

## Cloud Computing MapReduce in Heterogeneous Environments

#### Contents

- Looking at MapReduce performance in heterogeneous clusters
- Material is from the paper:

"Improving MapReduce Performance in Heterogeneous Environments", By Matei Zaharia, Andy Konwinski, Anthony D. Joseph, Randy Katz and Ion Stoica, published in Usenix OSDI conference, 2008

and their presentation at OSDI

#### **Motivation: MapReduce is becoming popular**

- Open-source implementation, Hadoop, used by Yahoo!, Facebook, Last.fm, ...
- Scale: 20 PB/day at Google, O(10,000) nodes at Yahoo, 3000 jobs/day at Facebook

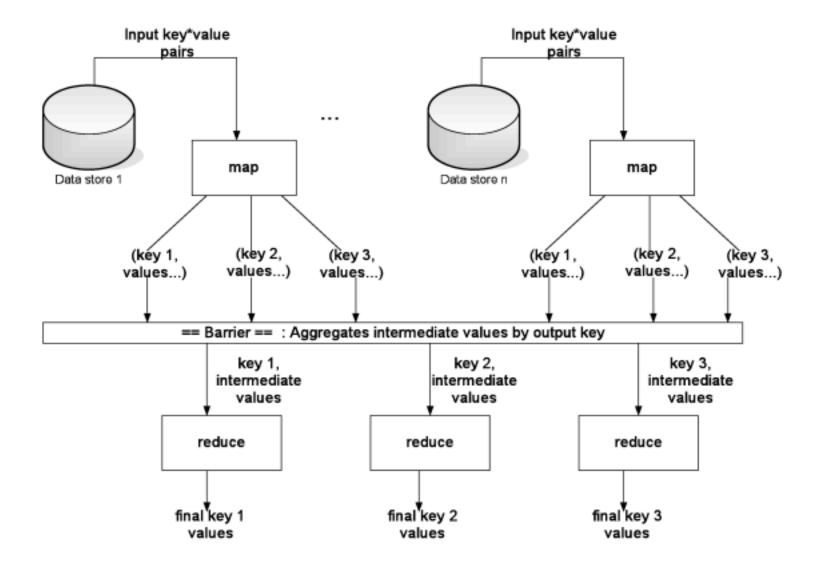
## **Stragglers in MapReduce**

- Straggler is a node that performs poorly or not performing at all.
- Original MapReduce mitigation approach was:
  - To run a speculative copy (called a backup task)
  - Whichever copy or original would finish first would be included
- Without speculative execution, a job would be slow as the slowest sub-task
- Google notes that speculative execution can improve job response times by 44%
- Is this approach good enough for modern clusters?

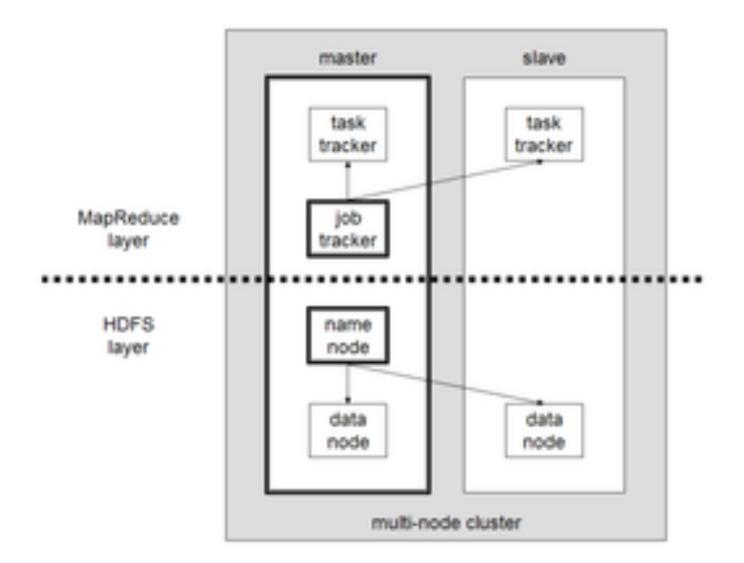
#### **Modern Clusters: Heterogeneity is the norm**

- Cloud computing providers like Amazon's Elastic Compute Cloud (EC2) provide cheap on-demand computing:
  - Price: 2 cents / VM / hour
  - Scale: thousands of VMs
  - Caveat: less control of performance
- Main challenge for Hadoop on EC2 is performance heterogeneity, which breaks task scheduler assumptions
- This lecture/paper is on a new LATE scheduler that can cut response time in half

#### **MapReduce Revised**



#### **MapReduce Implementation, Hadoop**



# **Scheduling in MapReduce**

- When a node has an empty slot, Hadoop chooses one from the three categories in the following priority:
  - 1. A failed task is given higher priority
  - 2. Unscheduled tasks. For maps, tasks with local data to the node are chosen first.
  - 3. Looks to run a speculative task.

## **Deciding on Speculative Tasks**

- Which task to execute speculatively?
- Hadoop monitors tasks progress using a progress score: a number from 0, ..., 1
- For mappers: the score is the fraction of input data read
- For reducers: the execution is divided into three equal phases, 1/3 of the score each:
  - Copy phase: percent of maps that output has been copied from
  - Sort phase: map outputs are sorted by key: percent of data merged
  - Reduce phase: percent of data passed through the reduce function
- Example: a task halfway through the copy phase has progress score = 1/2\*1/3 = 1/6.

• Example: a task halfway through the reduce phase has progress score = 1/3 + 1/3 + 1/2 \* 1/3 = 5/6

## **Deciding on Speculative Tasks** (con't)

- Hadoop looks at the average progress of each category of maps and reduces and defines a **threshold**:
- When a task's progress is less than the average for its category minus 0.2, and the task has run at least one minute, it is marked as a straggler:

threshold = avgProgress – 0.2

- All tasks with progress score < threshold are stragglers</p>
- Ties are broken by data locality
- This approach works reasonably well in homogeneous clusters

## **Scheduler's Assumptions**

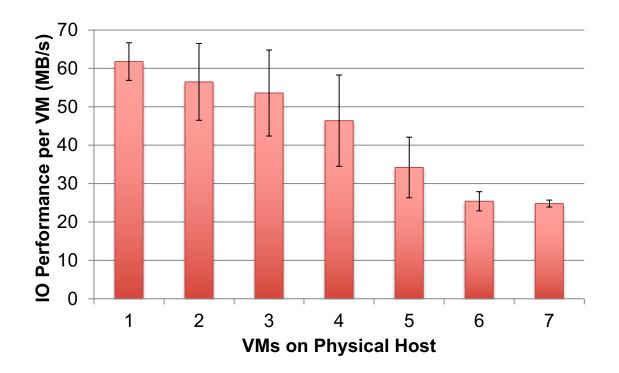
- 1. Nodes can perform work at roughly the same rate
- 2. Tasks progress at constant rate all the time
- 3. There is no cost to starting a speculative task
- 4. A task's progress is roughly equal to the fraction of its total work
- 5. Tasks tend to finish in waves, so a task with a low progress score is likely a slow task
- 6. Different task of the same category (maps or reduces) take roughly the same amount of work

## **Revising Scheduler's Assumptions**

- 1. Nodes can perform work at roughly the same rate
- 2. Tasks progress at constant rate all the time
- (1) In heterogeneous clusters some nodes are slower (older) than others
- (2) Virtualized clusters "suffer" from co-location interference

#### **Heterogeneity in Virtualized Environments**

- VM technology isolates CPU and memory, but disk and network are shared
  - Full bandwidth when no contention
  - Equal shares when there is contention
- Timed a dd command that wrote 5000 MB of zeroes from /dev/zero to a file in parallel on 871 virtual machines in EC2's production cluster.
- **2.5x** performance difference

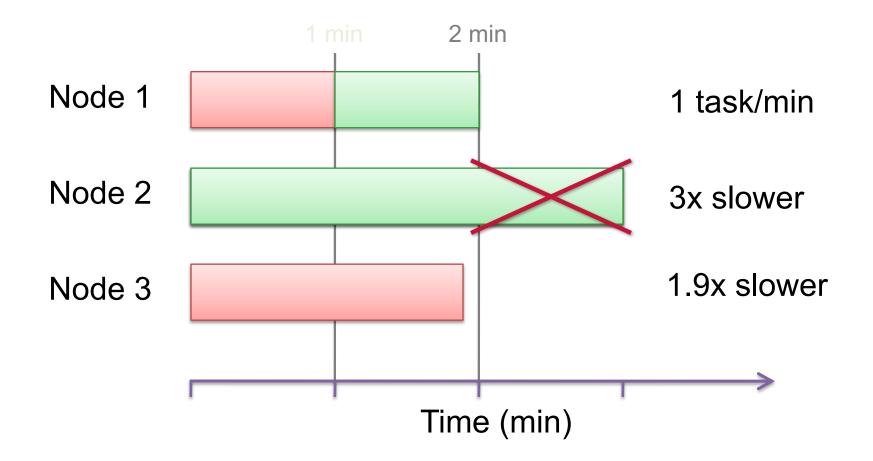


## **Revising Scheduler's Assumptions**

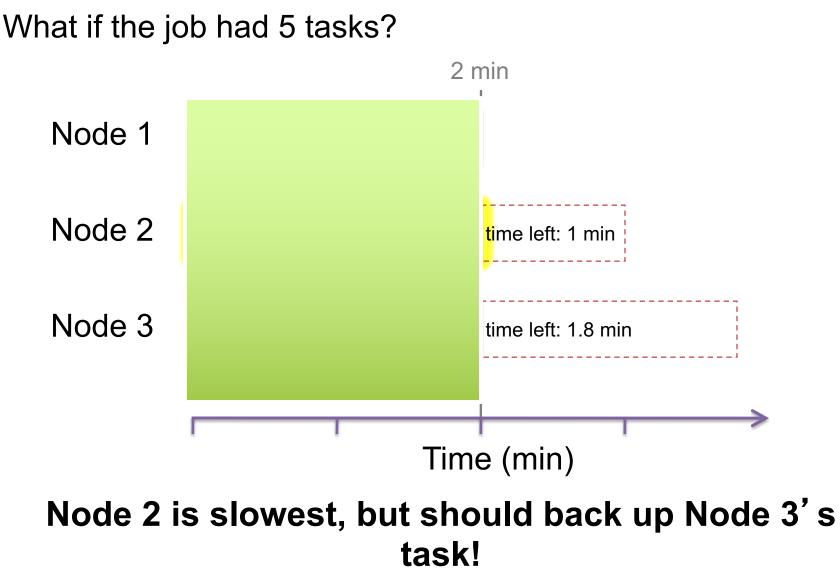
- 3. There is no cost to starting a speculative task
- 4. A task's progress is roughly equal to the fraction of its total work
- 5. Tasks tend to finish in waves, so a task with a low progress score is likely a slow task
- (3) Too many speculative tasks can take away resources from other running tasks
- (4) The copy phase of reducers is the slowest part, because it involves all-pairs communications. But this phase counts for 1/3 of the total reduce work.
- (5) Tasks from different generations will be executed concurrently. So newer faster tasks are considered with older show tasks, avgProgress changes a lot.

- Instead of using progress score values, compute progress rates, and back up tasks that are "far enough" below the mean
- Problem: can still select the wrong tasks

#### **Progress Rate Example**



#### **Progress Rate Example**



#### **Our Scheduler: LATE**

- Insight: back up the task with the largest estimated finish time
  - "Longest Approximate Time to End"  $\rightarrow$  LATE
  - Look forward instead of looking backward
- Sanity thresholds:
  - Cap number of backup tasks
  - Launch backups on fast nodes
  - Only back up tasks that are sufficiently slow

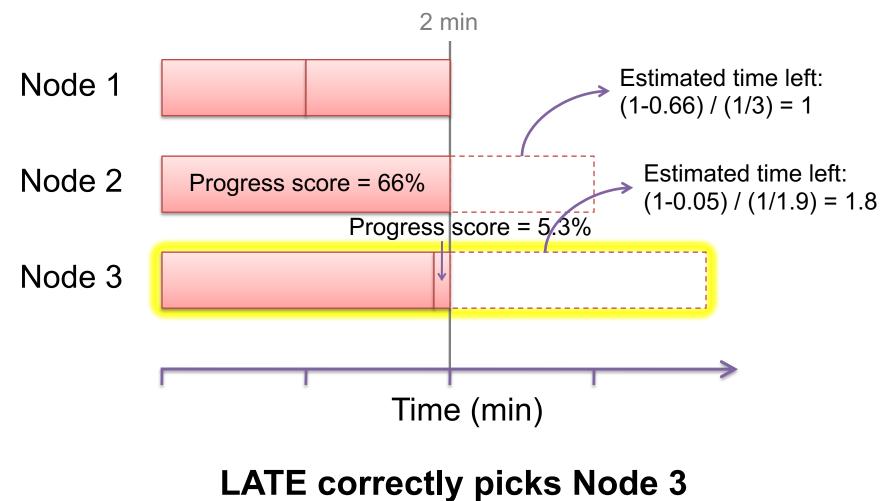
#### **LATE Details**

Estimating finish times:

 $progress rate = \frac{progress score}{execution time}$   $estimated time left = \frac{1 - progress score}{progress rate}$ 

- If a task slot becomes available and there are less than SpeculativeCap tasks running, then:
  - 1. Ignore the request if the node's total progress is below *SlowNodeThreshold* (=25<sup>th</sup> percentile)
  - 2. Rank currently running, non-speculatively executed tasks by estimated time left
  - 3. Launch a copy of the highest-ranked task with progress rate below *SlowTaskThreshold* (=25<sup>th</sup> percentile)
- Threshold values:
  - 10% cap on backups, 25<sup>th</sup> percentiles for slow node/task
  - Validated by sensitivity analysis

#### **LATE Example**



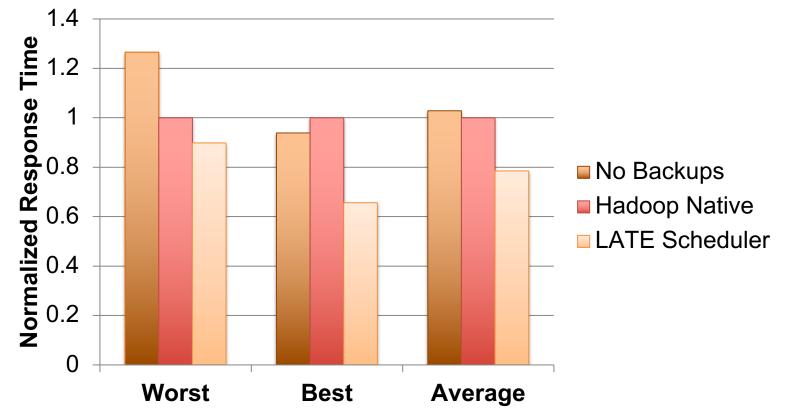
## **Evaluation**

#### Environments:

- EC2 (3 job types, 200-250 nodes)
- Small local testbed
- Self-contention through VM placement
- Stragglers through background processes

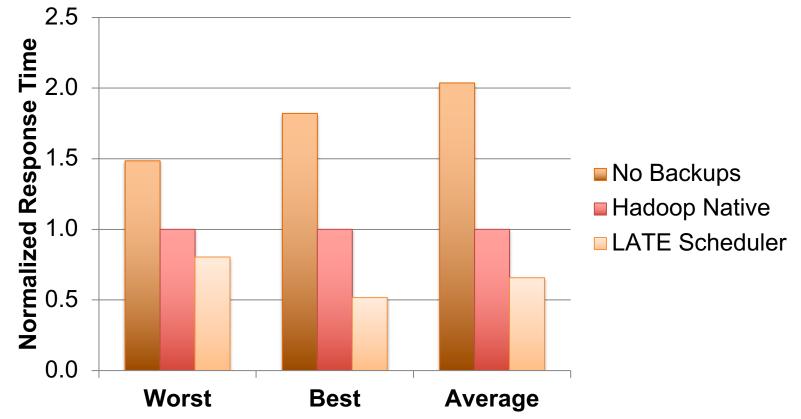
#### EC2 Sort without Stragglers (Sec 5.2.1)

- 106 machines , 7-8 VMs per machine  $\rightarrow$  total of 243 VMs
- 128 MB data per host, 30 GB in total
- 486 map tasks and 437 reduce tasks
- average 27% speedup over native, 31% over no backups



# EC2 Sort with Stragglers (Sec 5.2.2)

- 8 VMs are manually slowed down out of 100 VMs in total
- running background of CPU- and disk-intensive jobs
- average 58% speedup over native, 220% over no backups
- 93% max speedup over native



## Conclusion

- Heterogeneity is a challenge for parallel apps, and is growing more important
- Lessons:
  - Back up tasks which hurt response time most
- 2x improvement using simple algorithm

- MapReduce is a very powerful and expressive model
- Performance depends a lot on implementation details

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