Lecture 3 – Hadoop Technical Introduction

CSE 490H

Announcements

- My office hours: M 2:30—3:30 in CSE 212
- Cluster is operational; instructions in assignment 1 heavily rewritten
- Eclipse plugin is "deprecated"
- Students who already created accounts: let me know if you have trouble

Breaking news!

- Hadoop tested on 4,000 node cluster
 32K cores (8 / node)
 16 PB raw storage (4 x 1 TB disk / node) (about 5 PB usable storage)
- http://developer.yahoo.com/blogs/hadoop/2008/09/ scaling_hadoop_to_4000_nodes_a.html

You Say, "tomato..."

Google calls it:	Hadoop equivalent:
MapReduce	Hadoop
GFS	HDFS
Bigtable	HBase
Chubby	Zookeeper

Some MapReduce Terminology

- Job A "full program" an execution of a Mapper and Reducer across a data set
- Task An execution of a Mapper or a Reducer on a slice of data

□ a.k.a. Task-In-Progress (TIP)

Task Attempt – A particular instance of an attempt to execute a task on a machine

Terminology Example

- Running "Word Count" across 20 files is one job
- 20 files to be mapped imply 20 map tasks
 + some number of reduce tasks
- At least 20 map task attempts will be performed... more if a machine crashes, etc.

Task Attempts

- A particular task will be attempted at least once, possibly more times if it crashes
 - If the same input causes crashes over and over, that input will eventually be abandoned
- Multiple attempts at one task may occur in parallel with speculative execution turned on
 - Task ID from TaskInProgress is not a unique identifier; don't use it that way

MapReduce: High Level



Node-to-Node Communication

- Hadoop uses its own RPC protocol
- All communication begins in slave nodes
 Prevents circular-wait deadlock
 Slaves periodically poll for "status" message
- Classes must provide explicit serialization

Nodes, Trackers, Tasks

- Master node runs JobTracker instance, which accepts Job requests from clients
- TaskTracker instances run on slave nodes
- TaskTracker forks separate Java process for task instances

Job Distribution

- MapReduce programs are contained in a Java "jar" file + an XML file containing serialized program configuration options
- Running a MapReduce job places these files into the HDFS and notifies TaskTrackers where to retrieve the relevant program code
- Where's the data distribution?

Data Distribution

- Implicit in design of MapReduce!
 - All mappers are equivalent; so map whatever data is local to a particular node in HDFS
- If lots of data does happen to pile up on the same node, nearby nodes will map instead
 - □ Data transfer is handled implicitly by HDFS

Configuring With JobConf

- MR Programs have many configurable options
- JobConf objects hold (key, value) components mapping String → 'a
 - \Box e.g., "mapred.map.tasks" \rightarrow 20
 - JobConf is serialized and distributed before running the job
- Objects implementing *JobConfigurable* can retrieve elements from a JobConf

What Happens In MapReduce? Depth First

Job Launch Process: Client

- Client program creates a JobConf
 Identify classes implementing Mapper and
 - *Reducer* interfaces
 - JobConf.setMapperClass(), setReducerClass()
 - Specify inputs, outputs
 - FileInputFormat.addInputPath(),
 - FileOutputFormat.setOutputPath()
 - Optionally, other options too:
 - JobConf.setNumReduceTasks(), JobConf.setOutputFormat()...

Job Launch Process: JobClient

- Pass JobConf to JobClient.runJob() or submitJob()
 - □runJob() blocks, submitJob() does not
- JobClient:
 - Determines proper division of input into InputSplits
 - □ Sends job data to master *JobTracker* server

Job Launch Process: JobTracker

JobTracker:

- Inserts jar and JobConf (serialized to XML) in shared location
- □ Posts a *JobInProgress* to its run queue

Job Launch Process: TaskTracker

- TaskTrackers running on slave nodes periodically query JobTracker for work
- Retrieve job-specific jar and config
- Launch task in separate instance of Java
 main() is provided by Hadoop

Job Launch Process: Task

- TaskTracker.Child.main():
 Sets up the child *TaskInProgress* attempt
 Reads XML configuration
 Connects back to necessary MapReduce components via RPC
 - □ Uses *TaskRunner* to launch user process

Job Launch Process: TaskRunner

- TaskRunner, MapTaskRunner, MapRunner work in a daisy-chain to launch your Mapper
 - Task knows ahead of time which *InputSplits* it should be mapping
 - Calls Mapper once for each record retrieved from the InputSplit
- Running the *Reducer* is much the same

Creating the Mapper

- You provide the instance of Mapper
 Should extend MapReduceBase
- One instance of your Mapper is initialized by the MapTaskRunner for a TaskInProgress
 - Exists in separate process from all other instances of Mapper no data sharing!

Mapper

void map(K1 key, V1 value, OutputCollector<K2, V2> output, Reporter reporter)

K types implement *WritableComparable V* types implement *Writable*

What is Writable?

- Hadoop defines its own "box" classes for strings (*Text*), integers (*IntWritable*), etc.
- All values are instances of Writable
- All keys are instances of WritableComparable

Getting Data To The Mapper



Reading Data

- Data sets are specified by *InputFormats* Defines input data (e.g., a directory)
 - Identifies partitions of the data that form an InputSplit
 - Factory for *RecordReader* objects to extract (k, v) records from the input source

FileInputFormat and Friends

- TextInputFormat Treats each '\n'terminated line of a file as a value
- KeyValueTextInputFormat Maps '\n'terminated text lines of "k SEP v"
- SequenceFileInputFormat Binary file of (k, v) pairs with some add'l metadata
- SequenceFileAsTextInputFormat Same, but maps (k.toString(), v.toString())

Filtering File Inputs

- FileInputFormat will read all files out of a specified directory and send them to the mapper
- Delegates filtering this file list to a method subclasses may override
 - e.g., Create your own "xyzFileInputFormat" to read *.xyz from directory list

Record Readers

- Each InputFormat provides its own RecordReader implementation
 Provides (unused?) capability multiplexing
- LineRecordReader Reads a line from a text file
- KeyValueRecordReader Used by KeyValueTextInputFormat

Input Split Size

- FileInputFormat will divide large files into chunks
 - Exact size controlled by mapred.min.split.size
- RecordReaders receive file, offset, and length of chunk
- Custom InputFormat implementations may override split size – e.g., "NeverChunkFile"

Sending Data To Reducers

- Map function receives OutputCollector object
 - OutputCollector.collect() takes (k, v) elements
- Any (WritableComparable, Writable) can be used
- By default, mapper output type assumed to be same as reducer output type

WritableComparator

Compares WritableComparable data

 Will call WritableComparable.compare()
 Can provide fast path for serialized data

 JobConf.setOutputValueGroupingComparator()

Sending Data To The Client

- Reporter object sent to Mapper allows simple asynchronous feedback

 incrCounter(Enum key, long amount)
 setStatus(String msg)
- Allows self-identification of input
 InputSplit getInputSplit()

Partition And Shuffle



Partitioner

- int getPartition(key, val, numPartitions)
 Outputs the partition number for a given key
 One partition == values sent to one Reduce task
- HashPartitioner used by default
 Uses key.hashCode() to return partition num
 JobConf sets Partitioner implementation

Reduction

reduce(K2 key,

Iterator<V2> values, OutputCollector<K3, V3> output, Reporter reporter)

- Keys & values sent to one partition all go to the same reduce task
- Calls are sorted by key "earlier" keys are reduced and output before "later" keys

Finally: Writing The Output



OutputFormat

- Analogous to InputFormat
- TextOutputFormat Writes "key val\n" strings to output file
- SequenceFileOutputFormat Uses a binary format to pack (k, v) pairs
- NullOutputFormat Discards output

Questions?