CSE6242 / CX4242: Data & Visual Analytics

# Scaling Up Hadoop

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Partly based on materials by Professors Guy Lebanon, Jeffrey Heer, John Stasko, Christos Faloutsos, Parishit Ram (GT PhD alum; SkyTree), Alex Gray

#### How to handle data that is really large?

Really big, as in...

- Petabytes (PB, about 1000 times of terabytes)
- Or beyond: exabyte, zettabyte, etc.

Do we *really* need to deal with such scale?

Yes!

# "Big Data" is Common...

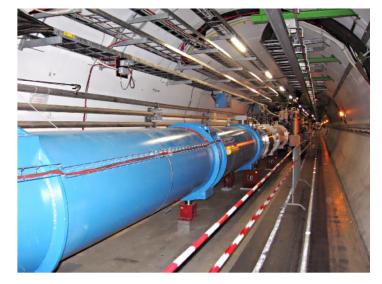
Google processed 24 PB / day (2009)

Facebook's add 0.5 PB / day to its data warehouses



Avatar's 3D effects took 1 PB to store

So, think **BIG!** 





First thing, how to store them?

Single machine? 60TB SSD announced. \$\$\$\$...

#### Cluster of machines?

- How many machines?
- Need data backup, redundancy, recovery, etc.
- Need to worry about machine and drive failure.

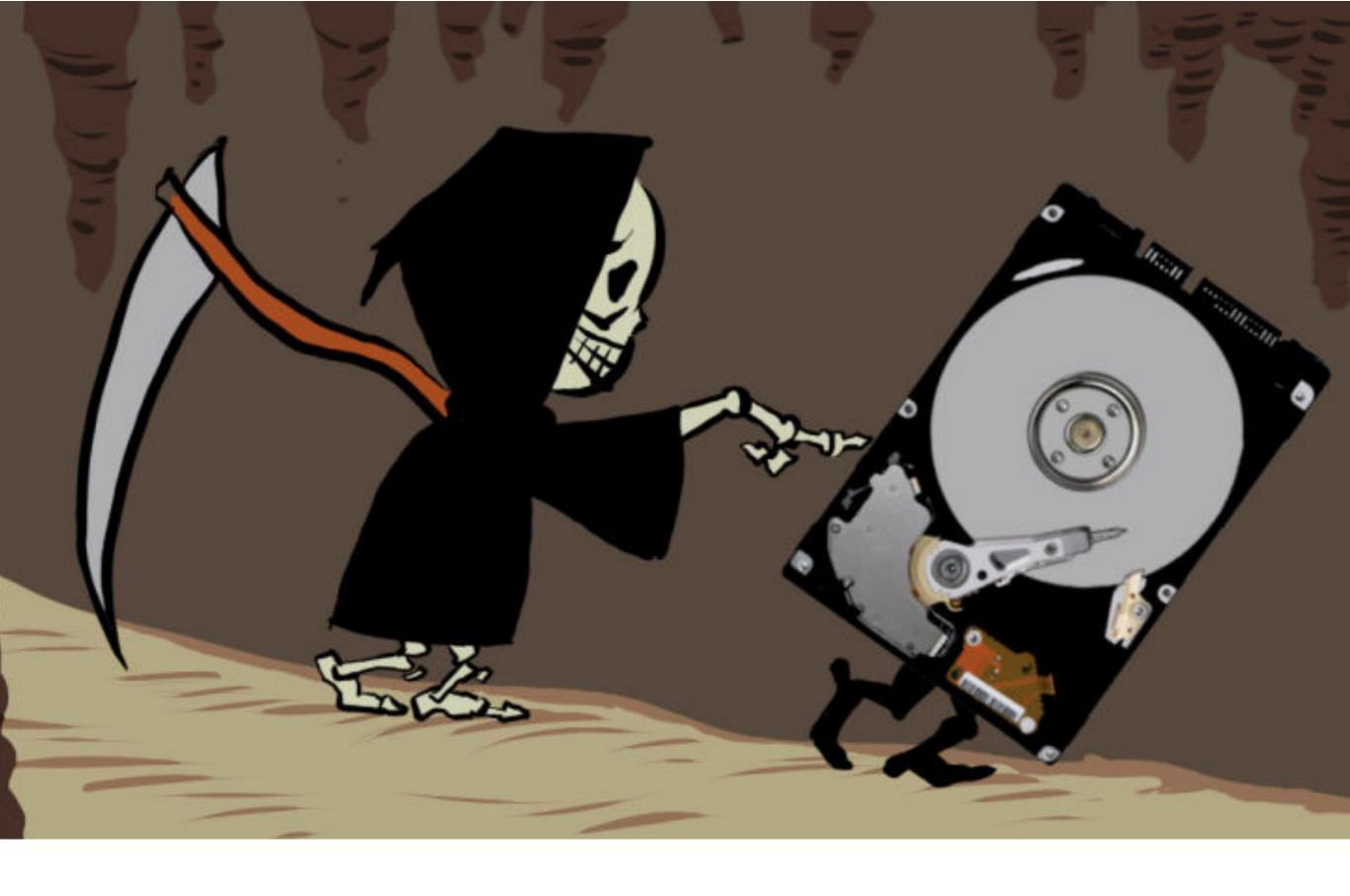
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# Really? Really???



http://lifehacker.com/how-long-will-my-hard-drives-really-last-1700405627

3% of 100,000 hard drives fail within first 3 months

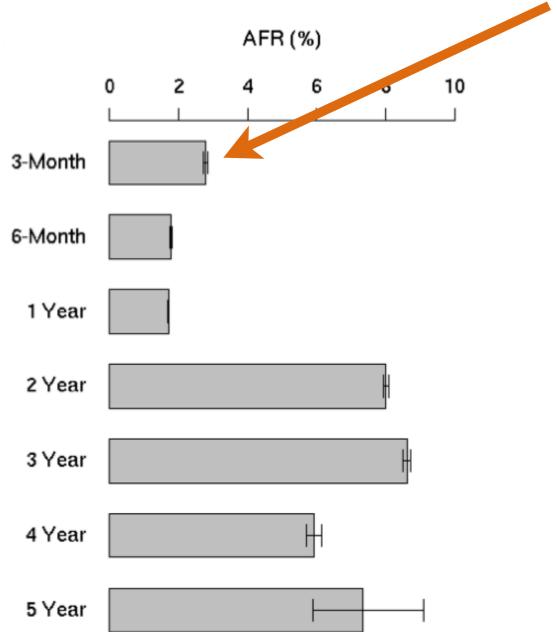
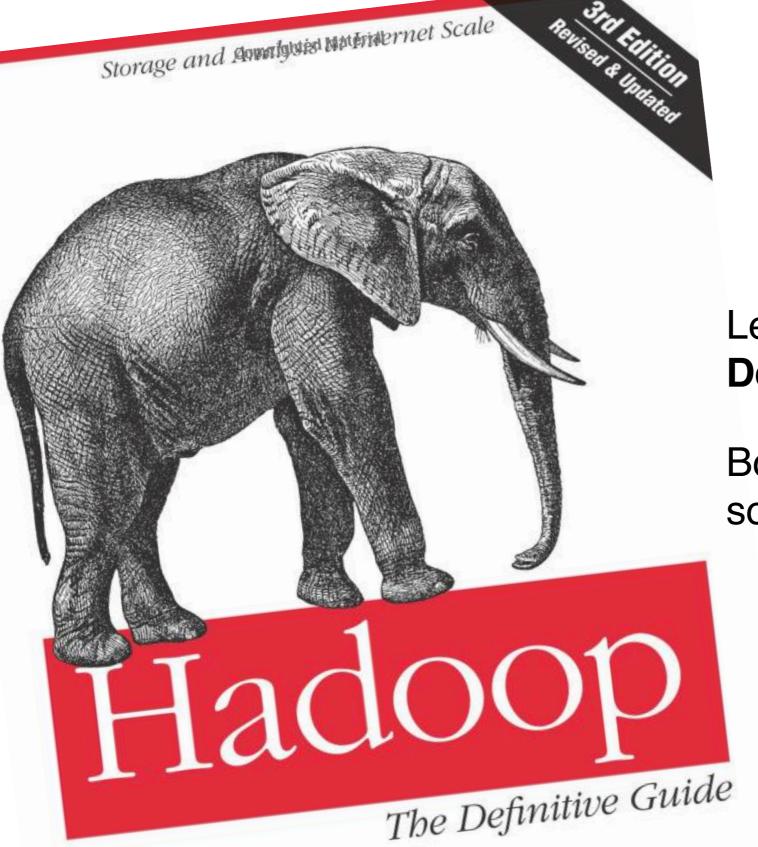


Figure 2: Annualized failure rates broken down by age groups

Failure Trends in a Large Disk Drive Population

How to analyze them?

- What software libraries to use?
- What programming languages to learn?
- Or more generally, what framework to use?



Lecture based on **Hadoop: The Definitive Guide** 

Book covers Hadoop, some Pig, some HBase, and other things.

http://goo.gl/YNCWN





Open-source software for reliable, scalable, distributed computing

Written in Java

Scale to thousands of machines

 Linear scalability (with good algorithm design): if you have 2 machines, your job runs twice as fast

Uses simple programming model (MapReduce)

Fault tolerant (HDFS)

 Can recover from machine/disk failure (no need to restart computation)

# Why learn Hadoop?

Fortune 500 companies use it

Many research groups/projects use it

Strong community support, and favored/backed my major companies, e.g., IBM, Google, Yahoo, eBay, Microsoft, etc.

It's free, open-source

Low cost to set up (works on commodity machines)

Will be an "essential skill", like SQL

# Elephant in the room



Hadoop created by Doug Cutting and Michael Cafarella while at Yahoo

Hadoop named after Doug's son's toy elephant

#### How does Hadoop scale up computation?

Uses master-worker architecture, and a simple computation model called **MapReduce** (popularized by Google's paper)

Simple way to think about it

- 1. Divide data and computation into smaller pieces; each machine works on one piece
- 2. Combine results to produce final results

#### How does Hadoop scale up computation?

More technically...

#### 1. Map phase

Master node divides data and computation into smaller pieces; each worker node ("mapper") works on one piece independently in parallel

# 2. Shuffle phase (automatically done for you) Master sorts and moves results to "reducers"

#### 3. Reduce phase

Worker nodes ("reducers") combines results independently in parallel

#### Example:

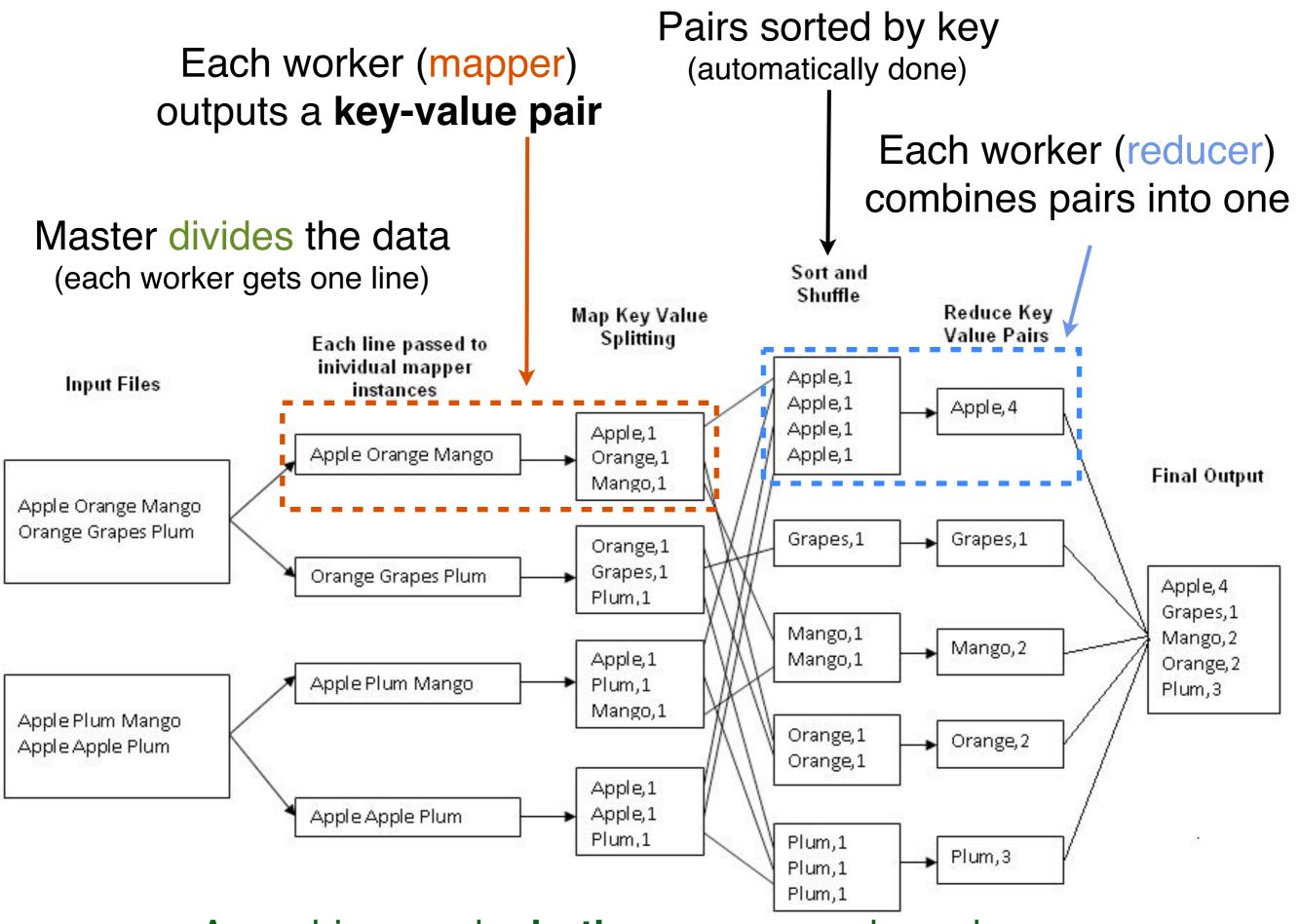
#### Find words' frequencies among text documents

#### Input

- "Apple Orange Mango Orange Grapes Plum"
- "Apple Plum Mango Apple Apple Plum"

#### **Output**

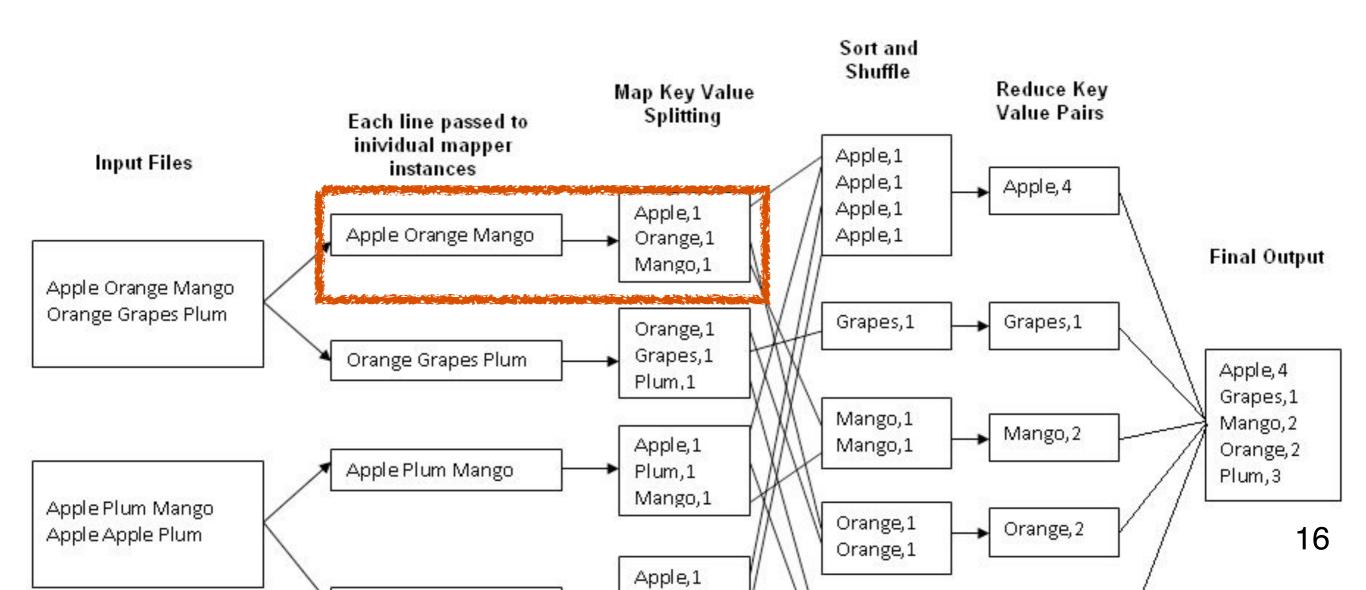
```
    Apple, 4
        Grapes, 1
        Mango, 2
        Orange, 2
        Plum, 3
```



A machine can be **both** a mapper and a reducer

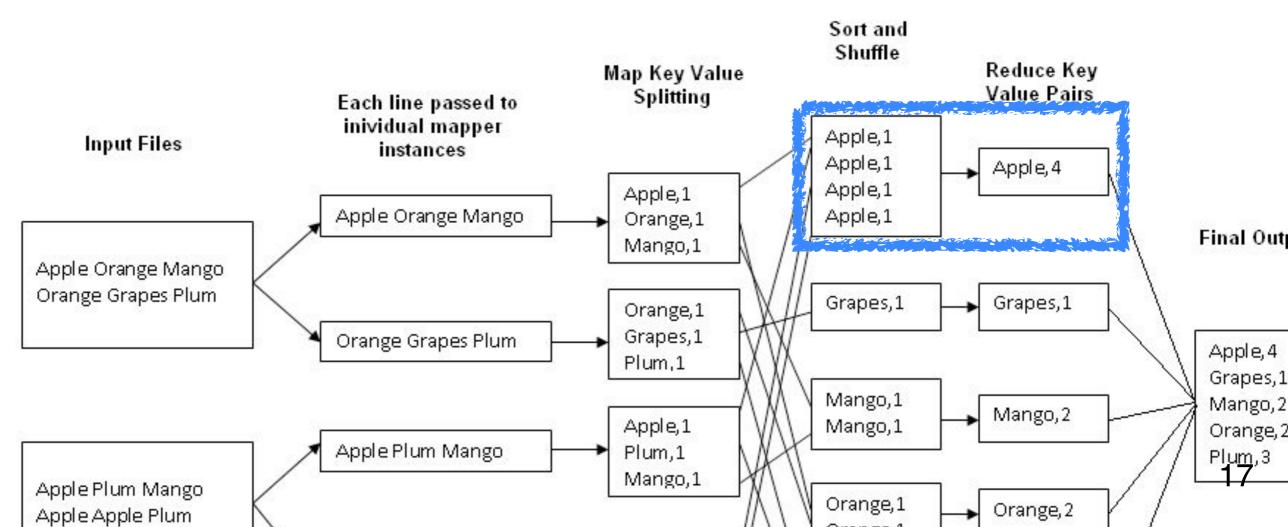
# How to implement this?

```
map(String key, String value):
    // key: document id
    // value: document contents
    for each word w in value:
        emit(w, "1");
```



# How to implement this?

```
reduce(String key, Iterator values):
    // key: a word
    // values: a list of counts
    int result = 0;
    for each v in values:
        result += ParseInt(v);
    Emit(AsString(result));
```



### What if a machine dies?

#### Replace it!

 "map" and "reduce" jobs can be redistributed to other machines

Hadoop's HDFS (Hadoop File System) enables this

# HDFS: <u>Hadoop File System</u>

A distribute file system

Built on top of OS's existing file system to provide redundancy and distribution

HDFS hides complexity of distributed storage and redundancy from the programmer

In short, you don't need to worry much about this!

# "History" of HDFS and Hadoop

#### Hadoop & HDFS based on...

- 2003 Google File System (GFS) paper http://cracking8hacking.com/cracking-hacking/Ebooks/Misc/pdf/The%20Google%20filesystem.pdf
- 2004 Google MapReduce paper

  http://static.googleusercontent.com/media/research.google.com/en/us/archive/mapreduce-osdi04.pdf

# What can you use Hadoop for?

As a "swiss knife".

Works for many types of analyses/tasks (but not all of them).

What if you want to write less code?

 There are tools to make it easier to write MapReduce program (Pig), or to query results (Hive)

# How to try Hadoop?

Hadoop can run on a single machine (e.g., your laptop)

Takes < 30min from setup to running</li>

Or a "home-grown" cluster

 Research groups often connect retired computers as a small cluster

Amazon EC2 (Amazon Elastic Compute Cloud), Microsoft Azure

- You only pay for what you use, e.g, compute time, storage
- You will use it in our next assignment