

Two Categories:

Private Cloud & Public Cloud

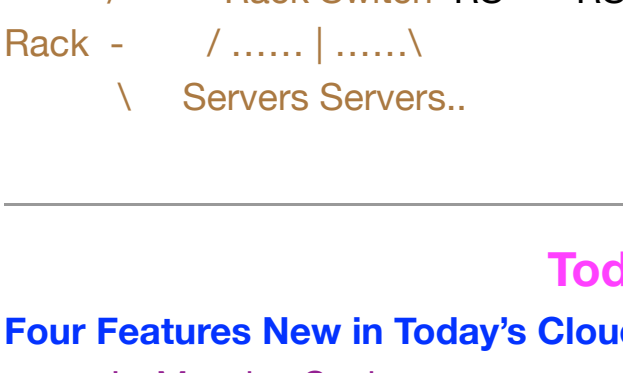
private cloud: accessible only to company employees

public cloud: provide service to any paying customer

- Amazon s3(simple storage service): Store arbitrary datasets, pay GBs/Month;
- Amazon EC2(Elastic computing cloud): Upload & run arbitrary OS images, pay CPUs/h
- Google App Engine/Comput Engine: Develop applications within their app engine framework, upload data that will be imported into their format and run

Cloud = Lots of storage + compute cycles nearby

Simple Cloud Topology:



Today's Cloud

Four Features New in Today's Cloud

- I - Massive Scale
- II - On-demand Access: Pay-as-you-go, no upfront commitment
Anyone can access it
- III - Data Intensive Nature: MBs become TBs, PBs and XBs
Daily logs, forensics, web data ...
Human has data numbness: Wiki compress is Only 10 GB
- IV - New Cloud Programming Paradigms: Mapreduce/Hadoop, NoSQL/Cassandra/MangoDB and many others
High in accessibility and ease of programmability
Lots of open-source

II - On-demand Access: *aaS Classification

Renting a cab vs. (previously) renting a car, or buying one. Ex.:

- AWS Elastic Compute Cloud (EC2) (renting a cab):
a few cents to a few \$ per CPU hour
- AWS Simple Storage Service (S3) (renting a car or having one already):
a few cents to a few \$ per GB-month

HaaS: Hardware as a Service

- You get access to barebones hardware machines, do whatever you want with them, ex: your own cluster
- Not always a good idea because of security risks

IaaS: Infrastructure as a Service

- You get access to flexible computing and storage infrastructure.
Virtualization is one way of achieving this (what's another way, e.g., using Linux). **Often said to subsume HaaS.**
- Ex: Amazon Web Services (AWS: EC2 and S3), Eucalyptus, Rightscale, Microsoft Azure

PaaS: Platform as a Service

- You get access to flexible computing and storage infrastructure, coupled with a software platform (often tightly)
- Ex: Google's AppEngine (Python, Java, Go)

SaaS: Software as a Service

- You get access to software services, when you need them. Often said to subsume SOA (Service Oriented Architectures).
- Ex: Google docs, MS Office on demand

III - Data-intensive Computing

Computation-Intensive Computing

- Example areas: MPI-based, high- performance computing, grids
- Typically run on supercomputers (e.g., NCSA Blue Waters)

Data-Intensive

- Typically store data at datacenters
- Use compute nodes nearby
- Compute nodes run computation services

In data-intensive computing, the focus shifts from computation to the data

- CPU utilization no longer the most important resource metric, instead I/O is (disk and/or network)

IV - New Cloud Programming Paradigms

Easy to write and run highly parallel programs in new cloud programming paradigms:

Google: MapReduce and Sawzall

Amazon: Elastic MapReduce service (pay-as-you-go)

Google (MapReduce)

- Indexing: a chain of 24 MapReduce jobs
- ~200K jobs processing 50PB/month (in 2006)

Yahoo! (Hadoop + Pig)

- WebMap: a chain of 100 MapReduce jobs
- 280 TB of data, 2500 nodes, 73 hours

Facebook (Hadoop + Hive)

- ~300TB total, adding 2TB/day (in 2008)
- 3K jobs processing 55TB/day

Similar numbers from other companies, e.g., Yeldex, eharmony.com, etc.

NoSQL

- MySQL is an industry standard, but Cassandra is 2400 times faster!

MapReduce Topic

Video: <https://class.coursera.org/cloudcomputing-001/wiki/Week1Overview>

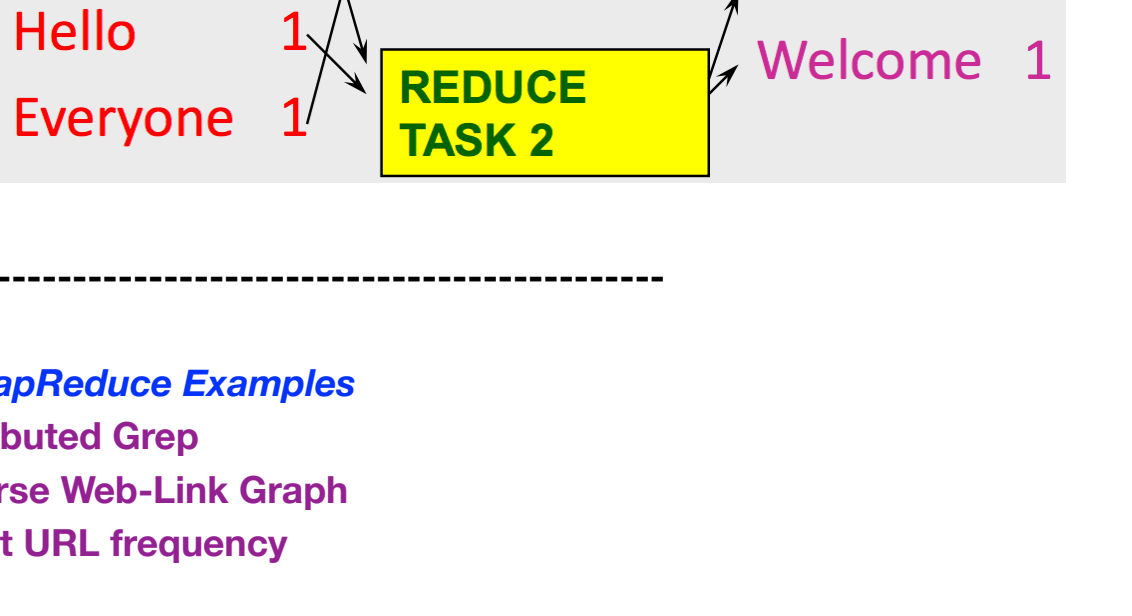
(1) MapReduce Paradigm

Slides:

https://d396qusza40orc.cloudfront.net/cloudcomputing/lecture_notes/w1/C3_Mapreduce_A_CSRAfinal.pdf

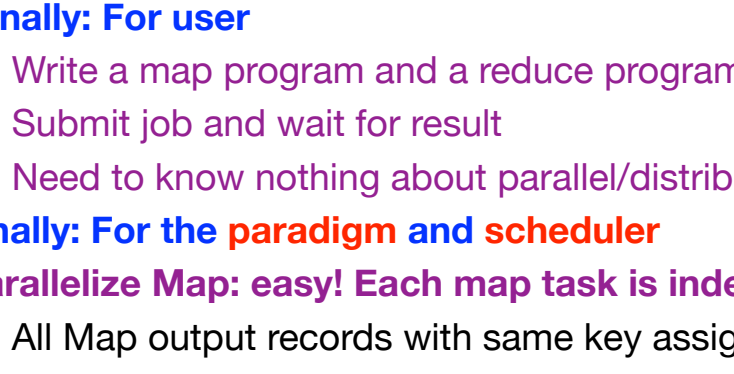
Map

- Process individual records to generate intermediate key/value pairs
- **Parallely** process individual records to generate intermediate key/value pairs



Reduce

- Reduce process and merges all intermediate value associated per key



(2) MapReduce Examples

Distributed Grep

Reverse Web-Link Graph

Count URL frequency

Sort

(3) MapReduce Scheduling

Externally: For user

- Write a map program and a reduce program
- Submit job and wait for result
- Need to know nothing about parallel/distributed programming

Internally: For the paradigm and scheduler

(1) Parallelize Map: easy! Each map task is independent of the other!

- All Map output records with same key assigned to same Reduce

(2) Transfer data from Map to Reduce:

- All Map output records with same key assigned to same Reduce task
- Use partitioning function: **hash(key) % (number of reducers)**

(3) Parallelize Reduce: easy! Each reduce task is independent of the other!

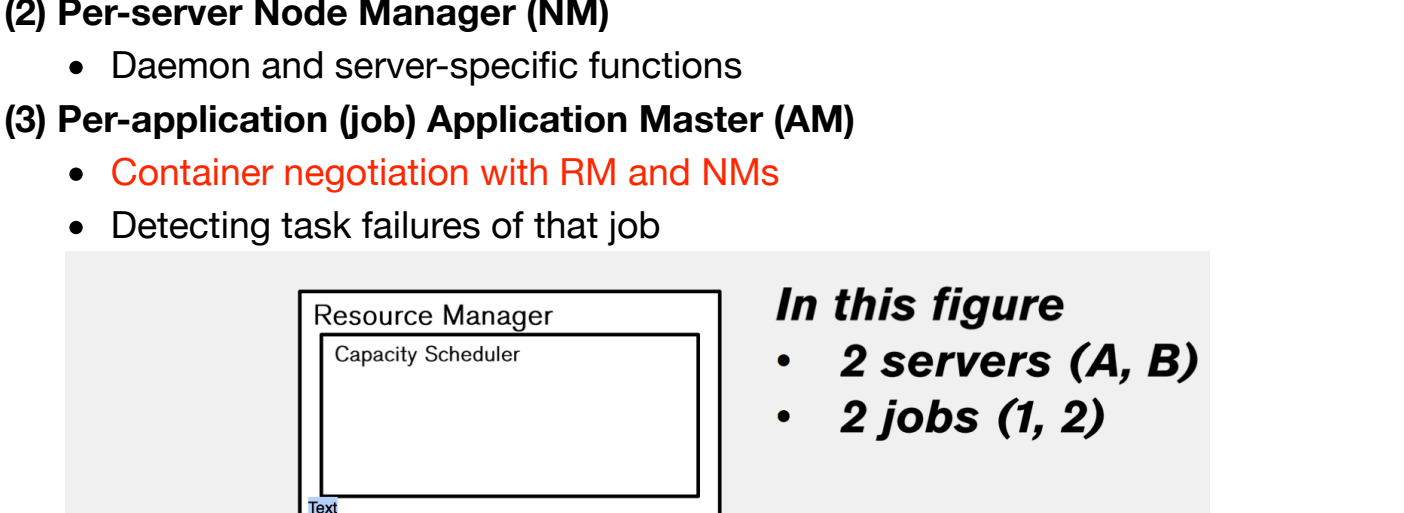
(4) Implement Storage for Map inout, Map output, Reduce input and Reduce output:

- Map input: from **Distributed File System**
- Map output: to local disk (at Map node); uses **Local File System**
- Reduce input: from (multiple) remote disks; uses **Local File System**
- Reduce output: to **Distributed File System**

Local File System = Linux FS, etc

Distributed File System = GFS (Google File System), HDFS (Hadoop Distributed File System)

Internal Workings of MapReduce



YARN Scheduler (Yet Another Resource Negotiator)

Used in Hadoop 2.x +

Treats each server as a collection of containers

- Container = some CPU + some memory

Has 3 main components

(1) Global Resource Manager (RM)

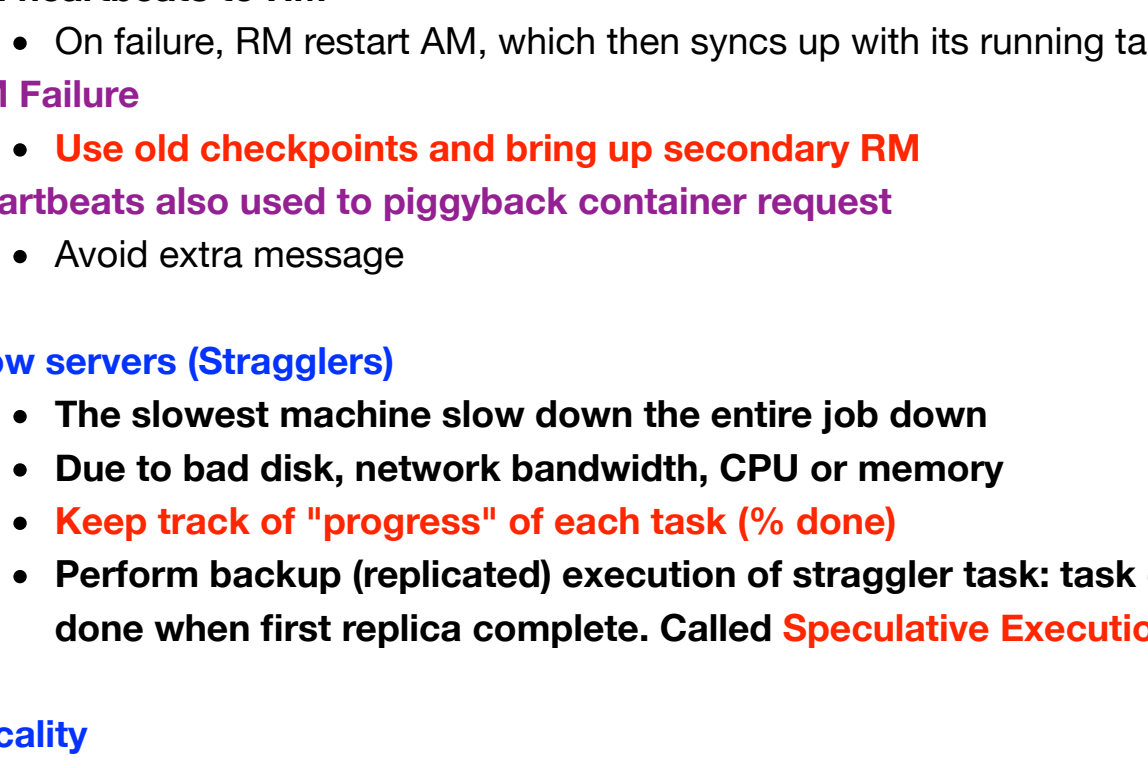
- Scheduling

(2) Per-server Node Manager (NM)

- Daemon and server-specific functions

(3) Per-application (job) Application Master (AM)

- Container negotiation with RM and NMs
- Detecting task failures of that job



In this figure

- 2 servers (A, B)
- 2 jobs (1, 2)

(4) MapReduce Fault Tolerance

Server Failure

NM (per-server) heartbeats to RM

- If server fails, RM then let all effected AMs know, and AMs take action

NM keeps track of each task running at its server

- If task fails while in-progress, mark the task as idle and restart it

AM heartbeats to RM

- On failure, RM restart AM, which then syncs up with its running tasks

RM Failure

- Use old checkpoints and bring up secondary RM

Heartbeats also used to piggyback container request

- Avoid extra message

Slow servers (Stragglers)

- The slowest machine slow down the entire job down
- Due to bad disk, network bandwidth, CPU or memory
- Keep track of "progress" of each task (% done)
- Perform backup (replicated) execution of straggler task: task consider done when first replica complete. Called **Speculative Execution**

Locality

Since cloud has hierarchical topology (e.g. racks)

GFS/HDFS stores 3 replicas of each of chunks (e.g., 64 MB in size)

- maybe on different racks

MapReduce attempts to schedule a map task on

- A machine that contains a replica of corresponding input data, or failing that
- The same rack as a machine containing the input, or failing that
- Anywhere