## Lecture: Networks, Disks, Datacenters

- Topics: networks wrap-up, disks and reliability, datacenters and energy proportionality


## Centralized Crossbar Switch



## Crossbar Properties

- Assuming each node has one input and one output, a crossbar can provide maximum bandwidth: $N$ messages can be sent as long as there are N unique sources and $N$ unique destinations
- Maximum overhead: $W N^{2}$ internal switches, where $W$ is data width and N is number of nodes
- To reduce overhead, use smaller switches as building blocks - trade off overhead for lower effective bandwidth


## Switch with Omega Network



## Omega Network Properties

- The switch complexity is now $\mathrm{O}(\mathrm{N} \log \mathrm{N})$
- Contention increases: P0 $\rightarrow$ P5 and P1 $\rightarrow$ P7 cannot happen concurrently (this was possible in a crossbar)
- To deal with contention, can increase the number of levels (redundant paths) - by mirroring the network, we can route from P0 to P5 via N intermediate nodes, while increasing complexity by a factor of 2


## Tree Network

- Complexity is $\mathrm{O}(\mathrm{N})$
- Can yield low latencies when communicating with neighbors
- Can build a fat tree by having multiple incoming and outgoing links



## Bisection Bandwidth

- Split N nodes into two groups of $\mathrm{N} / 2$ nodes such that the bandwidth between these two groups is minimum: that is the bisection bandwidth
- Why is it relevant: if traffic is completely random, the probability of a message going across the two halves is $1 / 2$ - if all nodes send a message, the bisection bandwidth will have to be N/2
- The concept of bisection bandwidth confirms that the tree network is not suited for random traffic patterns, but for localized traffic patterns


## Topology Examples



Grid


Torus


Hypercube

| Criteria <br> 64 nodes | Bus | Ring | 2Dtorus | Hypercube | Fully <br> connected |
| :---: | :--- | :--- | :--- | :--- | :---: |
| Performance <br> Bisection <br> bandwidth |  |  |  |  |  |
| Cost <br> Ports/switch <br> Total links |  |  |  |  |  |

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| :---: | :---: | :---: | :---: | :---: | :---: |
| Performance |  |  |  |  |  |
| Diameter | 1 | 32 | 8 | 6 | 1 |
| Bisection BW | 1 | 2 | 16 | 32 | 1024 |
| Cost |  |  |  |  |  |
| Ports/switch |  | 3 | 5 | 7 | 64 |
| Total links | 1 | 64 | 128 | 192 | 2016 |

## k-ary d-cube

- Consider a k-ary d-cube: a d-dimension array with k elements in each dimension, there are links between elements that differ in one dimension by $1(\bmod k)$
- Number of nodes $N=k^{d}$

Number of switches:
Switch degree
Number of links
Pins per node

Avg. routing distance:
Diameter
Bisection bandwidth :
Switch complexity

Should we minimize or maximize dimension?

## k-ary d-Cube

- Consider a k-ary d-cube: a d-dimension array with $k$ elements in each dimension, there are links between elements that differ in one dimension by $1(\bmod k)$
- Number of nodes $N=k^{d}$

| Number of switches: | N | Avg. routing distance: | $\mathrm{d}(\mathrm{k}-1) / 4$ |  |
| :--- | :--- | :--- | :--- | :--- |
| Switch degree $:$ | $2 \mathrm{~d}+1$ | Diameter $:$ | $:$ | $\mathrm{d}(\mathrm{k}-1) / 2$ |
| Number of links | $:$ | Nd | Bisection bandwidth: | $2 \mathrm{wk}^{\mathrm{d}-1}$ |
| Pins per node | $:$ | 2 wd | Switch complexity : | $(2 \mathrm{~d}+1)^{2}$ |

The switch degree, num links, pins per node, bisection bw for a hypercube are half of what is listed above (diam and avg routing distance are twice, switch complexity is $(\mathrm{d}+1)^{2}$ ) because unlike the other cases, a hypercube does not have right and left neighbors.

Should we minimize or maximize dimension?

## Warehouse-Scale Computer (WSC)

- 100K+ servers in one WSC
- ~ $\$ 150 \mathrm{M}$ overall cost
- Requests from millions of users (Google, Facebook, etc.)
- Cloud Computing: a model where users can rent compute and storage within a WSC; there's an associated service-level agreement (SLA)
- Datacenter: a collection of WSCs in a single building, possibly belonging to different clients and using different hardware/architecture


## PUE Metric and Power Breakdown

- PUE = Total facility power / IT equipment power (power utilization effectiveness)
- It is greater than 1; ranges from 1.33 to 3.03 , median of 1.69
- The cooling power is roughly half the power used by servers
- Within a server, the approximate power distribution is as follows: Processors (33\%), DRAM memory (30\%), Disks (10\%), Networking (5\%), Miscellaneous (22\%)


## CapEx and OpEx

- Capital expenditure: infrastructure costs for the building, power delivery, cooling, and servers
- Operational expenditure: the monthly bill for energy, failures, personnel, etc.
- CapEx can be amortized into a monthly estimate by assuming that the facilities will last 10 years, server parts will last 3 years, and networking parts will last 4


## CapEx/OpEx Case Study

- 8 MW facility : facility cost: \$88M, server/networking cost: \$79M
- Monthly expense: $\$ 3.8 \mathrm{M}$. Breakdown:
- Servers 53\% (amortized CapEx)
- Networking 8\% (amortized CapEx)
- Power/cooling infrastructure 20\% (amortized CapEx)
- Other infrastructure 4\% (amortized CapEx)
- Monthly power bill 13\% (true OpEx)
- Monthly personnel salaries 2\% (true OpEx)


## Improving Energy Efficiency

- An unloaded server dissipates a large amount of power
- Ideally, we want energy-proportional computing, but in reality, servers are not energy-proportional
- Can approach energy-proportionality by turning on a few servers that are heavily utilized
- See figures on next two slides for power/utilization profile of a server and a utilization profile of servers in a WSC


## Power/Utilization Profile



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## Server Utilization Profile



Figure 6.3 Average CPU utilization of more than $\mathbf{5 0 0 0}$ servers during a 6-month period at Google. Servers are rarely completely idle or fully utilized, in-stead operating most of the time at between $10 \%$ and $50 \%$ of their maximum utilization. (From Figure 1 in Barroso and Hölzle [2007].) The column the third from the right in Figure 6.4 calculates percentages plus or minus $5^{\%} \% 8$ to come up with the weightings; thus, $1.2 \%$ for the $90 \%$ row means that $1.2 \%$ of servers were between $85 \%$ and $95 \%$ utilized.

## Problem 1

Assume that a server consumes 100W at peak utilization and 50W at zero utilization. Assume a linear relationship between utilization and power. The server is capable of executing many threads in parallel. Assume that a single thread utilizes $25 \%$ of all server resources (functional units, caches, memory capacity, memory bandwidth, etc.). What is the total power dissipation when executing 99 threads on a collection of these servers, such that performance and energy are close to optimal?

## Problem 1

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For near-optimal performance and energy, use 25 servers. 24 servers at $100 \%$ utilization, executing 96 threads, consuming 2400 W . The 25th server will run the last 3 threads and consume 87.5~W.

## Other Metrics

- Performance does matter, both latency and throughput
- An analysis of the Bing search engine shows that if a 200 ms delay is introduced in the response, the next click by the user is delayed by 500 ms ; so a poor response time amplifies the user's non-productivity
- Reliability (MTTF) and Availability (MTTF/MTTF+MTTR) are very important, given the large scale
- A server with MTTF of 25 years (amazing!) : 50K servers would lead to 5 server failures a day; Similarly, annual disk failure rate is $2-10 \% \rightarrow 1$ disk failure every hour


## Important Problems

- Reducing power in power-down states
- Maximizing utilization
- Reducing cost with virtualization
- Reducing data movement
- Building a low-power low-cost processor
- Building a low-power low-cost hi-bw memory
- Low-power low-cost on-demand reliability


## Magnetic Disks

- A magnetic disk consists of 1-12 platters (metal or glass disk covered with magnetic recording material on both sides), with diameters between 1-3.5 inches
- Each platter is comprised of concentric tracks (5-30K) and each track is divided into sectors (100-500 per track, each about 512 bytes)
- A movable arm holds the read/write heads for each disk surface and moves them all in tandem - a cylinder of data is accessible at a time


## Disk Latency

- To read/write data, the arm has to be placed on the correct track - this seek time usually takes 5 to 12 ms on average - can take less if there is spatial locality
- Rotational latency is the time taken to rotate the correct sector under the head - average is typically more than 2 ms (15,000 RPM)
- Transfer time is the time taken to transfer a block of bits out of the disk and is typically $3-65 \mathrm{MB} /$ second
- A disk controller maintains a disk cache (spatial locality can be exploited) and sets up the transfer on the bus (controller overhead)
- Reliability and availability are important metrics for disks
- RAID: redundant array of inexpensive (independent) disks
- Redundancy can deal with one or more failures
- Each sector of a disk records check information that allows it to determine if the disk has an error or not (in other words, redundancy already exists within a disk)
- When the disk read flags an error, we turn elsewhere for correct data


## RAID 0 and RAID 1

- RAID 0 has no additional redundancy (misnomer) - it uses an array of disks and stripes (interleaves) data across the arrays to improve parallelism and throughput
- RAID 1 mirrors or shadows every disk - every write happens to two disks
- Reads to the mirror may happen only when the primary disk fails - or, you may try to read both together and the quicker response is accepted
- Expensive solution: high reliability at twice the cost
- Data is bit-interleaved across several disks and a separate disk maintains parity information for a set of bits
- For example: with 8 disks, bit 0 is in disk-0, bit 1 is in disk-1, ..., bit 7 is in disk-7; disk-8 maintains parity for all 8 bits
- For any read, 8 disks must be accessed (as we usually read more than a byte at a time) and for any write, 9 disks must be accessed as parity has to be re-calculated
- High throughput for a single request, low cost for redundancy (overhead: 12.5\%), low task-level parallelism


## RAID 4 and RAID 5

- Data is block interleaved - this allows us to get all our data from a single disk on a read - in case of a disk error, read all 9 disks
- Block interleaving reduces thruput for a single request (as only a single disk drive is servicing the request), but improves task-level parallelism as other disk drives are free to service other requests
- On a write, we access the disk that stores the data and the parity disk - parity information can be updated simply by checking if the new data differs from the old data
- If we have a single disk for parity, multiple writes can not happen in parallel (as all writes must update parity info)
- RAID 5 distributes the parity block to allow simultaneous writes


## Other Reliability Approaches

- High reliability is also expected of memory systems; many memory systems offer SEC-DED support - single error correct, double error detect; implemented with an 8-bit code for every 64-bit data word on ECC DIMMs
- Some memory systems offer chipkill support - the ability to recover from complete failure in one memory chip - many implementations exist, some resembling RAID designs
- Caches are typically protected with SEC-DED codes
- Some cores implement various forms of redundancy, e.g., DMR or TMR - dual or triple modular redundancy

