FS of Map and Reduce workers
 - Output is often input to another
 MapReduce task

Coordination: Master

Master node takes care of coordination:

- Task status: (idle, in-progress, completed)
- Idle tasks get scheduled as workers become available
- When a map task completes, it sends the master the location and sizes of its R intermediate files, one for each reducer
- Master pushes this info to reducers
- Master pings workers periodically to detect failures

Dealing with Failures

Map worker failure

- Map tasks completed or in-progress at worker are reset to idle.
- Idle tasks eventually rescheduled on other worker(s)

Reduce worker failure

- Only in-progress tasks are reset to idle
- Idle Reduce tasks restarted on other worker(s)

Master failure

MapReduce task is aborted and client is notified

How many Map and Reduce jobs?

- M map tasks, R reduce tasks
- Rule of thumb:

Make *M* much larger than the number of nodes in the cluster

One DFS chunk per map is common Improves dynamic load balancing and speeds up recovery from worker failures

Usually R is smaller than M

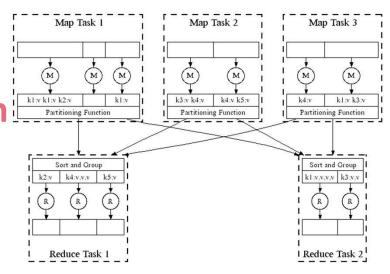
Because output is spread across R files

Map-Reduce

Refinements
Implementations

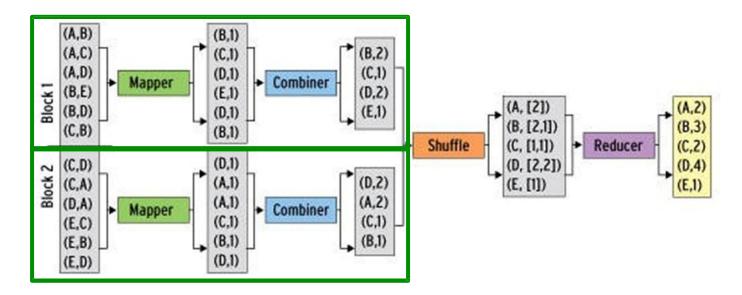
Refinement: Combiners (1)

- Often a Map task will produce many pairs of the form (k,v_1) , (k,v_2) , ... for the same key k
 - E.g., popular words in the word count example
- Can save network time by pre-aggregating values the mapper:
 - combine(k, list(v_1)) \rightarrow v_2
 - Combiner is usually same as the reduce function



Refinement: Combiners (2)

- Back to our word counting example:
 - Combiner combines the values of all keys of a single mapper (single node):



Much less data needs to be copied and shuffled!

Refinement: Combiners (3)

- Combiner trick works only if reduce function is commutative and associative.
- Sum

$$2 + (5 + 7) = (2 + 5) + 7$$

Average

Median

Refinement: Partition Function

- Want to control how keys get partitioned
 - The set of keys that go to a single reduce worker
- System uses a default partition function:
 - Hash (key) mod R
- Sometimes useful to override the hash function:
 - E.g., hash (hostname(URL)) mod R ensures URLs from a host end up in the same output file



Reading

- Jeffrey Dean and Sanjay Ghemawat: MapReduce:
 Simplified Data Processing on Large Clusters
 - http://labs.google.com/papers/mapreduce.html
- Sanjay Ghemawat, Howard Gobioff, and Shun-Tak Leung: The Google File System
 - http://labs.google.com/papers/gfs.html

Resources

- Hadoop Wiki
 - Introduction
 - http://wiki.apache.org/lucene-hadoop/
 - Getting Started
 - http://wiki.apache.org/lucene-hadoop/GettingStartedWithHadoop
 - Map/Reduce Overview
 - http://wiki.apache.org/lucene-hadoop/HadoopMapReduce
 - http://wiki.apache.org/lucene-hadoop/HadoopMapRedClasses
 - Eclipse Environment
 - http://wiki.apache.org/lucene-hadoop/EclipseEnvironment
- Javadoc
 - http://lucene.apache.org/hadoop/docs/api/

Resources

- Releases from Apache download mirrors
 - http://www.apache.org/dyn/closer.cgi/lucene/hadoop/
- Nightly builds of source
 - http://people.apache.org/dist/lucene/hadoop/ nightly/
- Source code from subversion
 - http://lucene.apache.org/hadoop/ version_control.html