



### Scalable Data Science

Lecture 18: Hadoop System

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# In the previous lectures:

#### Outline:

- What is Big Data and Hadoop?
- What is Map Reduce ?
- Example Map Reduce program.
- HDFS commands and internals.
- Implementation of Map-reduce program
  - Java
  - Other languages through streaming API.





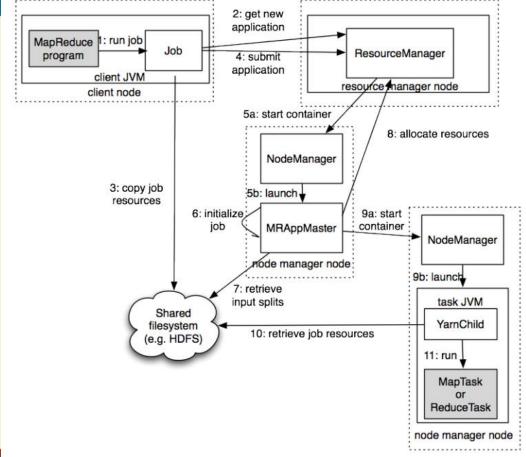
## In this Lecture:

- Outline:
  - Map-reduce implementation details
    - Coordination
    - Fault- tolerance
    - Pipelining
    - Refinements





# Hadoop(v2) MR job







# **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





# Fault tolerance

- ☐ Comes from scalability and cost effectiveness
- ☐ HDFS:
  - ☐ Replication
- ☐ Map Reduce
  - ☐ Restarting failed tasks: map and reduce
  - ☐ Writing map output to FS
  - ☐ Minimizes re-computation





## **Failures**

- ☐ Task failure
  - ☐ Task has failed report error to node manager, appmaster, client.
  - ☐ Task not responsive, JVM failure Node manager restarts tasks.
- Application Master failure
  - ☐ Application master sends heartbeats to resource manager.
  - ☐ If not received, the resource manager retrieves job history of the run tasks.
- Node manager failure





# **Dealing with Failures**

- Map worker failure
  - Map tasks completed or in-progress at worker are reset to idle
  - Reduce workers are notified when task is rescheduled on another worker.
- Reduce worker failure
  - Only in-progress tasks are reset to idle
  - Reduce task is restarted
- Master failure
  - MapReduce task is aborted and client is notified





# How many Map and Reduce jobs?

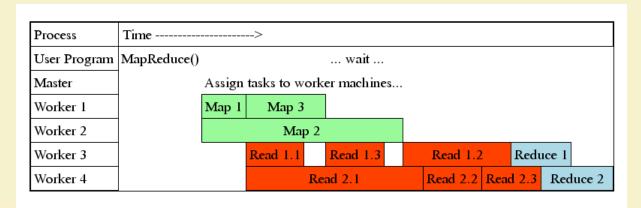
- M map tasks, R reduce tasks
- Rule of a thumb:
  - Make M much larger than the number of nodes in the cluster
  - One DFS chunk per map is common
  - Improves dynamic load balancing and speeds up recovery from worker failures
- Usually R is smaller than M
  - Because output is spread across R files





# Task Granularity & Pipelining

- Fine granularity tasks: map tasks >> machines
  - Minimizes time for fault recovery
  - Can do pipeline shuffling with map execution
  - Better dynamic load balancing





# Refinements: Backup Tasks

#### Problem

- Slow workers significantly lengthen the job completion time:
  - Other jobs on the machine
  - Bad disks
  - Weird things

#### Solution

- Near end of phase, spawn backup copies of tasks
  - Whichever one finishes first "wins"

#### Effect

Dramatically shortens job completion time





### **Refinement: Combiners**

• Often a Map task will produce many pairs of the form  $(k, v_1)$ ,  $(k, v_2)$ , ... for the same key k

Map Task 1

Partitioning Function

- E.g., popular words in the word count example
- Can save network time by pre-aggregating values in the mapper:
  - combine(k, list( $v_1$ ))  $\rightarrow v_2$
  - Combiner is usually same as the reduce function
- Works only if reduce function is commutative and associative





Map Task 3

Partitioning Function

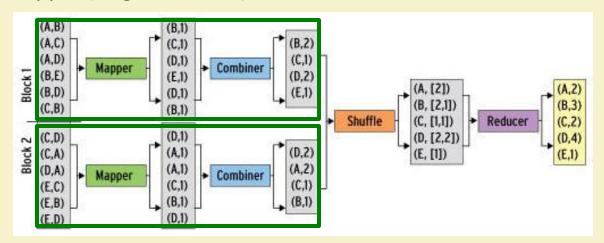
Reduce Task 2

Map Task 2

Partitioning Function

## **Refinement: Combiners**

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



Much less data needs to be copied and shuffled!





#### Refinement: Partition Function

- Want to control how keys get partitioned
  - Inputs to map tasks are created by contiguous splits of input file
  - Reduce needs to ensure that records with the same intermediate key end up at the same 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





## Conclusion:

- We have seen:
  - The structure of HDFS.
  - The shell commands.
  - The architecture of HDFS system.
  - Internal functioning of HDFS.





# References:

• Jure Leskovec, Anand Rajaraman, Jeff Ullman. **Mining of Massive Datasets.** 2<sup>nd</sup> edition. - Cambridge University Press. <a href="http://www.mmds.org/">http://www.mmds.org/</a>

• Tom White. **Hadoop: The definitive Guide.** Oreilly Press.





# Thank You!!



