CS370 Operating Systems

Colorado State University Yashwant K Malaiya Fall 2021 L23

Mass Storage



Slides based on

- Text by Silberschatz, Galvin, Gagne
- Various sources

FAQ

- Indexed allocation: Index blocks includes pointers to file data blocks.
- Inode: contains pointers to data blocks directly or indirectly.
 - All inodes are in the inode table on the disk.
 - Inode number gives the inode address and identifies a file.
- 2nd chance algorithm: If the reference bit is 1, give that page a second chance.
- Can windows mount files from linux and vice versa?
 Yes. Look up approaches.
- Why use a hard link? Avoid having another copy in another directory.
- Why use a symbolic link? Convenience.
- Average -
 - Average rotational latency why ½ a rotation time?
 - Average seek time is 1/3 of max seek time. Why?



Notes

- Project reports/slides: Due Th Dec 2, 2021
- Devp TA Demos: Dec 2-8 M-W. Sign-up for 15 min slot. [Videos]
- Research presentations: Dec 2-8 M-W [Videos]
 - Logistics & Details will be on Teams
- Peer reviews due Sat Dec 11, 2021
- Final: Comprehensive but mostly from second half
- Sec 001: Tu Dec 14, 6:20-8:20 PM
- Sec 801:
 - Local students with Sec 001 on-campus
 - Non-local: online during a 24 hr time window
- All SDC students (001, 801): at SDC as arranged.
 - Tu Dec 14, 4-8 PM (must stay until 6:20 PM)



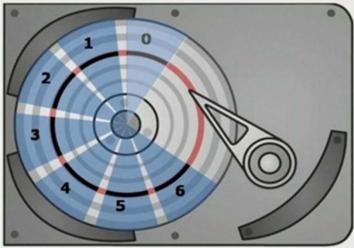
Hard Disk Performance

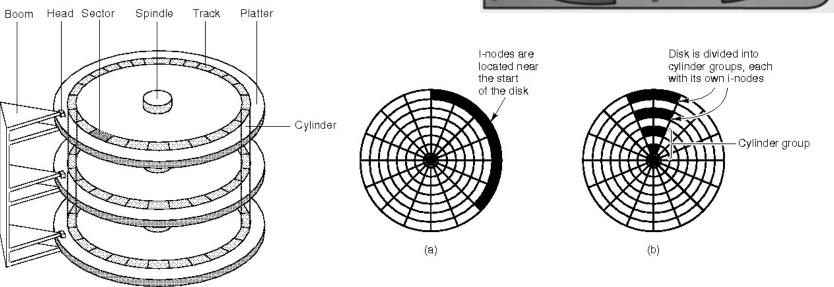
- Average I/O time = average access time + (amount to transfer / transfer rate) + controller overhead
 - Average access time = average seek time + average latency
- Example: Find expected I/O time to transfer a 4KB block on a 7200 RPM disk with a 5ms average seek time, 1Gb/sec transfer rate with a 0.1ms controller overhead.
 - Average access time = 5ms + 60/(7200*2) s = 5ms + 4.17ms
 - Transfer time = 4KB / 1Gb/s = 4x8K/G = 0.031 ms
 - Thus Average I/O time = 9.27ms + .031ms+0.1ms = 9.301ms

Strategy: memorize formula or understand how it works? Colorado State University

HDD addressing

- Physical: Drive, Cylinder, Head, sector
- Logical Block Addressing (LBA): blocks addressed by numbers.





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SSD Architecture

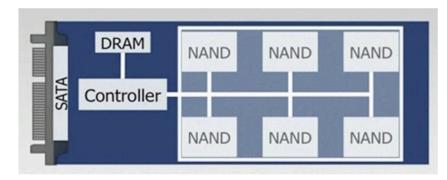
Controller

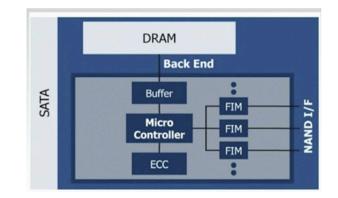
- Takes the raw data storage in the NAND flash and makes it look and act like hard disk drive
- Contains the micro controller, buffer, error correction, and flash interface modules

Micro Controller – a processor inside the controller that takes the incoming data and manipulates it

- Correcting errors
- Manages mapping
- Putting data into the flash or retrieving it from the flash

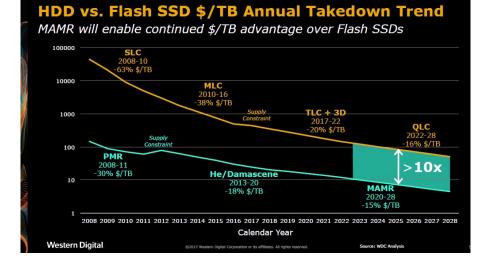
DRAM Cache – Reasonable amount of very low latency



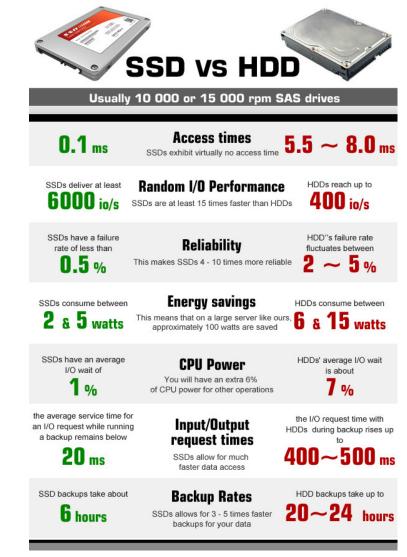


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SSD vs HDD



Enterprise SSD vs. Nearline / High-Cap HDDs - \$ per TB \$2,800 45x \$2,600 eSSDs vs. Nearline HDDs 40x \$2,400 ---- Enterprise SSDs - Nearline / High-Cap HDDs \$2,200 35x \$2,000 30x \$1,800 \$1.600 25x \$1,400 20x \$1.200 \$1,000 15x \$800 ő 10x \$600 \$400 5x \$200 \$0 0x 1013 2013 3013 2014 2014 2014 2015 2015 2015 2015 2015 2015 2016 2016 3016 4016 1017 4013 4017 1018 2018 3018 4018 1019 2019 2017 3017



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Source: IDC; TrendFocus; Wells Fargo Securities, LLC

HDD vs SSD

	HDD	SSD
	WD VelociRaptor	OCZ Vertex 3
Storage Capacity	600GB	120GB-360GB
Price for storage	48¢/ GB	2.08\$/GB x4
Seek Time/Rotational Speed	7ms/157 MB/s	
MTBF	1.4 million hours?	2 million hours?
Sequential Read/Write	1 MB/s	413.5/371.4 MB/s
Random Read	1 MB/s	68.8 MB/s
Random Write	1 MB/s	332.5 MB/s
IOPS	905	60,000 x60



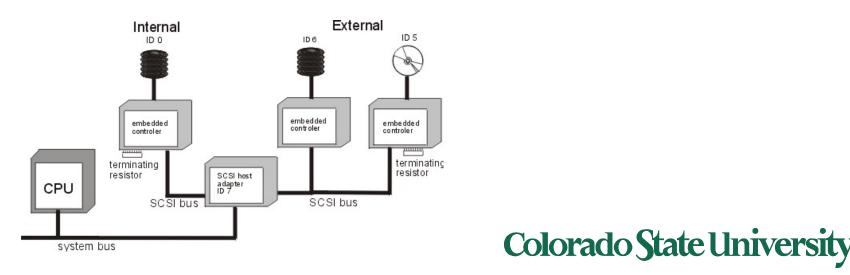
Magnetic Tape

- Was early secondary-storage medium (now tertiary)
 - Evolved from open spools to cartridges
- Relatively permanent and holds large quantities of data
 - Access time slow
 - Random access ~1000 times slower than disk
 - Once data under head, transfer rates comparable to disk
 - 140MB/sec and greater
- Mainly used for backup, storage of infrequentlyused data, transfer medium between systems
- Kept in spool and wound or rewound past readwrite head
- 200GB to 1.5TB typical storage Sony: New 330 TB



Disk Attachment: I/O busses

- Host-attached storage accessed through I/O ports talking to I/O busses
- SCSI itself is a bus, up to 16 devices on one cable, SCSI initiator (adapter) requests operation and SCSI targets (controller) perform tasks
 - Each target can have up to 8 logical units (disks attached to device controller)
- FC (fibre channel) is high-speed serial architecture
 - Can be switched fabric with 24-bit address space the basis of storage area networks (SANs) in which many hosts attach to many storage units



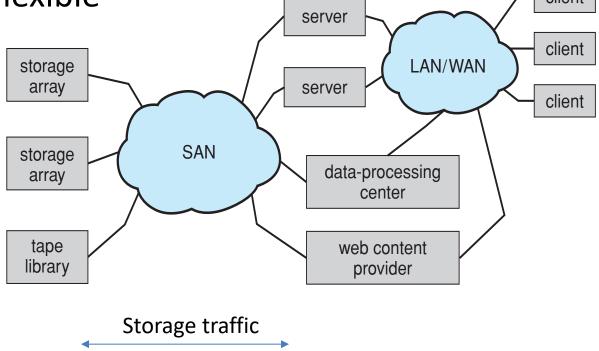
Storage Array

- Can just attach disks, or arrays of disks to an I/O port
- Storage Array has controller(s), provides features to attached host(s)
 - Ports to connect hosts to array
 - Memory, controlling software
 - A few to thousands of disks
 - RAID, hot spares, hot swap
 - Shared storage -> more efficiency



Storage Area Network

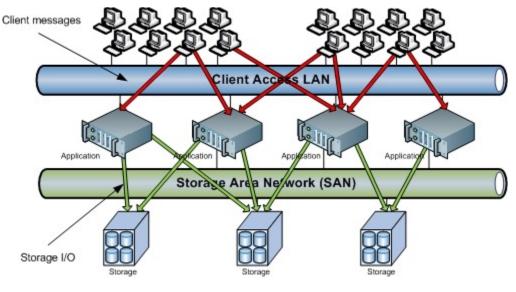
- Common in large storage environments
- Multiple hosts attached to multiple storage arrays
 flexible





Storage Area Network (Cont.)

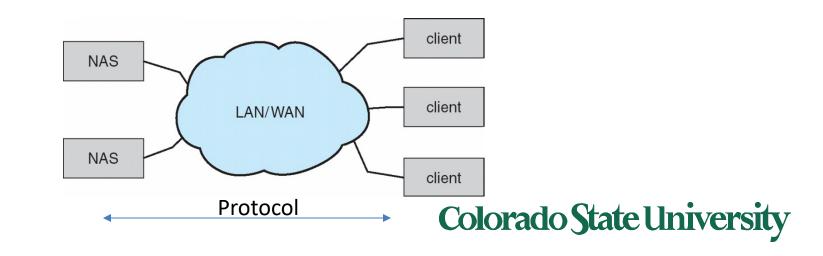
- SAN is one or more storage arrays
- Hosts also attach to the switches
- Storage made available from specific arrays to specific servers
- Easy to add or remove storage, add new host and allocate it storage
 - Over low-latency Fibre Channel fabric



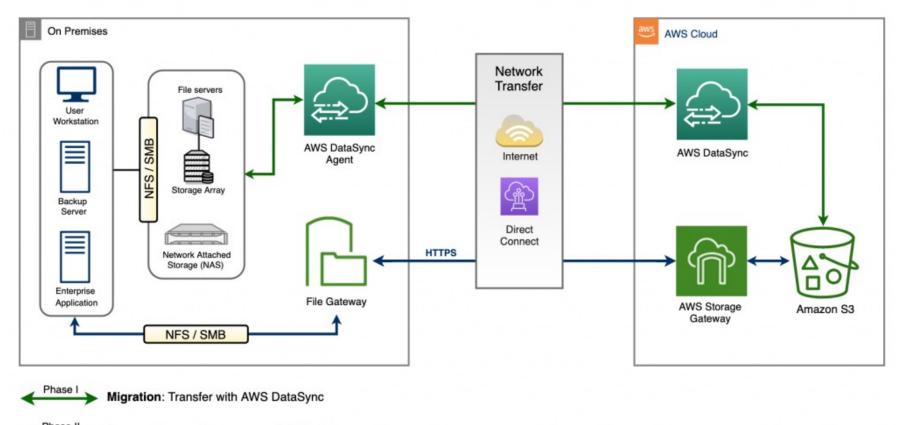


Network-Attached Storage

- Network-attached storage (NAS) is storage made available over a network rather than over a local connection (such as a bus)
 - Remotely attaching to file systems
- NFS and CIFS (windows) are common protocols
- Implemented via remote procedure calls (RPCs) between host and storage over typically TCP or UDP on IP network
- **iSCSI** protocol uses IP network to carry the SCSI protocol
 - Remotely attaching to devices (blocks)



AWS DataSync and Storage Gateway



Access: On premises access with File Gateway

Amazon S3 (Simple Storage Service)

Issues: Delay, security, availability, cost

https://aws.amazon.com/blogs/storage/from-on-premises-to-aws-hybrid-cloud-architecture-for network file share of State University

Disk Scheduling

- The operating system is responsible for using hardware efficiently — for the disk drives, this means having a fast access time and disk bandwidth
- Minimize seek time
- Seek time ≈∞ seek distance (between cylinders)
- Disk bandwidth is the total number of bytes transferred, divided by the total time between the first request for service and the completion of the last transfer



Disk Scheduling (Cont.)

- There are many sources of disk I/O request
 - OS
 - System processes
 - Users processes
- I/O request includes input or output mode, disk address, memory address, number of sectors to transfer
- OS maintains queue of requests, per disk or device
- Idle disk can immediately work on I/O request, busy disk means work must queue
 - Optimization algorithms only make sense when a queue exists



Disk Scheduling (Cont.)

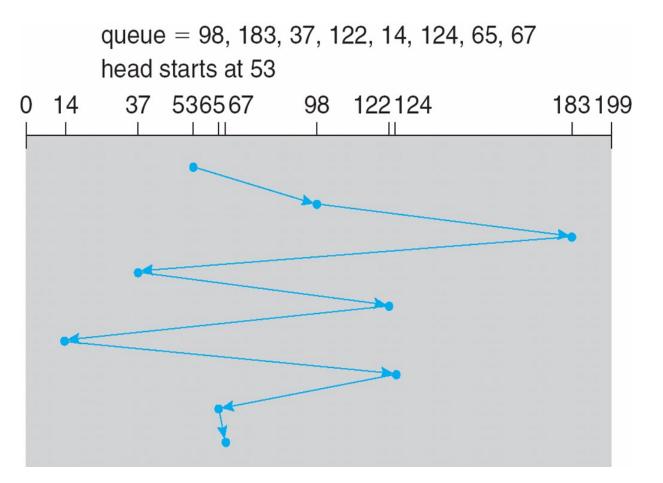
- Note that drive controllers have small buffers and can manage a queue of I/O requests (of varying "depth")
- Several algorithms exist to schedule the servicing of disk I/O requests
- The analysis is true for one or many platters
- We illustrate scheduling algorithms with a request queue (cylinders 0-199)

98, 183, 37, 122, 14, 124, 65, 67 Head pointer 53 (head is at cylinder 53)

Similar problems: limousine pickup/dropoff, elevator etc.

FCFS (First come first served)

Illustration shows total head movement. Cylinder 0 is outermost



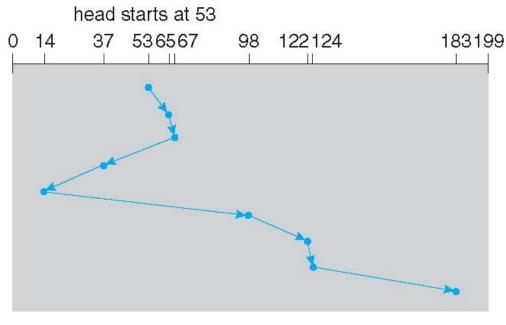
Total seek time = (98-53) += 640 cylinders Colorado State University

SSTF Shortest Seek Time First

- Shortest Seek Time First selects the request with the minimum seek time from the current head position
- SSTF scheduling is a form of SJF scheduling; may cause starvation of some requests

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total head movement of 236 cylinders



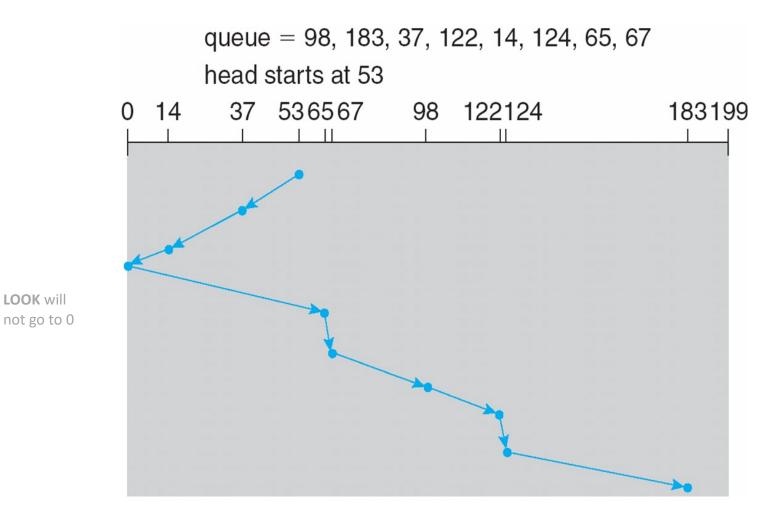
queue = 98, 183, 37, 122, 14, 124, 65, 67

SCAN

- The disk arm starts at one end of the disk, and moves toward the other end, servicing requests until it gets to the other end of the disk, where the head movement is reversed, and servicing continues.
- SCAN algorithm Sometimes called the elevator algorithm
- But note that if requests are uniformly dense, largest density at the other end of disk and those wait the longest
- Variation: Look: may not go to the very edge



SCAN (Cont.)



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Total 53+ 183= 236 cylinders

22

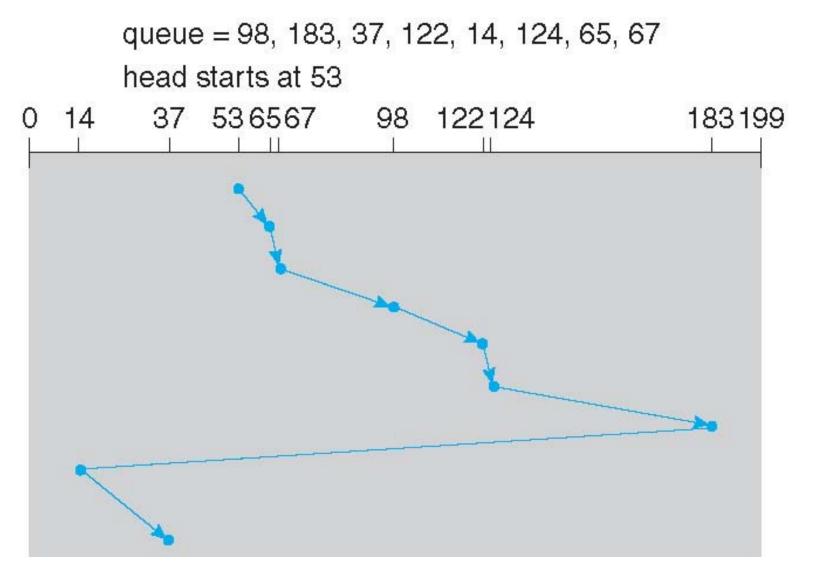
LOOK will

C-LOOK

- LOOK a version of SCAN, C-LOOK a version of C-SCAN
- Arm only goes as far as the last request in each direction, then reverses direction immediately, without first going all the way to the end of the disk
- Total number of cylinders?



C-LOOK (Cont.)



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Disk Management

- Low-level formatting, or physical formatting Dividing a disk into sectors that the disk controller can read and write
 - Each sector can hold header information (sector number), plus data, plus error correction code (ECC)
 - Usually 512 bytes of data but can be selectable
- To use a disk to hold files, the operating system still needs to record its own data structures on the disk
 - Partition the disk into one or more groups of cylinders, each treated as a logical disk
 - Logical formatting or "making a file system"
 - To increase efficiency most file systems group blocks into clusters
 - File I/O done in clusters
- Raw disk access for apps that want to do their own block management, keep OS out of the way (databases for example)



Disk Management (Cont.)

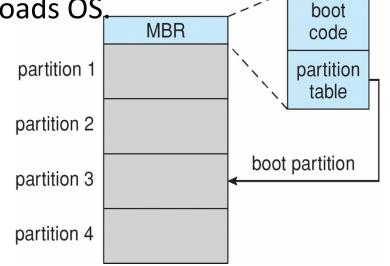
- Boot block initializes system
 - The tiny bootstrap code is stored in ROM
 - Bootstrap loader program stored in boot blocks of boot partition which loads OS.
- Methods such as sector sparing used to handle bad blocks



Booting from a Disk in Windows

- MBR: Master boot record: identifies boot partition
- Kernel loaded from boot partition
- Boot block initializes system
 - The tiny bootstrap code is stored in ROM
 - Full Bootstrap loader program identified in boot blocks of boot partition which loads OS.
- Methods such as sector sparing used to handle bad blocks

Boot disk: has a boot partition



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Reliability & RAIDs

Various sources

Reliability

- Storage is inherently unreliable. How can it be made more reliable?
- Redundancy
 - Complete mirrors of data: 2, 3 or more copies.
 - Use a good copy when there is failure,
 - Additional bits: Use parity bit/bits.
 - Use parity to reconstruct corrupted data
 - Rollback and retry
 - Go back to previously saved known good state and recompute.



RAID Structure

- RAID redundant array of inexpensive/independent disks. Multiple disk drives provides
 - Higher reliability, repair capability
 - Higher performance /storage capacity
 - A combination
- Increases the mean time to failure
- Mean time to repair exposure time when another failure could cause data loss
- Mean time to data loss based on above factors



RAID Techniques

- **Striping** uses multiple disks in parallel by splitting data: higher performance, no redundancy (ex. RAID 0)
- Mirroring keeps duplicate of each disk: higher reliability (ex. RAID 1)
- Block parity: One Disk hold parity block for other disks. A failed disk can be rebuilt using parity. Wear leveling if interleaved (RAID 5, double parity RAID 6).
- Ideas that did not work: Bit or byte level level striping (RAID 2, 3) Bit level Coding theory (RAID 2), dedicated parity disk (RAID 4).
- Nested Combinations:
 - RAID 01: Mirror RAID 0
 - RAID 10: Multiple RAID 1, striping
 - RAID 50: Multiple RAID 5, striping
 - others



RAID

- Replicate data for availability
 - RAID 0: no replication
 - RAID 1: mirror data across two or more disks
 - Google File System replicated its data on three disks, spread across multiple racks
 - RAID 5: split data across disks, with redundancy to recover from a single disk failure
 - RAID 6: RAID 5, with extra redundancy to recover from two disk failures

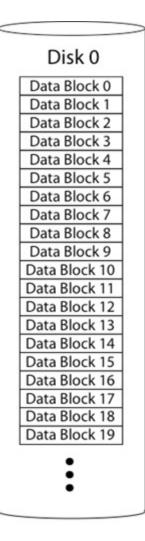


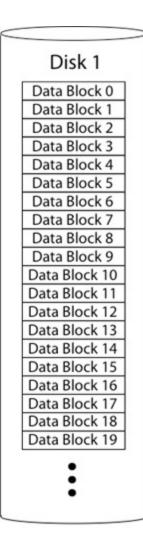
RAID 1: Mirroring

- Replicate writes to both disks
- Reads can go to either disk
- If they fail independently, consider disk with 100,000 hour *mean time* to failure and 10 hour *mean time* to repair
- One disk fails wile other is being repaired: data loss
 - probability that two will fail within 10 hours =

 $(2x10) /100,000^{2}$

— Mean time to data loss is 100,000²/(2x10) = 500x10⁶ hours, or 57,000 years!





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Parity

- Data blocks: Block1, block2, block3,
- Parity block: Block1 xor block2 xor block3 ...

10001101	block1
01101100	block2
11000110	block3
00100111	parity block (ensures even number of 1s)

Can reconstruct any missing block from the others

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Parity Exercise

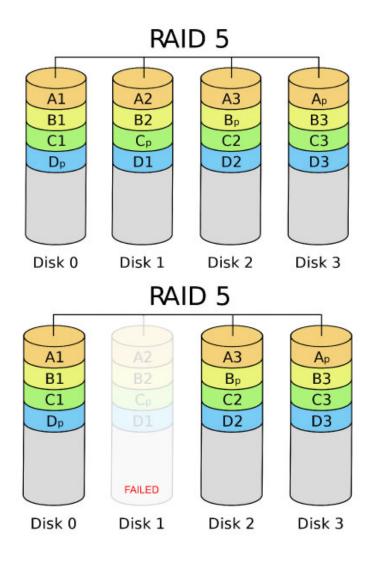
• Parity block: Block1 xor block2 xor block3 ...

10001101	block1
???????	block2 (bad)
11000110	block3
00100111	parity block (ensures even number of 1s)

• Can you reconstruct the bad block using the others?



RAID 5: Rotating Parity



Time to rebuild depends on disk capacity and data transfer rate



Read Errors and RAID recovery

- Example: RAID 5
 - Each bit has 10⁻¹⁵ probability of being bad.
 - 10 one-TB disks, and 1 fails
 - Read remaining disks to reconstruct missing data
- Probability of an error in reading 9 TB disks during recovery
 - = 10⁻¹⁵*(9 disks * 8 bits * 10¹² bytes/disk)

= 7.2%. Thus recovery probability = 92.8%

- Even better:
 - RAID-6: two redundant disk blocks
 - Can work even in presence of one bad disk
 - Scrubbing: read disk sectors in background to find and fix latent errors



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Big Data: Hadoop HDFS and mapreduce

Various sources

Hadoop: Distributed Framework for Big Data

Big Data attributes:

- Large volume: TB -> PB varies with Kryder's law: disk density doubles / 13 months
- Geographically Distributed: minimize data movement
- Needs: reliability, analytic approaches

History:

- Google file system 2003 and Map Reduce 2004 programming lang
- Hadoop to support distribution for the Yahoo search engine project '05, given to Apache Software Foundation '06
- Hadoop ecosystem evolves with Yarn '13 resource management, Pig '10 scripting, Spark '14 distributed computing engine. etc.

• MapReduce: Simplified Data Processing on Large Clusters. by Jeffrey Dean and Sanjay Ghemawat (2004)

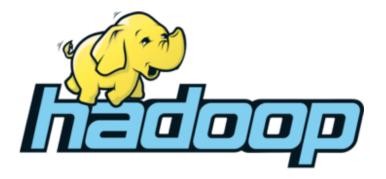


[•] The Google file system by Sanjay Ghemawat, Howard Gobioff, and Shun-Tak Leung (2003)

Hadoop: Distributed Framework for Big Data

Recent development.

- Big data: multi-terabyte or more data for an app
- Distributed file system
 - Reliability through replication (Fault tolerance)
- Distributed execution
 - Parallel execution for higher performance





Hadoop: Core components

- Hadoop (originally): MapReduce + HDFS
- MapReduce: A programming framework for processing parallelizable problems across huge datasets using a large number of commodity machines.
- HDFS: A distributed file system designed to efficiently allocate data across multiple commodity machines, and provide self-healing functions when some of them go down

• Commodity machines: lower performance per machine, lower cost, perhaps lower reliability compared with special high performance machines.



Challenges in Distributed Big Data

Common Challenges in Distributed Systems

- Node Failure: Individual computer nodes may overheat, crash, have hard drive failures, or run out of memory or disk space.
- Network issues: Congestion/delays (large data volumes), Communication Failures.
- Bad data: Data may be corrupted, or maliciously or improperly transmitted.
- Other issues: Multiple versions of client software may use slightly different protocols from one another.
- Security



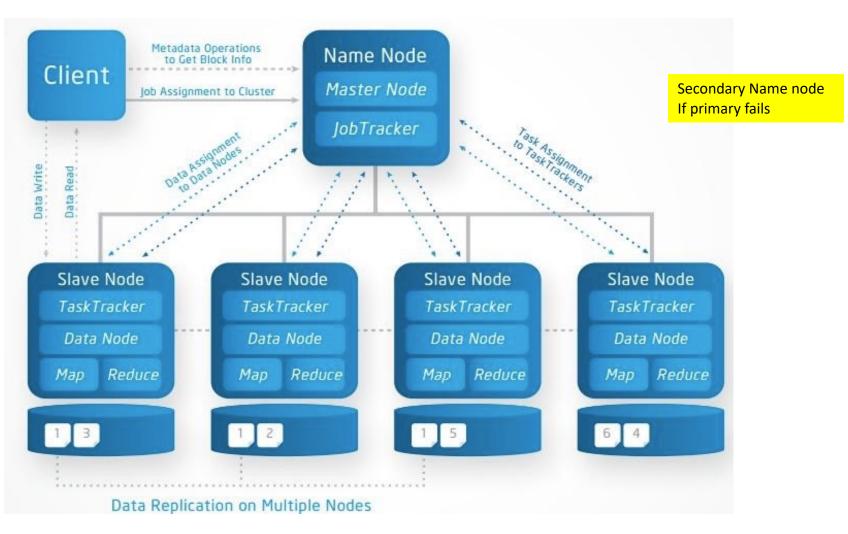
HDFS Architecture

Hadoop Distributed File System (HDFS):

- HDFS Block size: 64-128 MB ext4: 4KB
- HDFS file size: "Big"
- Single HDFS FS cluster can span many nodes possibly geographically distributed. datacenters-racks-blades
- Node: system with CPU and memory
- Metadata (corresponding to superblocks, Inodes)
- Name Node: metadata giving where blocks are physically located
- Data (files blocks)
- Data Nodes: hold blocks of files (files are distributed)



HDFS Architecture



http://a4academics.com/images/hadoop/Hadoop-Architecture-Read-Write.jpg

