Counting at Scale APAM E4990 Modeling Social Data

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Previously

Claim:

Solving the counting problem at scale enables you to investigate many interesting questions in the social sciences

Learning to count

Last week:

Counting at small/medium scales on a single machine

Learning to count

Last week:

Counting at small/medium scales on a single machine

This week:

Counting at large scales in parallel

What?



What?

"... to create building blocks for programmers who just happen to have lots of data to store, lots of data to analyze, or lots of machines to coordinate, and who don't have the time, the skill, or the inclination to become distributed systems experts to build the infrastructure to handle it."

-Tom White Hadoop: The Definitive Guide

What?

Hadoop contains many subprojects:

- Hadoop Common: The common utilities that support the other Hadoop subprojects.
- <u>Chukwa</u>: A data collection system for managing large distributed systems.
- <u>HBase</u>: A scalable, distributed database that supports structured data storage for large tables.
- HDFS: A distributed file system that provides high throughput access to application data.
- <u>Hive</u>: A data warehouse infrastructure that provides data summarization and ad hoc querying.
- MapReduce: A software framework for distributed processing of large data sets on compute clusters.
- · Pig: A high-level data-flow language and execution framework for parallel computation.
- ZooKeeper: A high-performance coordination service for distributed applications.

We'll focus on distributed computation with MapReduce.

An overly brief history

pre-2004

Doug Cutting and Mike Cafarella develop open source projects for web-scale indexing, crawling, and search



2004

Dean and Ghemawat publish MapReduce programming model, used internally at Google

MapReduce: Simplified Data Processing on Large Clusters

Jeffrey Dean and Sanjay Ghemawat

jeff@google.com, sanjay@google.com

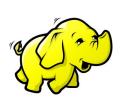
Google, Inc.

Abstract

MapReduce is a programming model and an associated implementation for processing and generating large data sets. Users specify a map function that processes a key/value pair to generate a set of intermediate key/value pairs, and a reduce function that merges all intermediate values associated with the same intermediate key and year load tasks are expressible in this model, as shown in the paper.

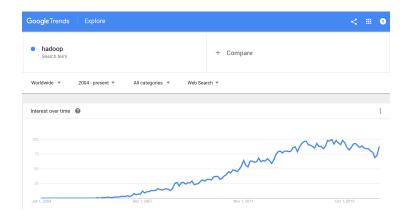
given day, etc. Most such computations are conceptually straightforward. However, the input data is usually large and the computations have to be distributed across hundreds or thousands of machines in order to finish in a reasonable amount of time. The issues of how to parallelize the computation, distribute the data, and handle failures conspire to obscure the original simple computation with large amounts of complex code to deal with these issues.

2006 Hadoop becomes official Apache project, Cutting joins Yahoo!, Yahoo adopts Hadoop









Where?



http://wiki.apache.org/hadoop/PoweredBy

Why yet another solution?

(I already use too many languages/environments)

Why a distributed solution?

(My desktop has TBs of storage and GBs of memory)



Roughly how long to read 1TB from a commodity hard disk?



Roughly how long to read 1TB from a commodity hard disk?

$$\frac{1}{2}\frac{\mathsf{Gb}}{\mathsf{sec}} \times \frac{1}{8}\frac{\mathsf{B}}{\mathsf{b}} \times 3600\frac{\mathsf{sec}}{\mathsf{hr}} \approx 225\frac{\mathsf{GB}}{\mathsf{hr}}$$

Roughly how long to read 1TB from a commodity hard disk?¹

 \approx 4hrs

MAY 11, 2009

Hadoop Sorts a Petabyte in 16.25 Hours and a Terabyte in 62 Seconds

We used **Apache Hadoop** to compete in **Jim Gray's Sort** benchmark. Jim's Gray's sort benchmark consists of a set of many related benchmarks, each with their own rules. All of the sort benchmarks measure the time to sort different numbers of 100 byte records. The first 10 bytes of each record is the key and the rest is the value. The **minute sort** must finish end to end in less than a minute. The **Gray sort** must sort more than 100 terabytes and must run for at least an hour. The best times we observed were:

Bytes	Nodes	Maps	Reduces	Replication	Time
500,000,000,000	1406	8000	2600	1	59 seconds
1,000,000,000,000	1460	8000	2700	1	62 seconds
100,000,000,000,000	3452	190,000	10,000	2	173 minutes
1,000,000,000,000,000	3658	80,000	20,000	2	975 minutes

http://bit.ly/petabytesort

Typical scenario

Store, parse, and analyze high-volume server logs,

[16/May/2010;07;28:49 -0400] "GET /autonomous_css/style.css HTTP/1.1" 200 2806 "http://www.jakehofman.com/" "Mozilla/4.0 (compatible; MSIE 8.0; Windows HT 6.1; WWW64; Trident/4.0; GTB6.4; SLCC2; NET CLR 2.0,50727; NET CLR 3.5,30729; NET CLR 3.0,30729; Media Center PC 6.0; OfficeliveCornec tor.1.4; OfficeliveEdtch.1.3)

[18/May/200/07/2849 -0400] "GET /cleanlooks/test3.gif HTTP/1.1" 404 339 "http://www.jakehofhwn.com/" "hozilla/4,0 (compatible: MSE 8,0: Windows ME 6.1: W0064; Trident/4,0; GTB6.4; SLCC2; NET CLR 2,0.50727; .NET CLR 3,0.30729; Media Center PC 6.0: OfficeLiveConnector_1,4: OfficeLiveFatch_1.3)"

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.4: OfficeLivePatch.1.3"

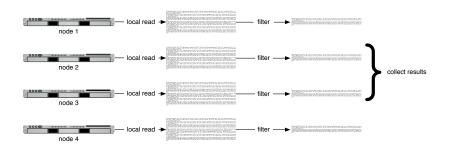
e.g. how many search queries match "icwsm"?

MapReduce: 30k ft

Break large problem into smaller parts, solve in parallel, combine results

Typical scenario

"Embarassingly parallel" (or nearly so)



Typical scenario++

How many search queries match "icwsm", grouped by month?

[16/May/2010:07:28:49 -0400] "GET /autonomous_css/style.css HTTP/1.1" 200 2006 "http://www.jakehofman.com/" "Mozilla/4.0 (compatible; MSIE 8.0; Windows NT 6.1; WOM64; Trident/4.0; GTB6.4; SLCC2; "NET CLR 2.0,50727; "NET CLR 3.5,30729; "NET CLR 3.0,30729; Media Center PC 6.0; OfficeLiveConnec tor.1.4; Official.vePatch.1.3;

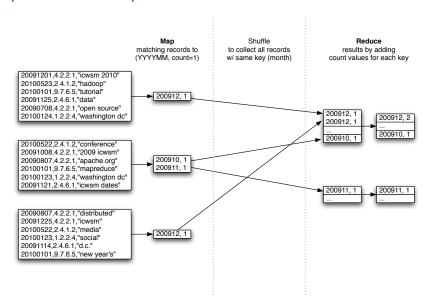
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MapReduce: example



Programmer specifies map and reduce functions

Map: tranforms input record to intermediate (key, value) pair

```
def mapper(record):
    # input: a single record

    # parse / transform / filter record
    ...

# output: intermediate key(s) and value(s)
    output( (key, value) )
```

Shuffle: collects all intermediate records by key

Record assigned to reducers by hash(key) % num_reducers

Reducers perform a merge sort to collect records with same key

Reduce: transforms all records for given key to final output

Distributed read, shuffle, and write are transparent to programmer

MapReduce: principles

- Move code to data (local computation)
- Allow programs to scale transparently w.r.t size of input
- Abstract away fault tolerance, synchronization, etc.

MapReduce: strengths

- Batch, offline jobs
- Write-once, read-many across full data set
- Usually, though not always, simple computations
- I/O bound by disk/network bandwidth

!MapReduce

What it's not:

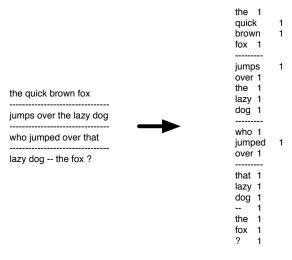
- High-performance parallel computing, e.g. MPI
- Low-latency random access relational database
- Always the right solution

the quick brown fox jumps over the lazy dog who jumped over that lazy dog -- the fox ?

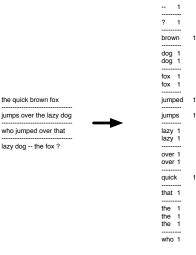


dog 2
-- 1
the 3
brown
fox 2
jumped
lazy 2
jumps
over 2
quick
that 1
who 1
? 1

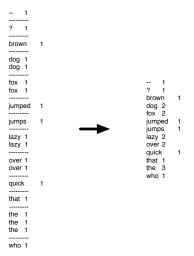
Map: for each line, output each word and count (of 1)

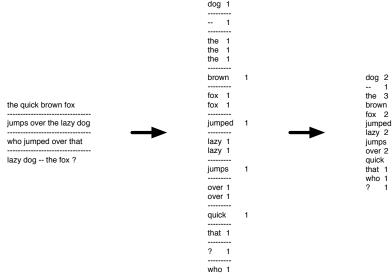


Shuffle: collect all records for each word



Reduce: add counts for each word





dog 1

3

1

1

1

WordCount.java

```
import java.io.IOExceptions
import java.util.*;
import org.apache.hadoop.fs.Pathr
import org.apache.hadoop.conf.*;
import org.apache.hadoop.io.*;
 public static class Map extends MapReduceBase implements Mapper<LongWritable, Text, Text, IntWritable> (
  private final static IntWritable one = new IntWritable(1);
  private Text word - new Text();
 public static class Reduce extends MapReduceBase implements Reducer<Text, IntNritable, Text, IntNritable> {
  public void reduce(Text key, Iterator(IntWritable) values, OutputCollector(Text, IntWritable) output, Reporter reporter) throws IOException (
   while (values.hasNext()) {
     sum += values.next().get();
    output.collect(key, new IntWritable(sum));
  JobConf conf = new JobConf (WordCount.class);
   conf.setJobName("wordcount");
   conf.setMapperClass(Map.class);
   conf.setReducerClass(Reduce.class);
  FileOutputFormat.setOutputPath(conf, new Path(args[1]));
```

Hadoop streaming

Hadoop streaming is a utility that comes with the Hadoop distribution. The utility allows you to create and run map/reduce jobs with any executable or script as the mapper and/or the reducer. For example:

```
$HADOOP_HOME/bin/hadoop jar $HADOOP_HOME/hadoop-streaming.jar \
-input myInputDir \
-output myOutputDir \
-mapper /bin/cat \
-reducer /bin/wc
```

Hadoop streaming

MapReduce for *nix geeks²:

```
# cat data | map | sort | reduce
```

Where the sort is a hack to approximate a distributed group-by:

- Mapper reads input data from stdin
- Mapper writes output to stdout
- Reducer receives input, grouped by key, on stdin
- Reducer writes output to stdout

wordcount.sh

Locally:

```
# cat data | tr " " "\n" | sort | uniq -c
```

wordcount.sh

Locally:

```
# cat data | tr " " \n | sort | uniq -c
```

 \Downarrow

Distributed:

```
$HADOOP_HOME/bin/hadoop jar $HADOOP_HOME/hadoop-streaming.jar \
-input README.txt \
-output wordcount \
-mapper 'tr " " "\n" \
-reducer 'uniq -c'
```

Transparent scaling

Use the same code on MBs locally or TBs across thousands of machines.

wordcount.py

```
from hstream import HStream
import sys
import re
from collections import defaultdict
class WordCount(HStream):
    def mapper(self, record):
        for word in " ".join(record).split():
            self.write_output((word, 1))
    def reducer(self, key, records):
        total = 0
        for record in records:
            word, count = record
            total += int(count)
        self.write output( (word, total) )
if __name__ == '__main__':
    WordCount()
```

Higher level abstractions

Higher level languages like Pig or Hive provide robust implementations for many common MapReduce operations

(e.g., filter, sort, join, group by, etc.)

They also allow for user-defined map and reduce functions

Higher level abstractions

Pig looks like SQL, but is more procedural

```
input lines = LOAD '/tmp/my-copy-of-all-pages-on-internet' AS (line:chararray);
-- Extract words from each line and put them into a pig bag
-- datatype, then flatten the bag to get one word on each row
words = FOREACH input lines GENERATE FLATTEN(TOKENIZE(line)) AS word;
-- filter out any words that are just white spaces
filtered words = FILTER words BY word MATCHES '\\w+';
-- create a group for each word
word groups = GROUP filtered words BY word;
-- count the entries in each group
word count = FOREACH word groups GENERATE COUNT(filtered words) AS count, group AS word;
-- order the records by count
ordered word count = ORDER word count BY count DESC;
STORE ordered word count INTO '/tmp/number-of-words-on-internet':
```

pig.apache.org

Higher level abstractions

Hive is closer to SQL, with nested declarative statements

```
1 DROP TABLE IF EXISTS docs;
2 CREATE TABLE docs (line STRING);
3 LOAD DATA INPATH 'input_file' OVERWRITE INTO TABLE docs;
4 CREATE TABLE word_counts AS
5 SELECT word, count(1) AS count FROM
6 (SELECT explode(split(line, '\s')) AS word FROM docs) temp
7 GROUP BY word
8 ORDER BY word;
```

hive.apache.org