



Data Center Architecture

- Earlier days of parallel computing
 - Scientific computing is parallelizable, but complex synchronization glue the parallel processes
 - Distributed databases are also tightly coupled
 - Many points of failure in commodity machine clusters
 - Supercomputers with custom network and node design: close integration but expensive, hard to scale
- Late 90s, early 2000
 - Embarrassingly parallel web services: web search, e-commerce
 - Commodity machines and networks, glued together by software
 - Turn independent failures from a curse to a strength, migrate a server farm
 - Cheaper at scale since the software cost is amortized

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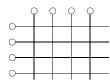
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Network Fabric

- Network matters to distributed data processing
- A Gbps network switch allows simultaneous traffic at full link speed
 - Crossbar: Complexity grows up quickly with the number of ports



- Switch with a small number of ports is cheaper (~\$30 per port)
- Cost goes up non-linearly (an order of magnitude higher per port cost)
- One design choice
 - Cheap, few-port switch to link machines on a rack
 - Expensive, many-port switch to link all the racks together

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Network Switching at Data Centers

Thousands of servers with hierarchical network switching



- Reduce congestions:
 - Network load balance across nodes
 - Place highly communicating server modules within same switch domain, or better yet, in same machines
 - Redundant backbone links → Equal-Cost Multi-Path (ECMP) routing

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Storage Architecture

- Each machine has local storage, managed by distributed file system such as GFS (Colossus), HDFS
- In addition (alternatively), storage-level parallelism
 - specialized storage systems plugged into the network switch that manages replication, parallel I/O etc. (scalable RAID system)
- Contrast
 - Storage-level parallelism generally provides better I/O performance
 - File system parallelism integrates better with higher-level software (such as MapReduce), facilitates data locality through smart software management, and puts less stress on network

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Memory/Storage Hierarchy

- CPU cache, memory, Flash SSD, and disk
 - Larger space but longer latency
- In a data center, we have CPU cache, local memory, local disk, rack memory, rack disk, data center memory, data center disk
 - Figures 1.3, 1.4 (page 10)

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Data Center Workloads

- Unique, large jobs
 - E.g., compute page ranks from the 50B crawled web pages
- Many, small tasks
 - E.g., Amazon cloud computing, Google search service
- Parallelism
 - Data parallelism
 - User / request parallelism

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Platform / Application Adaptation

- Platform: basic hardware and infrastructure system software
- Applications developed on top
- Platform / application adaptation
 - Platforms are somewhat homogeneous
 - Application upgrades are fast
 - ⇒ Focus on basic cost-effectiveness on the platform construction and let application software adapt to the platform

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Infrastructure Software: Buy, Adapt, Build

- Infrastructure software: operating system, file systems, databases, parallel computation, replication, ...
- Choices:
 - Buy from vendors
 - Take open-source software and make adaptation
 - Build new software from scratch
- Google's argument: adapt or build, but don't buy
 - Need full control (performance debugging), flexibility, customization (nosql)
 - Few vendors (think of databases) are capable of building and testing their software at the datacenter scale

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Cost-Efficient Analysis on Multiprocessor

- You have N parallel tasks
 - c is the computation time, d is the number of data accesses, I_{local} is the latency of local data access, I_{remote} is the latency of remote data access
- If each task runs on a separate machine
 - Assume data is evenly distributed, the chance of a data access to be local is 1/N
 - Then the execution time is: $c + d \cdot (I_{local} \cdot 1/N + I_{remote} \cdot (1-1/N))$
- If you have a large multiprocessor (host all N tasks):
 - Then the execution time is: c + d · I_{local}
- In practice, your multiprocessor isn't large enough to host all (host n<N tasks)
 - Then the execution time is: $c + d \cdot (I_{local} \cdot n/N + I_{remote} \cdot (1-n/N))$
 - Figure 3.2 (page 36)

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Power

- Power consumption of a large data center is significant
- The power distribution system is also expensive
 - In building cost and power lost during distribution
- People care about the total energy usage and peak power
- What do people do?
 - Build your data center next to a hydro power plant
 - Energy conservation through server consolidation
 - Weigh spare capacity against load spikes and speed of server powerup
 - Lower power spikes through load shifting or throttling

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Conventional vs. Wimpy CPUs

- Conventional CPU (regular CPUs on desktops / servers)
- Wimpy CPU (what you have in a smart phone)
- Why wimpy CPUs?
 - With good parallelism, a larger number of wimpy CPUs can do the job done by a smaller number of conventional CPUs
 - Cluster of wimpy CPUs may save power and save money
 - E.g., a 2-Watt wimpy processor is 5 times slower than a 50-Watt server processor
 potentially 5 times higher energy efficiency
- But the assumption of "good parallelism"
 - Now you know that good parallelism isn't easy to get

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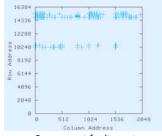
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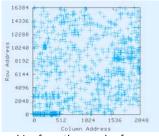
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Reliability at Scale

Chance of failures/anomalies to occur in a large system is high





- Permanent faults on two memory chips from thousands of memory chips we examined in a production system
- Transient faults were also found

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Reliability at Scale

- Why does a machine fail?
 - Hardware problems, software problems
- Reliability rate for one machine is p, for n machines is pⁿ
- In a very large system, you may not always be able to find out problem cause and fix the problem, so
 - Detect problem and tolerate / mitigate it
 - Redundancy / replication
 - Memory ECC
 - Machines
 - Data storage
 - Replication of mutable states introduces consistency problems
 - Networking
 - Power supply

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Reliability at Scale

- Performance anomalies
 - A particular request / map task is very slow
 - Technique to tolerate slow tail: perform work redundantly if the current run is too slow
- Recovery-oriented computing
 - Components are inherently unreliable, design a system in which components can be independently recovered (rebooted) without significant effects on others
 - Make sure that restart is fast (multi-scaled restart)

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Failure Correlation

- Correlated catastrophic failures
 - A bug-triggering input / request: Canaries
 - Load-triggered failures
 - Malicious activities

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Management

- Easy to add/remove components
 - Container-based data center modules

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Disclaimer

 Preparation of this class was helped by materials in the book "The Datacenter as a Computer" by Luiz A. Barroso, Jimmy Clidaras, and Urs Holzle.

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