The Google File System

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(These slides modified from Alex Moshchuk, University of Washington – used during Google lecture series.)



Outline

- Filesystems Overview
- GFS (Google File System)
 - Motivations
 - □ Architecture
 - Algorithms
- HDFS (Hadoop File System)

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Filesystems Overview

- Permanently stores data
- Usually layered on top of a lower-level physical storage medium
- Divided into logical units called "files"
 - ☐ Addressable by a *filename* ("foo.txt")
 - Usually supports hierarchical nesting (directories)
- A file path = relative (or absolute) directory + file name
 - □/dir1/dir2/foo.txt



Distributed Filesystems

- Support access to files on remote servers
- *Must* support concurrency
 - Make varying guarantees about locking, who "wins" with concurrent writes, etc...
 - Must gracefully handle dropped connections
- Can offer support for replication and local caching
- Different implementations sit in different places on complexity/feature scale



Motivation

- Google needed a good distributed file system
 - Redundant storage of massive amounts of data on cheap and unreliable computers
- Why not use an existing file system?
 - □ Google's problems are different from anyone else's
 - Different workload and design priorities
 - □ GFS is designed for Google apps and workloads
 - □ Google apps are designed for GFS



Assumptions

- High component failure rates
 - Inexpensive commodity components fail all the time
- "Modest" number of HUGE files
 - □ Just a few million
 - □ Each is 100MB or larger; multi-GB files typical
- Files are write-once, mostly appended to
 - Perhaps concurrently
- Large streaming reads
- High sustained throughput favored over low latency

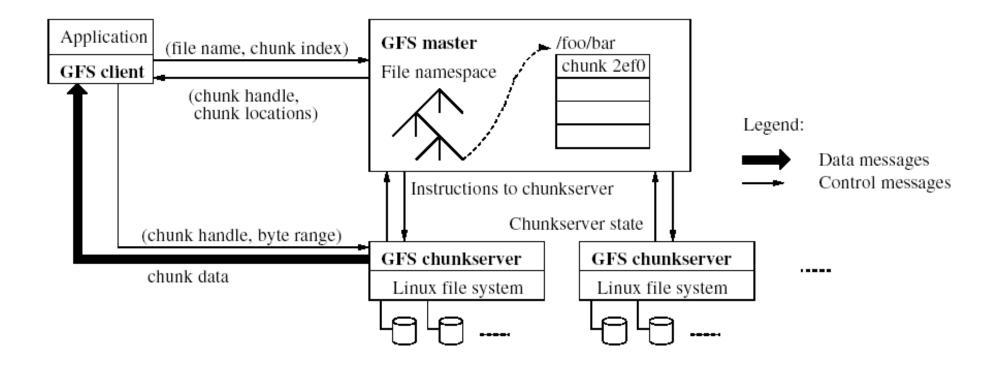
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GFS Design Decisions

- Files stored as chunks
 - ☐ Fixed size (64MB)
- Reliability through replication
 - □ Each chunk replicated across 3+ *chunkservers*
- Single master to coordinate access, keep metadata
 - □ Simple centralized management
- No data caching
 - ☐ Little benefit due to large data sets, streaming reads
- Familiar interface, but customize the API
 - □ Simplify the problem; focus on Google apps
 - Add snapshot and record append operations

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GFS Architecture



...Can anyone see a potential weakness in this design?



Single master

- Problem:
 - □ Single point of failure
 - □ Scalability bottleneck
- GFS solutions:
 - □ Shadow masters
 - Minimize master involvement
 - never move data through it, use only for metadata
 - □ and cache metadata at clients
 - large chunk size
 - master delegates authority to primary replicas in data mutations (chunk leases)
- Simple, and good enough for Google's concerns



Metadata

- Global metadata is stored on the master
 - ☐ File and chunk namespaces
 - Mapping from files to chunks
 - □ Locations of each chunk's replicas
- All in memory (64 bytes / chunk)
 - □ Fast
 - □ Easily accessible



Metadata

- Master has an operation log for persistent logging of critical metadata updates
 - □ Persistent on local disk
 - □ Replicated
 - □ Checkpoints for faster recovery



Master's Responsibilities

- Metadata storage
- Namespace management/locking
- Periodic communication with chunkservers
 - □ give instructions, collect state, track cluster health
- Chunk creation, re-replication, rebalancing
 - □ balance space utilization and access speed
 - spread replicas across racks to reduce correlated failures
 - re-replicate data if redundancy falls below threshold
 - rebalance data to smooth out storage and request load



Master's Responsibilities

- Garbage Collection
 - □ simpler, more reliable than traditional file delete
 - master logs the deletion, renames the file to a hidden name
 - □ lazily garbage collects hidden files
- Stale replica deletion
 - □ detect "stale" replicas using chunk version numbers

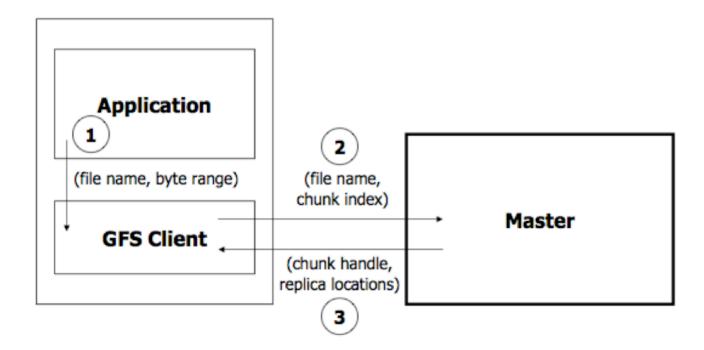


Mutations

- Mutation = write or record append
 - Must be done for all replicas
- Goal: minimize master involvement
- Lease mechanism:
 - ■Master picks one replica as primary; gives it a "lease" for mutations
- Data flow decoupled from control flow



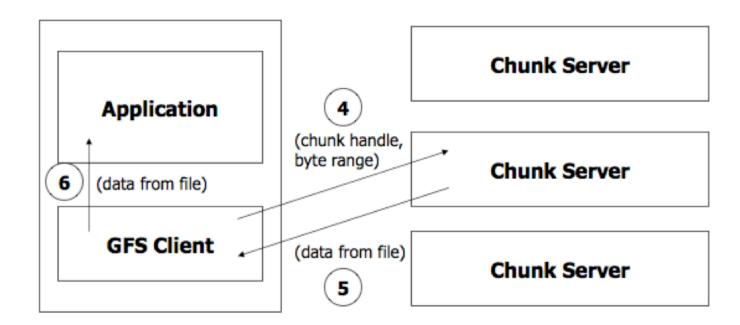
- Application originates the read request
- 2. GFS client translates request and sends it to master
- Master responds with chunk handle and replica locations





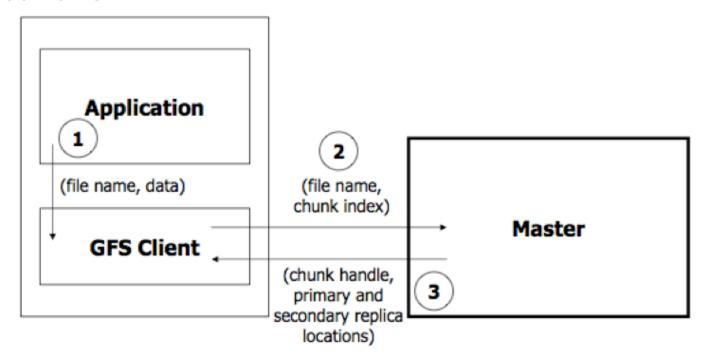
Read Algorithm

- 4. Client picks a location and sends the request
- 5. Chunkserver sends requested data to the client
- 6. Client forwards the data to the application





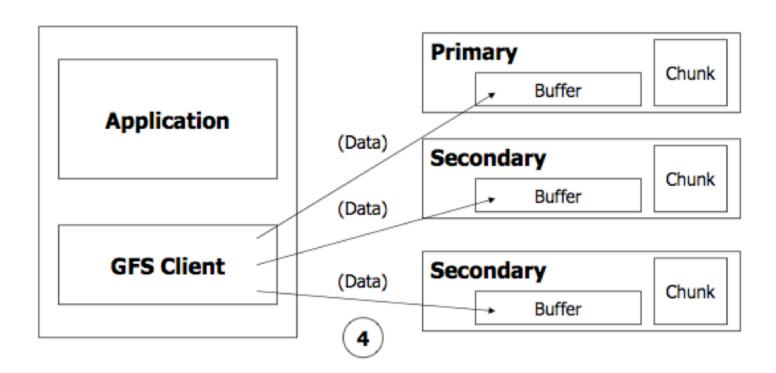
- 1. Application originates the request
- 2. GFS client translates request and sends it to master
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Write Algorithm

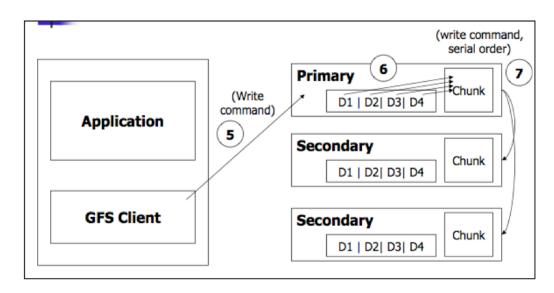
4. Client pushes write data to all locations. Data is stored in chunkserver's internal buffers





Write Algorithm

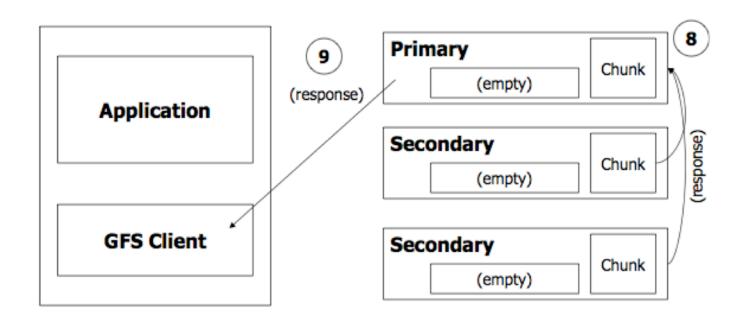
- Client sends write command to primary
- 6. Primary determines serial order for data instances in its buffer and writes the instances in that order to the chunk
- Primary sends the serial order to the secondaries and tells them to perform the write





Write Algorithm

- Secondaries respond back to primary
- 9. Primary responds back to the client





Atomic Record Append

- GFS appends it to the file atomically at least once
 - ☐ GFS *picks* the offset
 - □ Works for concurrent writers
- Used heavily by Google apps
 - e.g., for files that serve as multiple-producer/singleconsumer queues
 - □ Merge results from multiple machines into one file

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Record Append Algorithm

- Same as write, but no offset and...
 - 1. Client pushes write data to all locations
 - Primary checks if record fits in specified chunk
 - 3. If the record does not fit:
 - Pads the chunk
 - 2. Tells secondary to do the same
 - 3. Informs client and has the client retry
 - 4. If record fits, then the primary:
 - Appends the record
 - Tells secondaries to do the same
 - 3. Receives responses and responds to the client

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Relaxed Consistency Model

- Consistent = all replicas have the same value
- Defined = replica reflects the mutation, consistent
- Some properties:
 - concurrent writes leave region consistent, but possibly undefined
 - ☐ failed writes leave the region *inconsistent*
- Some work has moved into the applications:
 - □ e.g., self-validating, self-identifying records
- "Simple, efficient"
 - ☐ Google apps can live with it

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Fault Tolerance

- High availability
 - □ Fast recovery
 - master and chunkservers restartable in a few seconds
 - □ Chunk replication
 - default: 3 replicas.
 - □ Shadow masters
- Data integrity
 - ☐ Checksum every 64KB block in each chunk

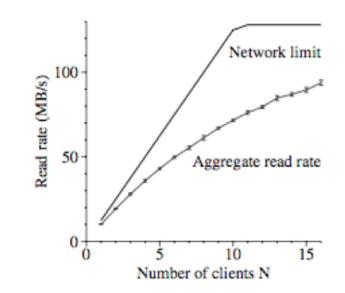


Performance Test

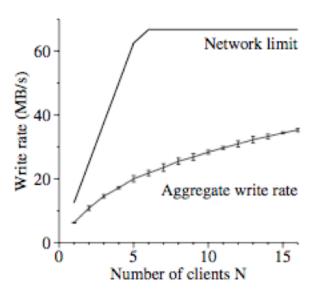
- Cluster setup:
 - □1 master
 - □ 16 chunkservers
 - □16 clients
- Server machines connected to central switch by 100 Mbps Ethernet
- Switches connected with 1 Gbps link



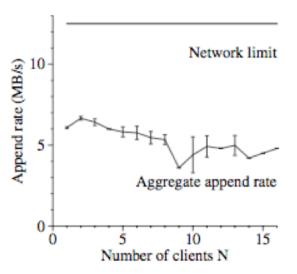
Reads



Write



Record Append



- •1 client:
 - •10 MB/s, 80% limit
- •16 clients:
 - •6 MB/s, 75% limit

- •1 client:
 - •6.3 MB/s, 50% limit
- •16 clients:
 - •35 MB/s, 50% limit
 - •2.2 MB/s per client

- •1 client:
 - •6 MB/s
- •16 clients:
 - •4.8 MB/s per client

Performance

Cluster	А	В
Chunkservers	342	227
Available disk space	72 TB	180 TB
Used disk space	55 TB	155 TB
Number of Files	735 k	737 k
Number of Dead files	22 k	232 k
Number of Chunks	992 k	1550 k
Metadata at chunkservers	13 GB	21 GB
Metadata at master	48 MB	60 MB

Cluster	A	В
Read rate (last minute)	583 MB/s	380 MB/s
Read rate (last hour)	562 MB/s	384 MB/s
Read rate (since restart)	589 MB/s	49 MB/s
Write rate (last minute)	1 MB/s	101 MB/s
Write rate (last hour)	2 MB/s	117 MB/s
Write rate (since restart)	25 MB/s	13 MB/s
Master ops (last minute)	325 Ops/s	533 Ops/s
Master ops (last hour)	381 Ops/s	518 Ops/s
Master ops (since restart)	202 Ops/s	347 Ops/s



Deployment in Google

- Many GFS clusters
- Hundreds/thousands of storage nodes each
- Managing petabytes of data
- GFS is under BigTable, etc.



Conclusion

- GFS demonstrates how to support large-scale processing workloads on commodity hardware
 - □ design to tolerate frequent component failures
 - optimize for huge files that are mostly appended and read
 - □ feel free to relax and extend FS interface as required
 - ☐ go for simple solutions (e.g., single master)
- GFS has met Google's storage needs, therefore good enough for them.

Hadoop File System



HDFS Design Assumptions

- Single machines tend to fail
 - □ Hard disk, power supply, ...
- More machines = increased failure probability
- Data doesn't fit on a single node
- Desired:
 - Commodity hardware
 - ■Built-in backup and failover

... Does this look familiar?



Namenode and Datanodes

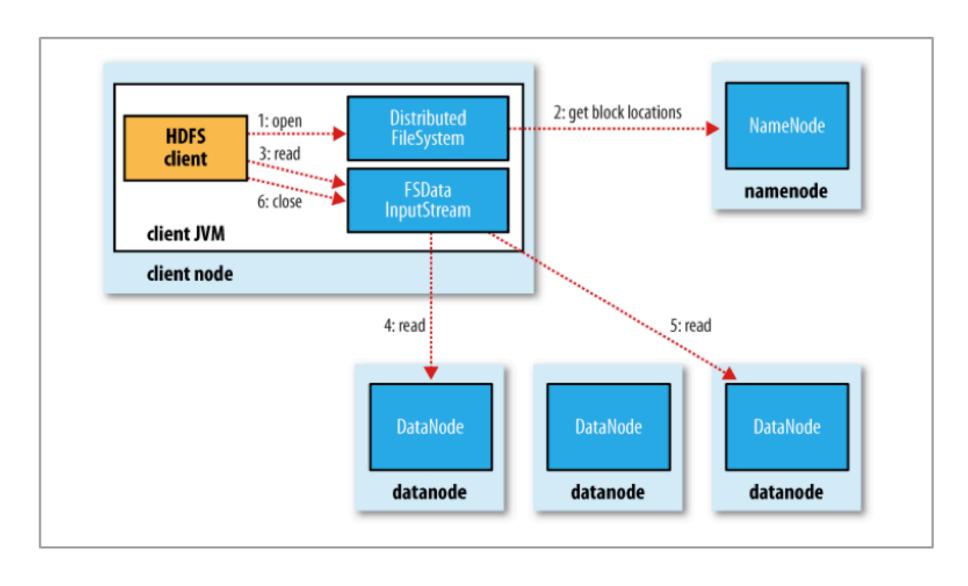
- Namenode (Master)
 - Metadata:
 - Where file blocks are stored (namespace image)
 - Edit (Operation) log
 - □ Secondary namenode (Shadow master)
- Datanode (Chunkserver)
 - ☐ Stores and retrieves blocks
 - ...by client or namenode.
 - Reports to namenode with list of blocks they are storing

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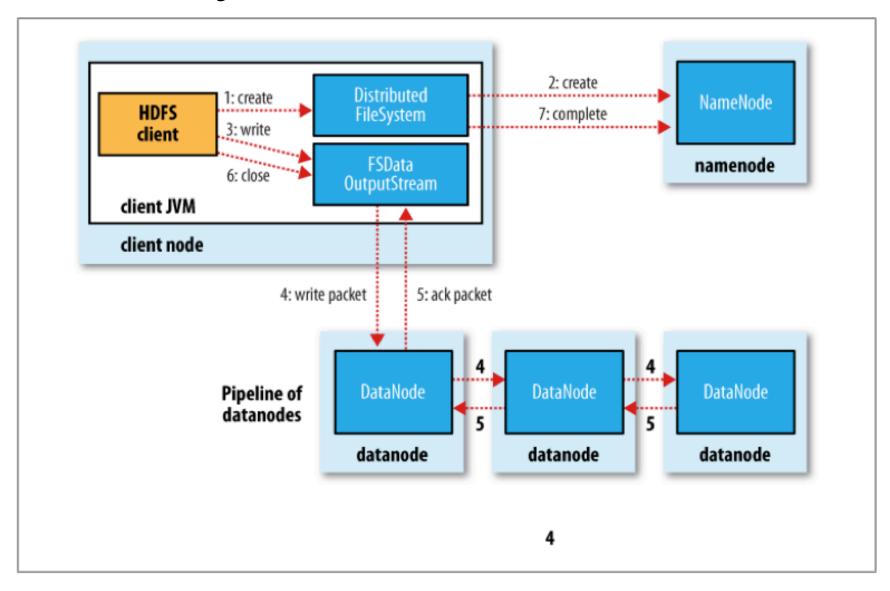
Noticeable Differences from GFS

- Only single-writers per file.
 - No record append operation.
- Open source
 - □ Provides many interfaces and libraries for different file systems.
 - S3, KFS, etc.
 - Thrift (C++, Python, ...), *libhdfs* (C), FUSE

Anatomy of a File Read



Anatomy of a File Write



```
// Print the contents of the HDFS file pathName to stdout.
public static void PrintHDFSFile(Context context, String pathName)
        throws IOException {
   // Load the HDFS library.
    Configuration conf = context.getConfiguration();
    FileSystem fs = FileSystem.get(conf);
   // Load the file input stream.
    Path hdfsPath = new Path(pathName);
    FSDataInputStream in = fs.open(hdfsPath);
    String line = null;
    while ((line = in.readUTF()) != null) {
        System.out.println(line);
    in.close();
```



Additional Topics

- Replica placements:
 - □ Different node, rack, and center
- Coherency model:
 - Describes data visibility
 - Current block being written may not be visible to other readers
- Web demo



Questions?

- Additional slides taken from:
 - http://www.cs.rochester.edu/~naushad/survey/nz-google-file-system.ppt.pdf