CS744: Big Data Systems Notes

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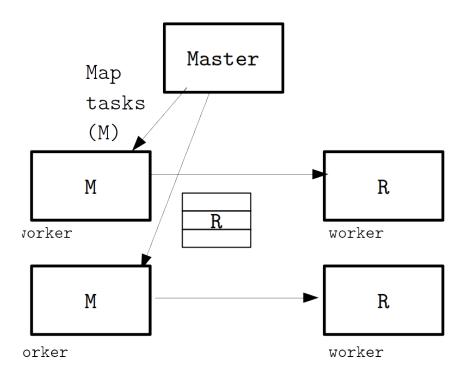
February 18, 2019

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1 MapReduce (2.4.19)

- Programming model
- Execution
- Runtime issues
- M-R library handles execution and run-time issues
 - Transparent to programmers



1.1 Operators

- 1. Map
 - Input = $(\text{key}, \text{value}) \rightarrow (\text{key}, < \text{v}>)$

2. Reduce

- Operates share a key
- (key,value) is sorted and values passed to reducer

1.2 Failures and Slowdowns

• Handled by the master

1.2.1 Possible failures

- 1. Map / Reduce
 - Worker fails, some maps and some reduces completed
 - Reduce data is already written to HDFS, doesn't need to be recomputed
 - Maps must be re-executed to recover intermediate data, since it hasn't been written to HDFS

2 Spark (2.6.19)

• Programming model

2.1 RDDs

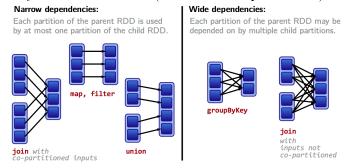
- Partitioned collection of records
- SQL, D-Streams, Graphx
- Intermediate data stored in memory
- Low overhead fault tolerance achieved through lineage

2.2 Benefits

- 1. Speed up iterative computations
- 2. Load datasets into memory
 - can't be done in MapReduce
- 3. Higher level programs

Verbatim RDD -> transformations -> action

- Persist (deserialized, serialized, on-disk)
 - RDDs only exist logically unless persist is called
 - * Only then materialized (unless wide dependencies)



- REL (reliable flag): checkpoint to disk or other memory locations
- Partitioning
- Lazy computation

2.3 Example: PageRank

- General process:
 - 1. Gather
 - 2. Applies
 - 3. Scatter

$3 \operatorname{Spark} + \operatorname{Job Manager} (2.8.19)$

3.1 Job Manager

- Scheduling: resources, ready, location
 - 1. How important is the task? (critical path) -> modeling, use some tasks to estimate others

3.2 Stragglers

- 1. Rate of progress -> progress report
 - Compare report with model
 - 50% slower -> this task is a straggler
 - Detect and react to stragglers early
 - No early detection of stragglers in MapReduce
- 2. Worthwhile to react to stragglers?
 - Does cloning lower overall resource use?

4 Cluster Schedulers – Mechanism (2.11.19)

4.1 Mechanisms

- YARN: one-level resource allocator
- Mesos: two-level pessimistic scheduler
- Omega/Borg: optimistic scheduler
- Fair allocation delay scheduling

4.2 YARN

- Statistical-multiplexing
 - Work-conserving allocation discipline
 - * No resources go to waste
- Node Manager (NM): local available resources
- Resource Manager (RM): global state
- Application Master (AM): requesting for resources
- De-coupled from programming model
 - Gang scheduling, Message Passing Interface (MPI)
 - * Revoking vs. admission control
- Late binding (kill-and-restart, e.g.)

4.3 Mesos

- Two-level scheduler
- Resource offers
 - Slaves report what resources are available
 - Mesos offers these resources to individual tasks
 - * Frameworks can accept or deny offers (maybe they are not data-local)
- May kill tasks that run for too long to allow better sharing of resources

4.3.1 Weaknesses

• Applications cannot request specific resources, must wait until they get offered what they want

5 Omega Scheduler (2.13.19)

- HFS (delay scheduling)
- DRF: instantaneous fairness
 - 1. all-or-nothing
 - 2. priority

5.1 Dominant Resource Fairness

- 1. Slots -> rigid
- 2. Tasks are uniform
- 3. Equal number of slots is fair

$6 \quad (2.18.19)$

- Resource variabilities (Clarenet, QOOP) | Batch analytics
- WAN bandwidth -> Geo distributed analytics
- Compute resources:

- Spot market
- Small clusters

6.1 Geo Distributed Analytics

6.1.1 Constraints

- Bandwidth varies over links (logical link)
- Control Plane
 - Scheduler
 - Job Manager
 - Parameter server
 - Partitioning/replication/coordination of the above
- Latency
 - Staleness
 - Heartbeats
- Legal & Privacy Issues
 - Multi-party computation

6.1.2 Clarinet

- Mutually-trusting data centers
- Batch computation multiple queries
- Ignores legal & privacy issues
- Ignores partitioning scheduler or job manager
- 1. Big ideas
 - (a) Query plans in a network-aware manner

```
Query => Query Optimizer (n/w aware) => Scheduler =>
```

- Placement of tasks
- Scheduling, or "when"