INFM 603: Information Technology and Organizational Context Session II: Cloud Computing and Big Data



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Thursday, November 14, 2013

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What is the Matrix?

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What is cloud computing?

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Source: Wikipedia (Clouds)

Source: flickr (60in3/2338247189)

The best thing since sliced bread?

- Before clouds...
 - Grids
 - Connection machines
 - Vector supercomputers
 - ...
- Cloud computing means many different things:
 - Large-data processing
 - Rebranding of web 2.0
 - Utility computing
 - Everything as a service

Rebranding of web 2.0

- Rich, interactive web applications
 - Clouds refer to the servers that run them
 - AJAX as the de facto standard (for better or worse)
 - Examples: Facebook, YouTube, Gmail, ...
- "The network is the computer": take two
 - User data is stored "in the clouds"
 - Rise of the tablets, smartphones, etc.
 - Browser is the OS

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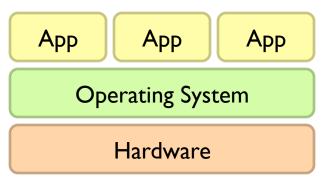
Utility Computing

- What?
 - Computing resources as a metered service ("pay as you go")
 - Ability to dynamically provision virtual machines
- Why?
 - Cost: capital vs. operating expenses
 - Scalability: "infinite" capacity
 - Elasticity: scale up or down on demand
- Does it make sense?
 - Benefits to cloud users
 - Business case for cloud providers

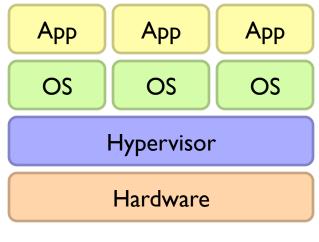
I think there is a world market for about five computers.



Enabling Technology: Virtualization



Traditional Stack



Virtualized Stack

Everything as a Service

- Utility computing = Infrastructure as a Service (IaaS)
 - Why buy machines when you can rent them?
 - Examples: Amazon's EC2, Rackspace
- Platform as a Service (PaaS)
 - Give me nice API and take care of the maintenance, upgrades, ...
 - Example: Google App Engine
- Software as a Service (SaaS)
 - Just run it for me!
 - Example: Gmail, Salesforce

Different Types of Clouds

- Public clouds
- Private clouds
- Hybrid clouds

Our World: Large Data

Source: Wikipedia (Hard disk drive)

Google[™] processes 20 PB a day (2008)

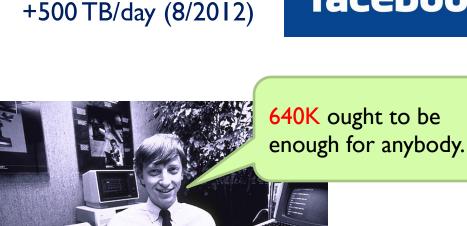


Wayback Machine: 10 PB web archive (10/2012)

10 PB data in Hadoop/Teradata 75B DB calls/day (6/2012)

LHC: 15 PB a year





100+ PB user data

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facebook.



LSST: 6-10 PB a year (~2015)

How much data?

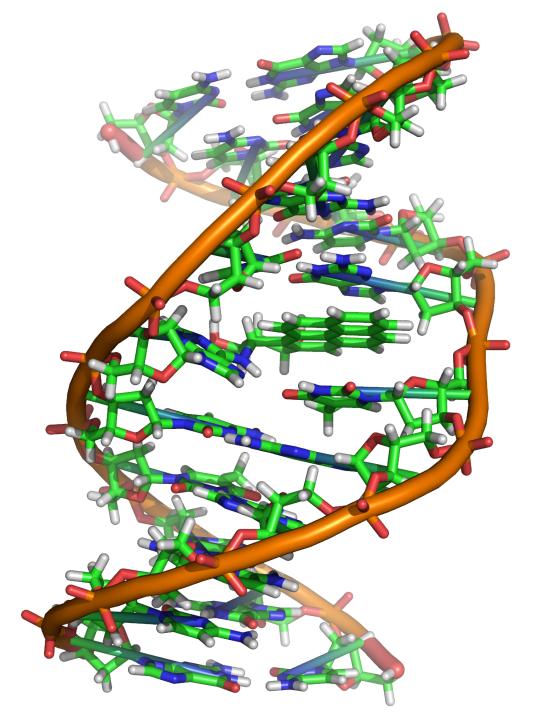
Why large data? Science Engineering Commerce



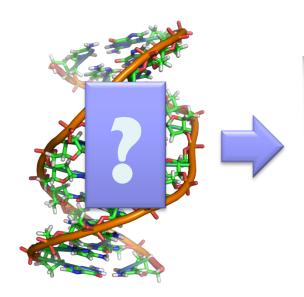
Science

- Emergence of the 4th Paradigm
- Data-intensive e-Science





Source: Wikipedia (DNA)



Subject genome



GATGCTTACTATGCGGGGCCCC CGGTCTAATGCTTACTATGC GCTTACTATGCGGGGCCCCTT AATGCTTACTATGCGGGGCCCCTT TAATGCTTACTATGC AATGCTTAGCTATGCGGGCC AATGCTTACTATGCGGGGCCCCTT CGGTCTAGATGCTTACTATGC AATGCTTACTATGCGGGCCCCTT CGGTCTAATGCTTAGCTATGC ATGCTTACTATGCGGGCCCCTT

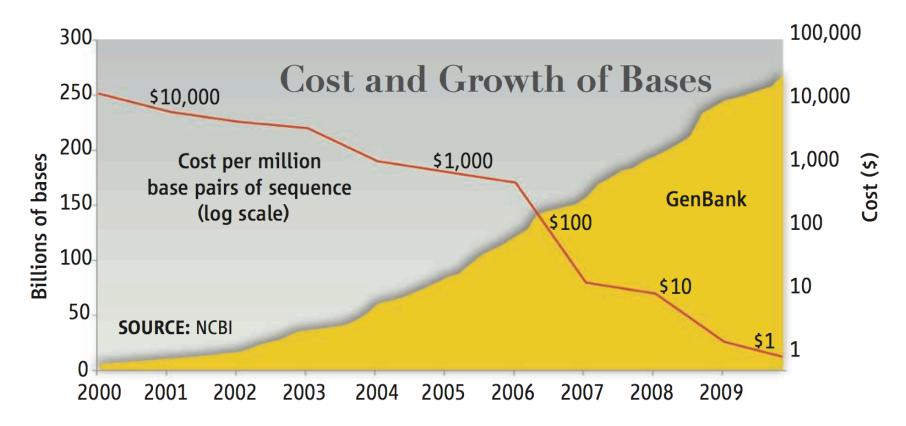
Reads

Human genome: 3 gbp A few billion short reads (~100 GB compressed data)

Sequencer

DNA Data Tsunami

Current world-wide sequencing capacity exceeds 13 Pbp/year and is growing at 5x per year!



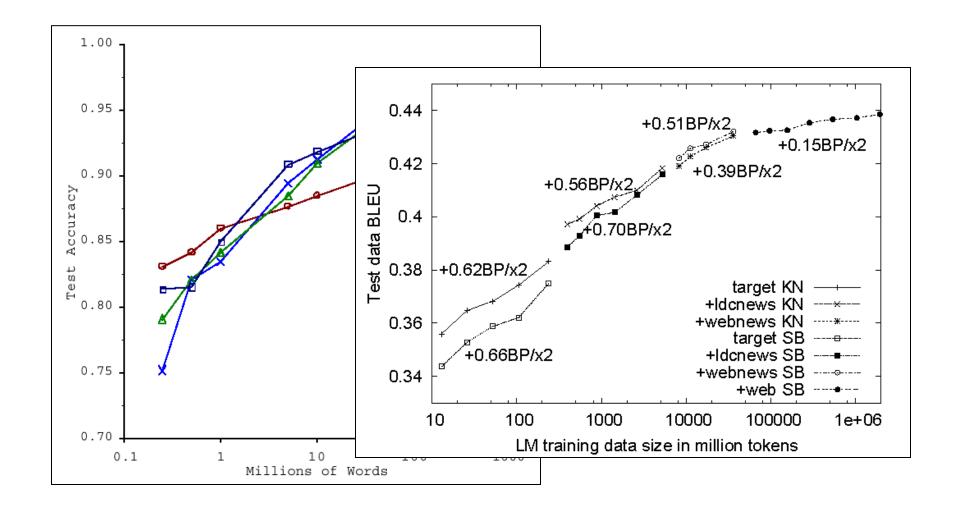
[&]quot;Will Computers Crash Genomics?" Elizabeth Pennisi (2011) Science. 331(6018): 666-668.



Engineering

- The unreasonable effectiveness of data
- Count and normalize!

No data like more data!



(Banko and Brill, ACL 2001) (Brants et al., EMNLP 2007)

What to do with more data?

• Answering factoid questions

- Pattern matching on the Web
- Works amazingly well

Who shot Abraham Lincoln? \rightarrow X shot Abraham Lincoln

- Learning relations
 - Start with seed instances
 - Search for patterns on the Web
 - Using patterns to find more instances

Wolfgang Amadeus Mozart (1756 - 1791) Einstein was born in 1879

Birthday-of(Mozart, 1756) Birthday-of(Einstein, 1879)

PERSON (DATE – PERSON was born in DATE



Commerce

- Know thy customers
- Data \rightarrow Insights \rightarrow Competitive advantages

Business Intelligence

- Premise: more data leads to better business decisions
 - Periodic reporting as well as ad hoc queries
 - Rise of the data scientist
 - Listen to your customers, not the HiPPO
- Examples:
 - Slicing-and-dicing activity by different dimensions to better understand the marketplace
 - Analyzing log data to improve front-end experience
 - Analyzing log data to better optimize ad placement
 - Analyzing purchasing trends for better supply-chain management
 - Mining for correlations between otherwise unrelated activities

Database Workloads

• OLTP (online transaction processing)

- Typical applications: e-commerce, banking, airline reservations
- User facing: real-time, low latency, highly-concurrent
- Tasks: relatively small set of "standard" transactional queries
- Data access pattern: random reads, updates, writes (involving relatively small amounts of data)

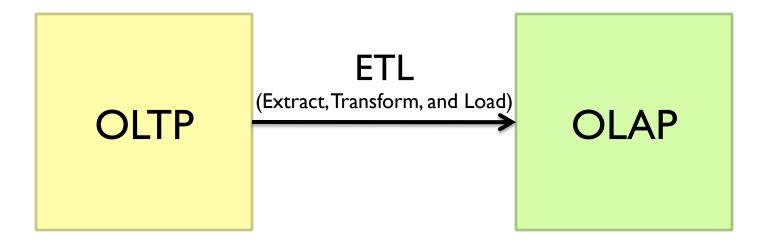
• OLAP (online analytical processing)

- Typical applications: business intelligence, data mining
- Back-end processing: batch workloads, less concurrency
- Tasks: complex analytical queries, often ad hoc
- Data access pattern: table scans (involving large amounts of data)

One Database or Two?

- Downsides of co-existing OLTP and OLAP workloads
 - Poor memory management
 - Conflicting data access patterns
 - Variable latency
- Solution: separate databases
 - User-facing OLTP database for high-volume transactions
 - Data warehouse for OLAP workloads
 - How do we connect the two?

OLTP/OLAP Architecture



OLTP/OLAP Integration

- OLTP database for user-facing transactions
 - Retain records of all activity
 - Periodic ETL (e.g., nightly)
- Extract-Transform-Load (ETL)
 - Extract records from source
 - Transform: clean data, check integrity, aggregate, etc.
 - Load into OLAP database
- OLAP database for data warehousing
 - Business intelligence: reporting, ad hoc queries, data mining, etc.
 - Feedback to improve OLTP services

Challenge of Big Data

- Volume
- O Cost
- ETL Latency

cloud computing meets big data

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Source: Wikipedia (Clouds)

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Cloud Computing Meets Big Data

- Rise of social media and user-generated content
 - Cloud services exacerbates big data problems
- Utility computing democratizes big data capabilities
 - Efficient dynamic allocation of large-scale computing resources

What really is the cloud?

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Source: Wikipedia (Clouds)

Source: Wikipedia (The Dalles, Oregon)











Source: Facebook





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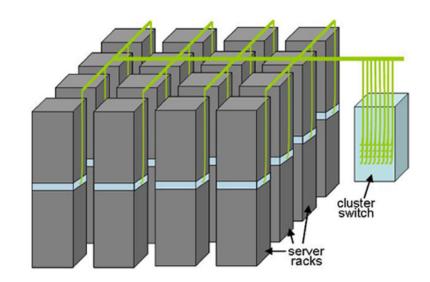
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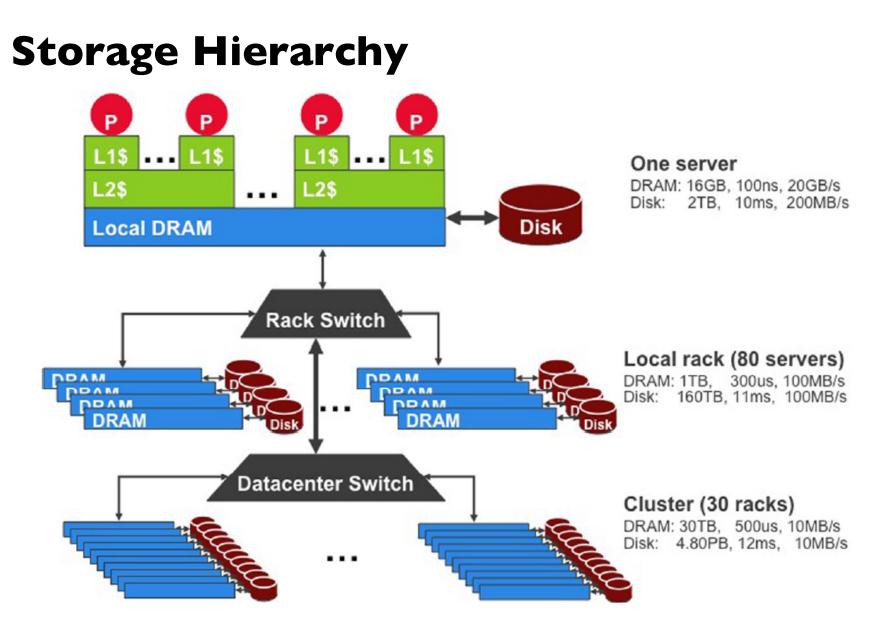
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Building Blocks





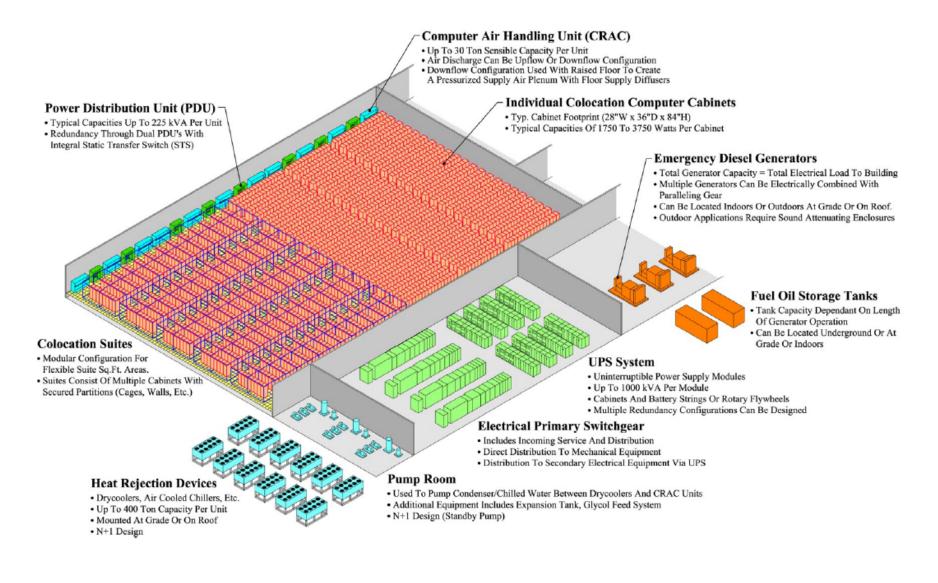




Funny story about sense of scale...

Source: Barroso and Urs Hölzle (2009)

Anatomy of a Datacenter





Source: CumminsPower

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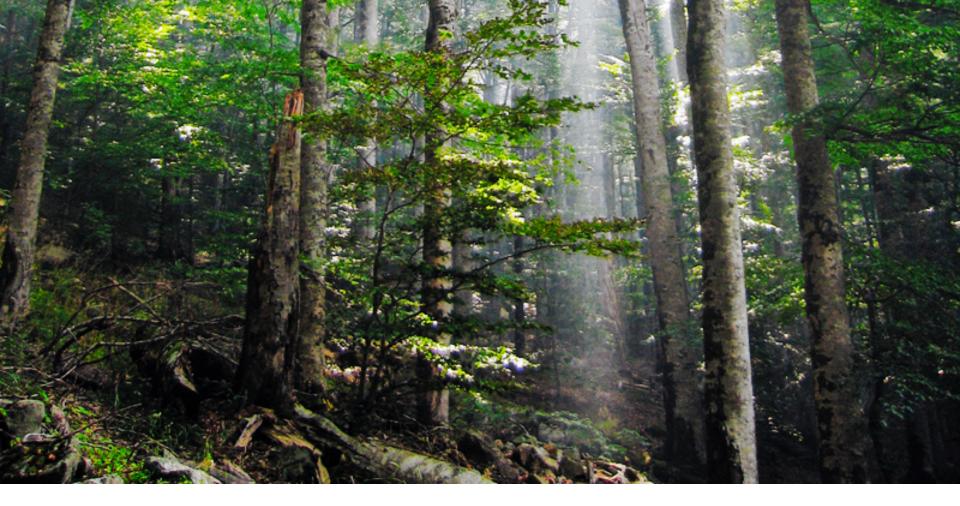
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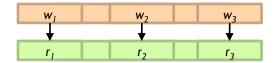
How large data?

Source: Wikipedia (Everest)



Divide et impera

- Chop problem into smaller parts
- Combine partial results



Synchronization Challenges

- How to split large chunks up into smaller ones
- How to integrate results from each chunk
- How to distribute shared data
- How to update shared data
- How to coordinate access to shared resources
- How to schedule different processing chunks
- How to cope of machine failure

Source: Ricardo Guimarães Herrmann

Typical Large-Data Problem

• Iterate over a large number of records

Maaxtract something of interest from each

• Shuffle and sort intermediate results

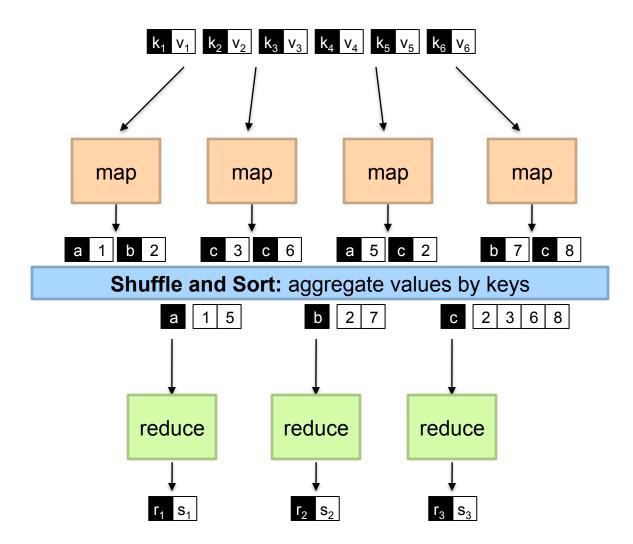
- Aggregate intermediate results Reduce
- Generate final output

MapReduce

• Programmers specify two functions:

map $(k, v) \rightarrow \langle k', v' \rangle^*$ reduce $(k', v') \rightarrow \langle k', v' \rangle^*$

- All values with the same key are sent to the same reducer
- The execution framework handles everything else...



MapReduce

• Programmers specify two functions:

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- The execution framework handles everything else...

What's "everything else"?

MapReduce "Runtime"

- Handles scheduling
 - Assigns workers to map and reduce tasks
- Handles "data distribution"
 - Moves processes to data
- Handles synchronization
 - Gathers, sorts, and shuffles intermediate data
- Handles errors and faults
 - Detects worker failures and restarts

MapReduce Word Count

Map(String docid, String text):

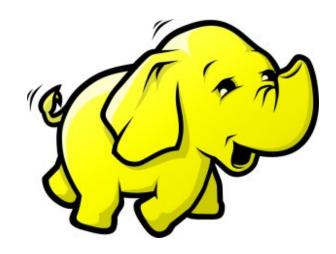
for each word w in text: Emit(w, I);

Reduce(String term, Iterator<Int> values):

int sum = 0; for each v in values: sum += v; Emit(term, value);

MapReduce Implementations

- Google has a proprietary implementation
- Hadoop is an open-source implementation in Java
 - Originally developed by Yahoo, now an Apache project
 - Center of a rapidly expanding software ecosystem



Now you know...

- Cloud computing
- Big data
- Relationship between the two
- Challenges with big data processing
- MapReduce/Hadoop

Questions?

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Source: Wikipedia (Clouds)