

Google File System (GFS)

MapReduce assumptions

- Commodity hardware
 - Economies of scale!
 - Commodity networking with less bisection bandwidth
 - Commodity storage (hard disks) is cheap

Failures are common

 Replicated, distributed file system for data storage

Fault tolerance

- If a task crashes:
 - Retry on another node
 - Why this is okay?
 - If the same task repeatedly fails, end the job

Fault tolerance

- If a task crashes:
 - Retry on another node
 - Why this is okay?
 - If the same task repeatedly fails, end the job
- If a node crashes:
 - Relaunch its current tasks on another node
 - What about task inputs?

Google file system (GFS)

- Goal: a global (distributed) file system that stores data across many machines
 - Need to handle 100's TBs

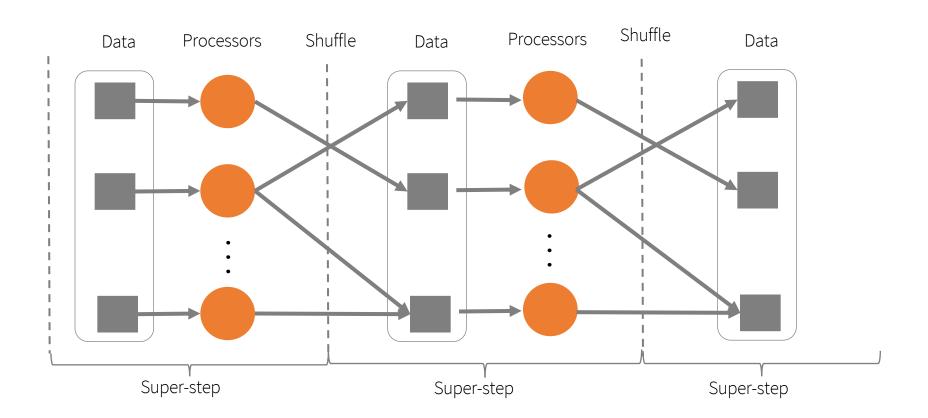
- Google published details in 2003
- Open source implementation:
 - Hadoop Distributed File System (HDFS)



Workload-driven design

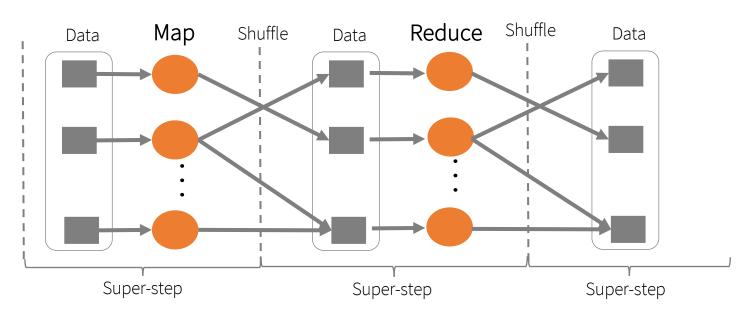
- MapReduce workload characteristics
 - Huge files (GBs)
 - Almost all writes are appends
 - Concurrent appends common
 - High throughput is valuable
 - Low latency is not

Example workloads: Bulk Synchronous Processing (BSP)



*Leslie G. Valiant, A bridging model for parallel computation, Communications of the ACM, Volume 33 Issue 8, Aug. 1990

MapReduce as a BSP system



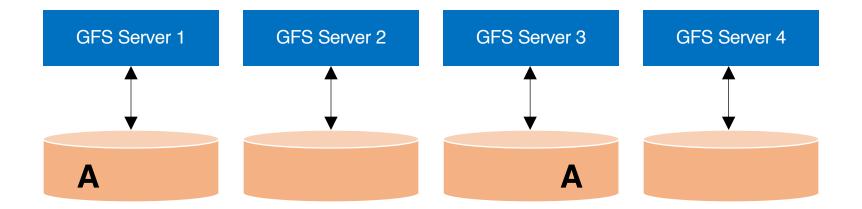
- Read entire dataset, do computation over it
 - Batch processing
- Producer/consumer: many producers append work to file concurrently; one consumer reads and does work

Workload-driven design

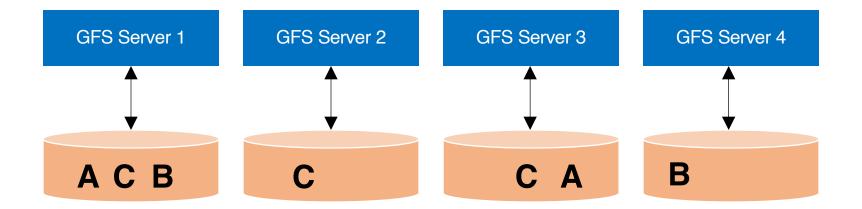
 Build a global (distributed) file system that incorporates all these application properties

- Only supports features required by applications
- Avoid difficult local file system features, e.g.:
 - rename dir
 - links

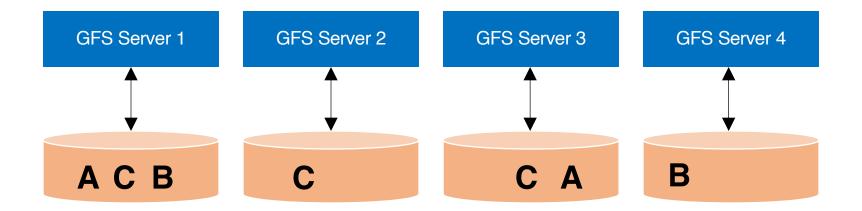
Replication



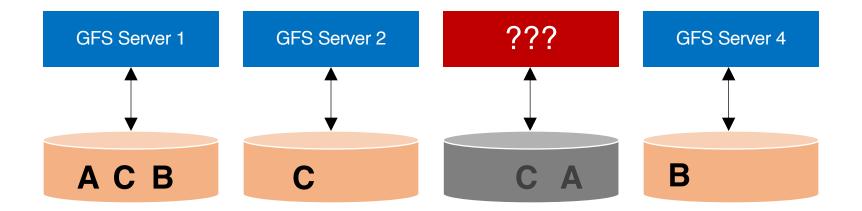
Replication

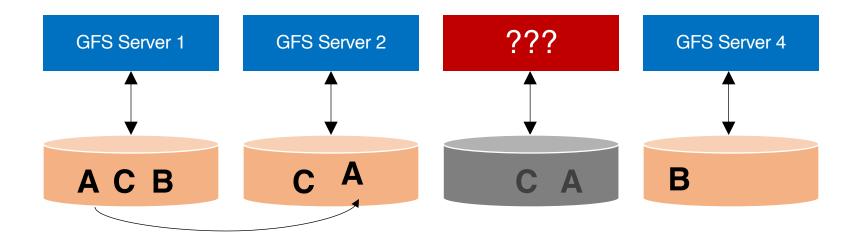


Resilience against failures

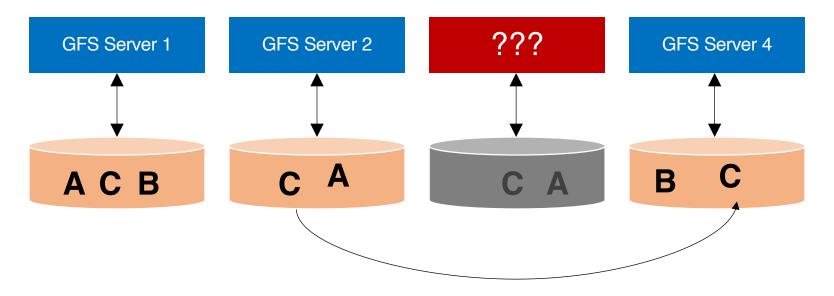


Resilience against failures

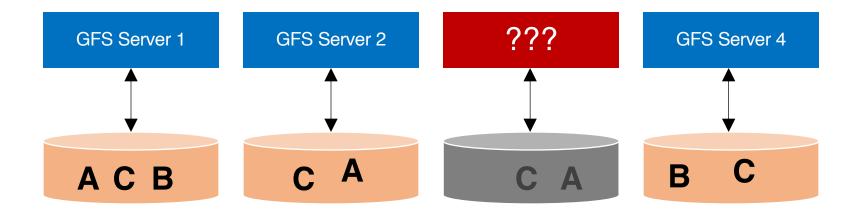




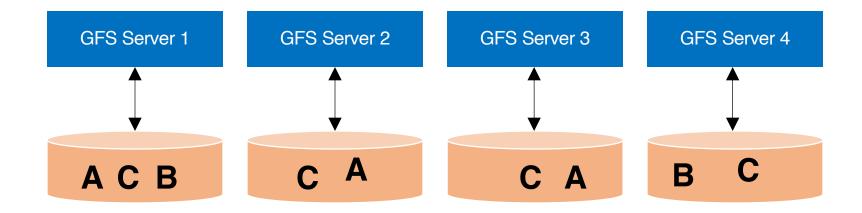
Replicating A to maintain a replication factor of 2



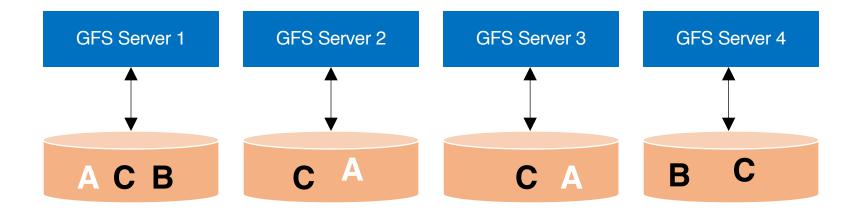
Replicating C to maintain a replication factor of 3

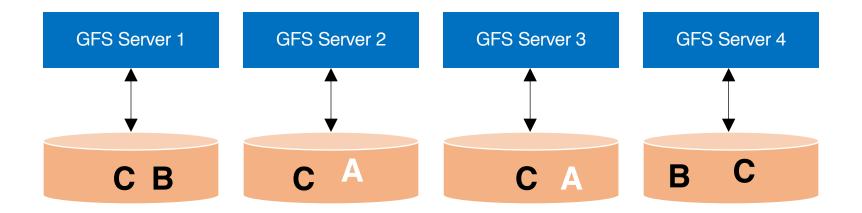


Machine may be dead forever, or it may come back



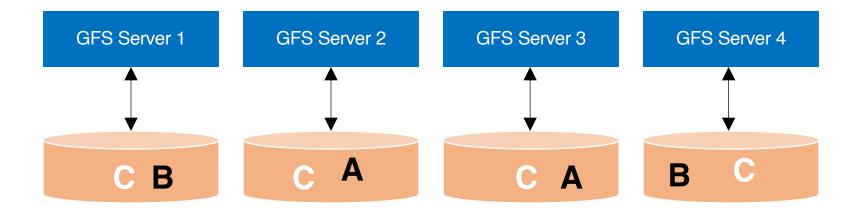
Machine may be dead forever, or it may come back

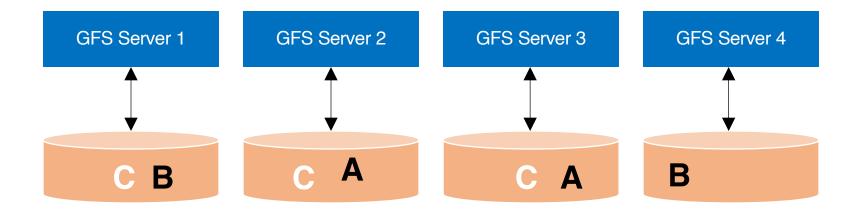




Data Rebalancing

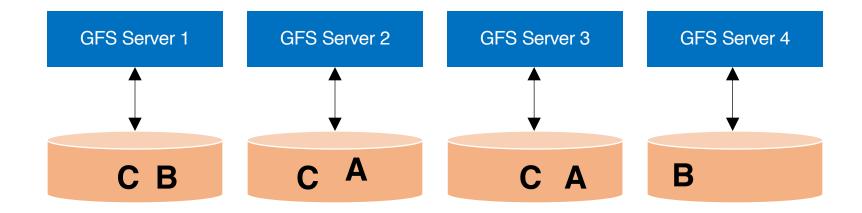
Deleting one A to maintain a replication factor of 2





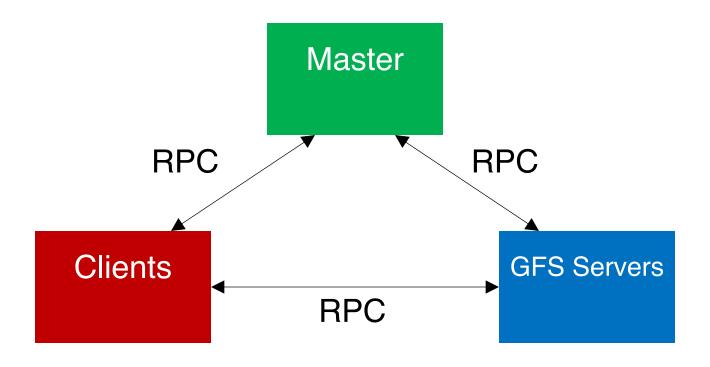
Data Rebalancing

Deleting one C to maintain a replication factor of 3

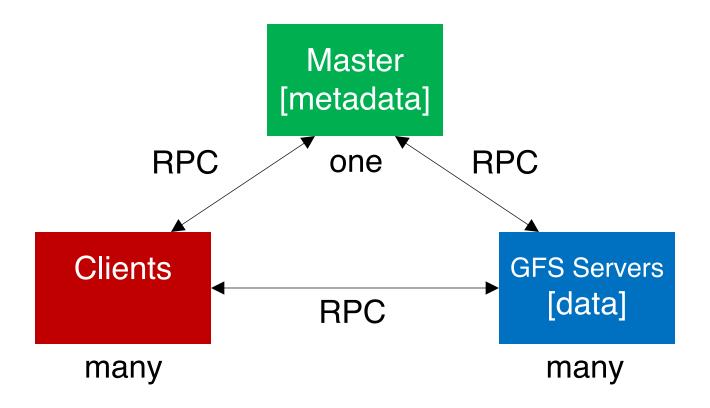


Question: how to maintain a global view of all data distributed across machines?

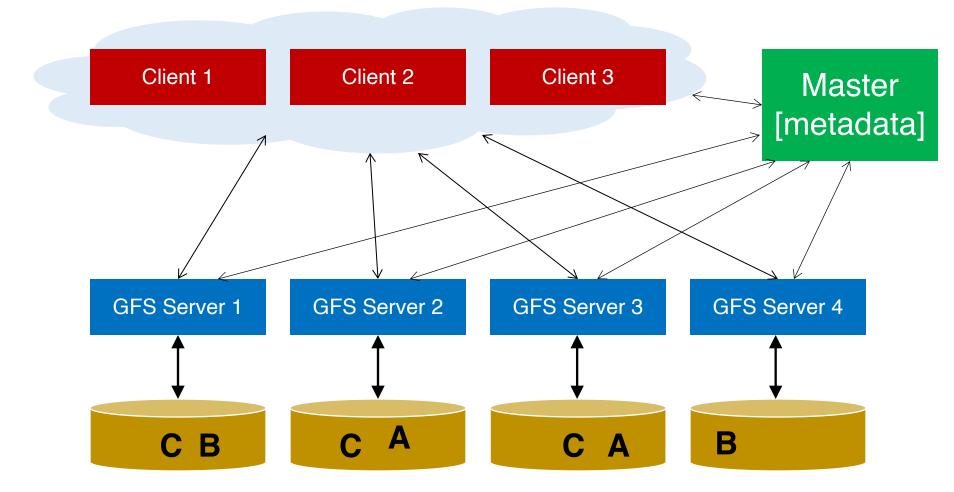
GFS architecture



GFS architecture



GFS architecture



Data chunks

• Break large GFS files into coarse-grained data chunks (e.g., 64MB)

 GFS servers store physical data chunks in local Linux file system

 Centralized master keeps track of mapping between logical and physical chunks

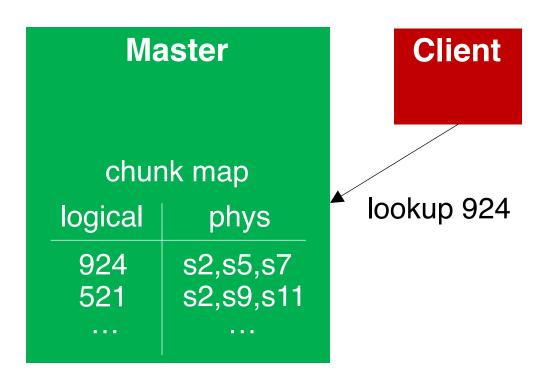
Chunk map

chunk map logical phys 924 s2,s5,s7 521 s2,s9,s11 ...

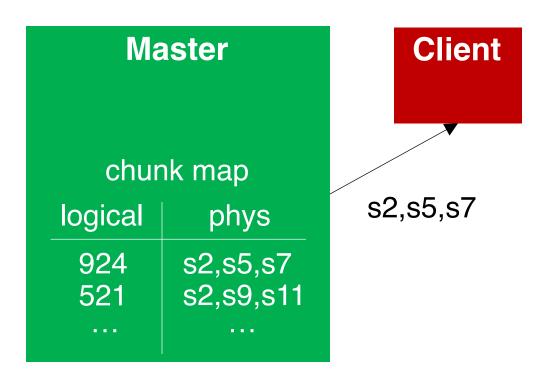
GFS server s2

Master	
chunk map	
logical	phys
924 521 	s2,s5,s7 s2,s9,s11

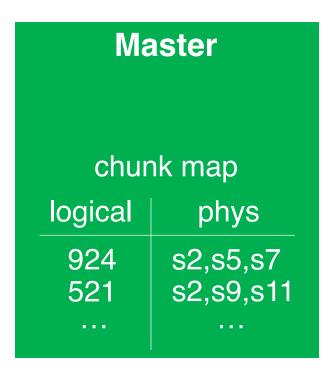








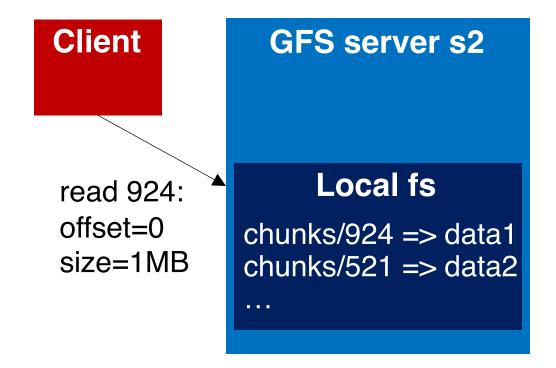




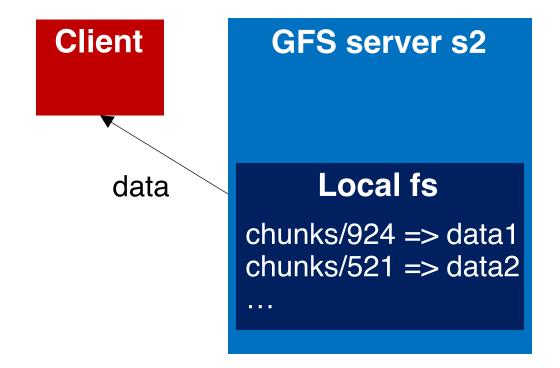




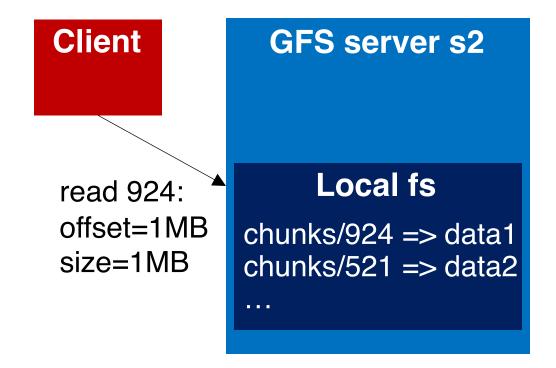




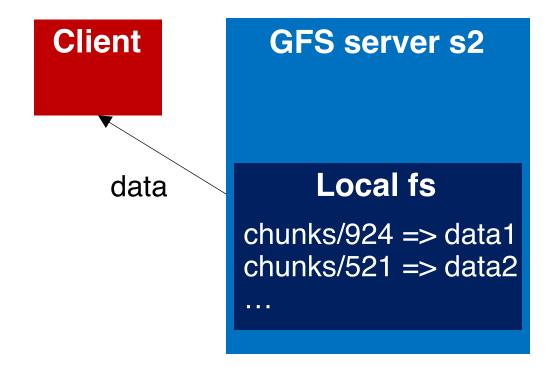




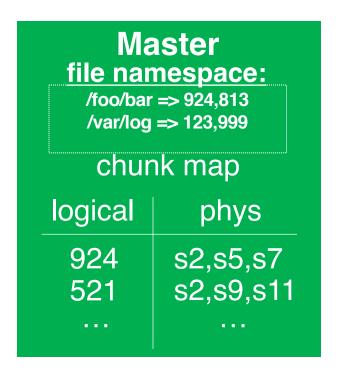








File namespace

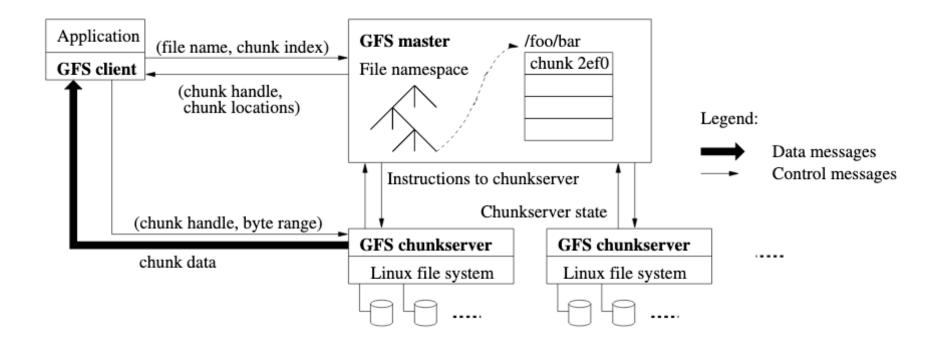




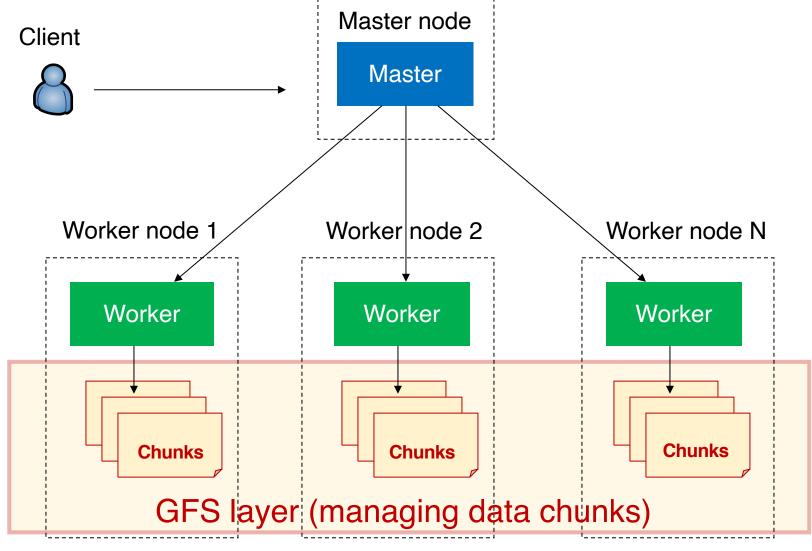


path names mapped to logical names

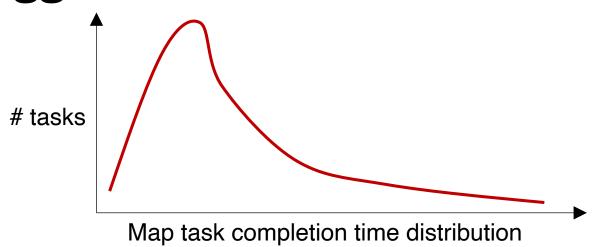
GFS architecture (original paper)



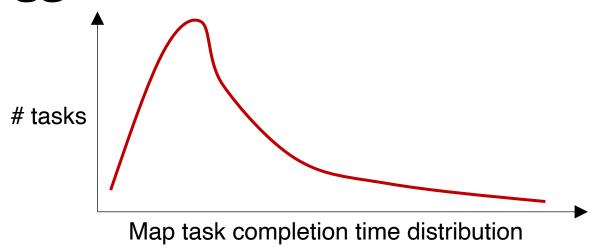
MapReduce+GFS: Put everything together



Stragglers



Stragglers



- Tail latency means some workers (always) finish late
- Q: How can MapReduce work around this?
 - Hint: its approach to fault-tolerance provides the right tool

Resilience against stragglers

- If a task is going slowly (i.e., straggler):
 - Launch second copy of task on another node
 - Take the output of whichever finishes first

More design

Master failure

Locality

Task granularity

GFS usage at Google

- 200+ clusters
- Many clusters of 1000s of machines
- Pools of 1000s of clients
- 4+ PB filesystems
- 40 GB/s read/write load
 - In the presence of frequent hardware failures

* Jeff Dean, LADIS 2009

MapReduce usage statistics over time

	Aug, '04	Mar, '06	Sep, '07	Sep, '09
Number of jobs	29K	171K	2,217K	3,467K
Average completion time (secs)	634	874	395	475
Machine years used	217	2,002	11,081	25,562
Input data read (TB)	3,288	52,254	403,152	544,130
Intermediate data (TB)	758	6,743	34,774	90,120
Output data written (TB)	193	2,970	14,018	57,520
Average worker machines	157	268	394	488

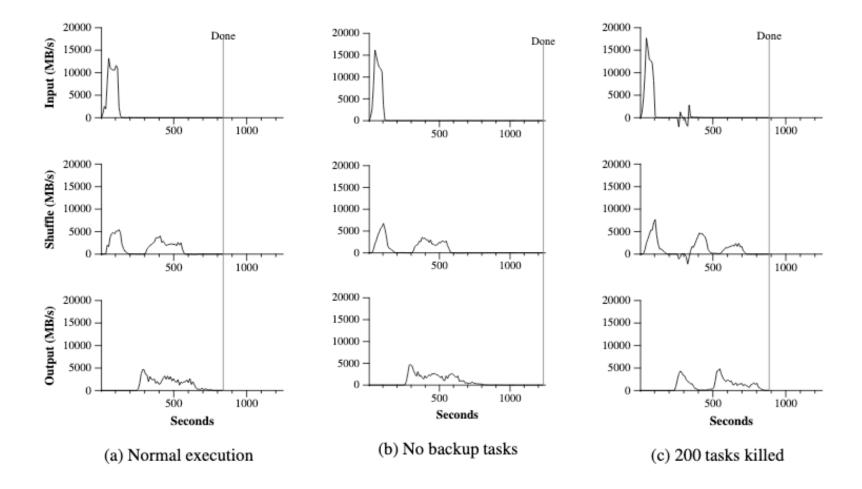
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 What will likely serve as a performance bottleneck for Google's MapReduce back in 2004 (or even earlier)? CPU? Memory? Disk? Network? Anything else?

 What will likely serve as a performance bottleneck for Google's MapReduce back in 2004 (or even earlier)? CPU? Memory? Disk? Network? Anything else?

 How does MapReduce reduce the effect of slow network?

 How does MapReduce jobs get good load balance across worker machines?



Network File System (NFS)

Primary goal

 Local FS: processes on same machine access shared files

 Network FS: processes on different machines access shared files in same way

Sub-objectives

- Fast + simple crash recovery
 - Both clients and file server may crash

- Transparent access
 - Can't tell it's over the network
 - Normal UNIX semantics

Reasonable performance

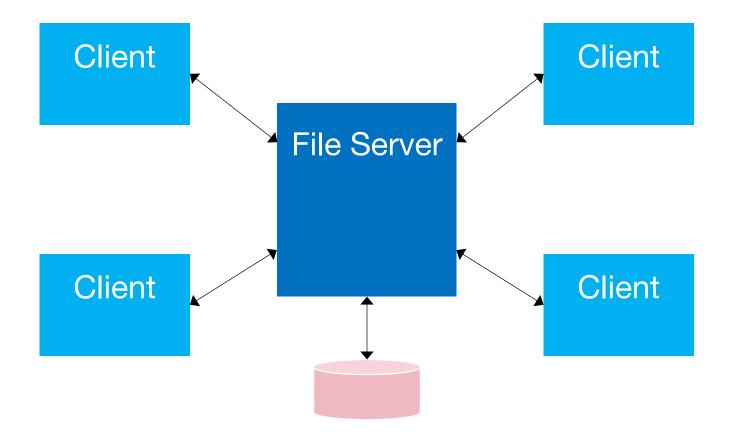
NFS agenda

Architecture

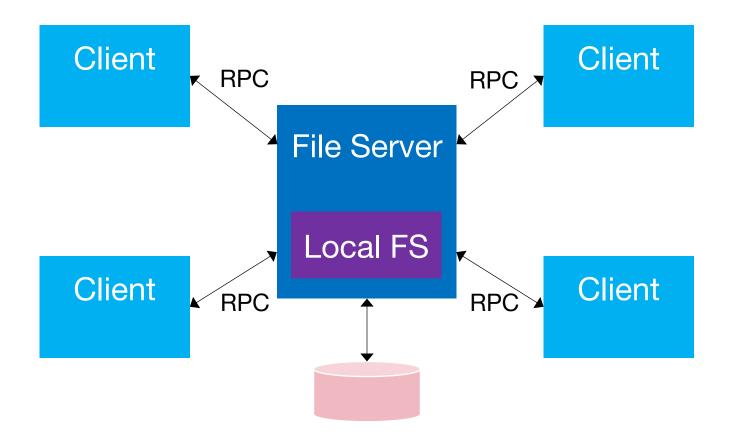
Network API

Cache

NFS architecture



NFS architecture



Main design decisions

What functions to expose via RPC?

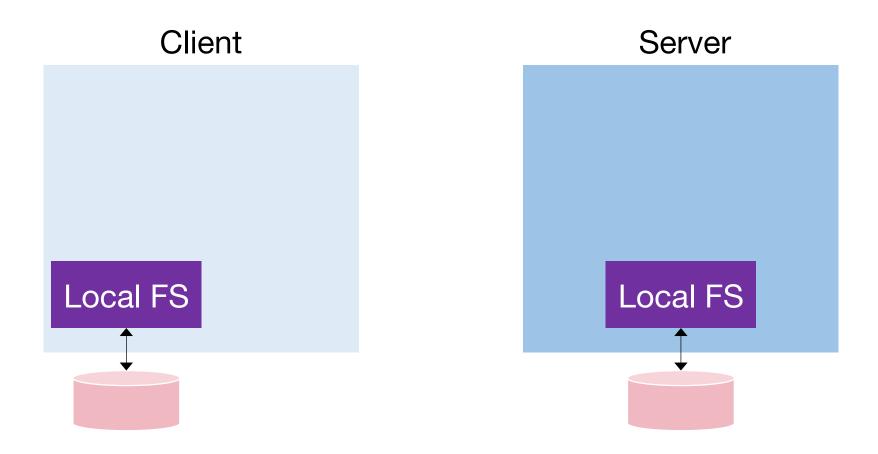
 Think of NFS as more of a protocol than a particular file system implementation

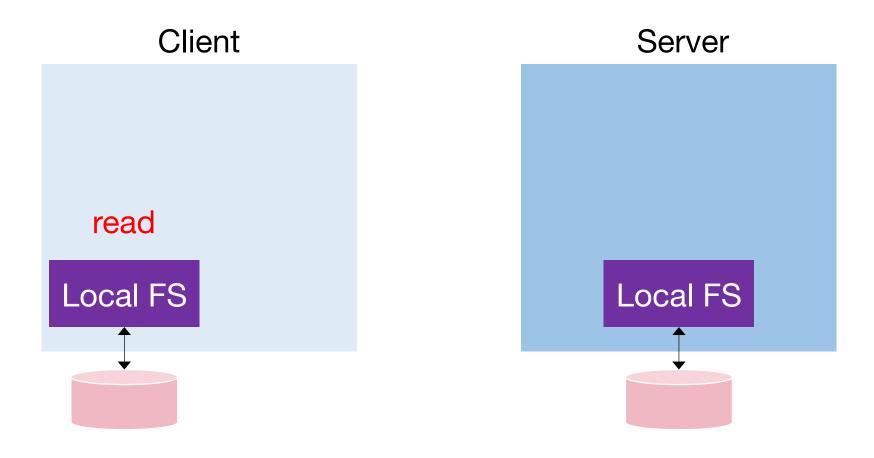
- Many companies have implemented NFS:
 - Oracle/Sun, NetApp, (Dell) EMC, IBM, etc.

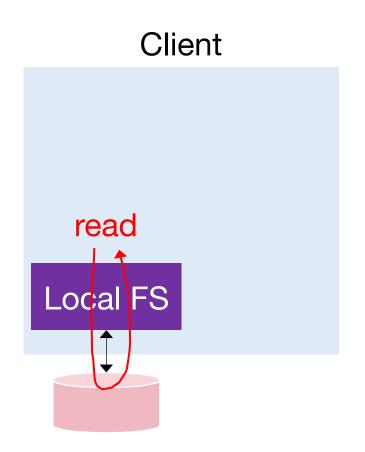
Today's lecture

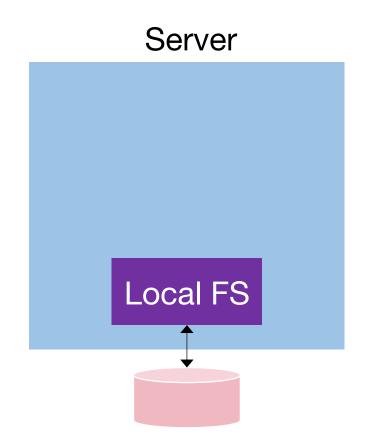
We're looking at NFSv2

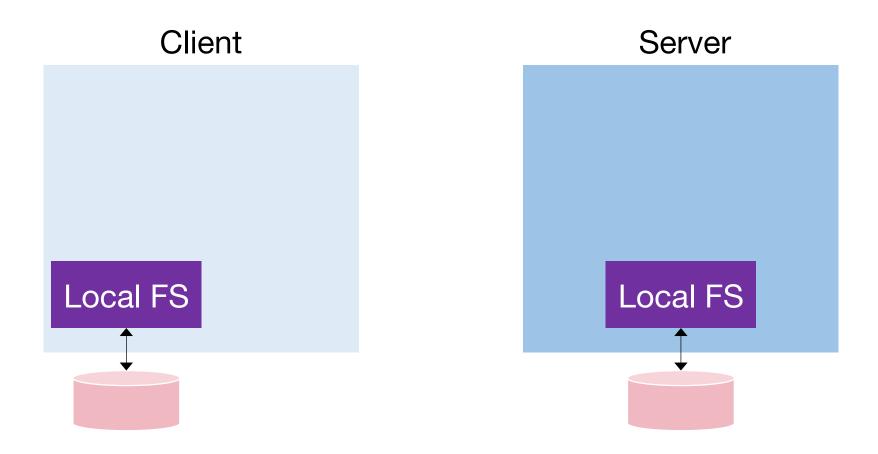
 There is now an NFSv3 and NFSv4 with many changes

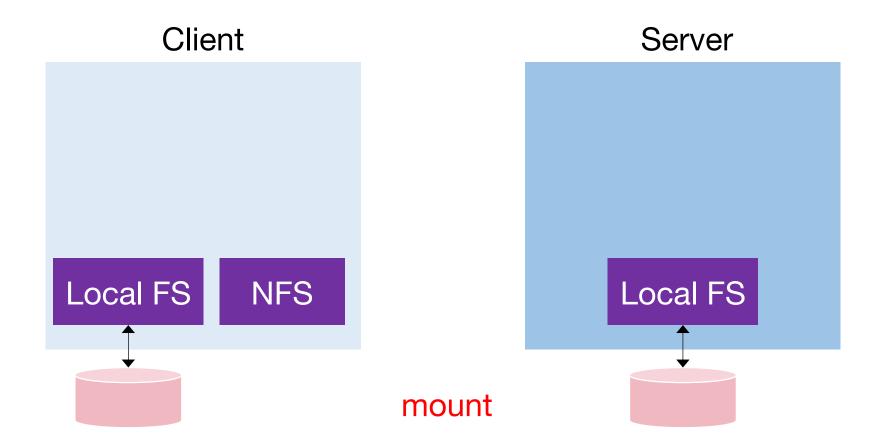


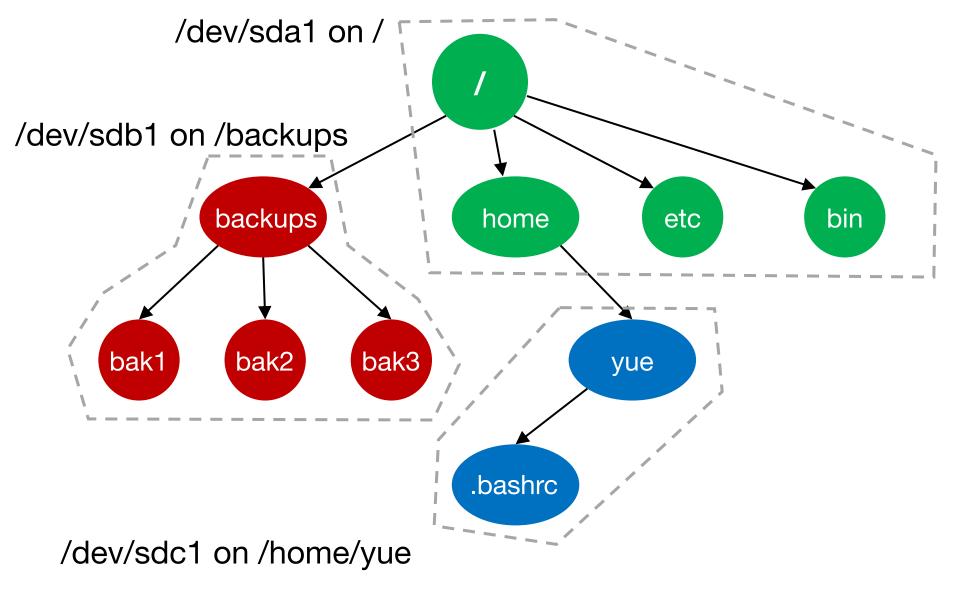


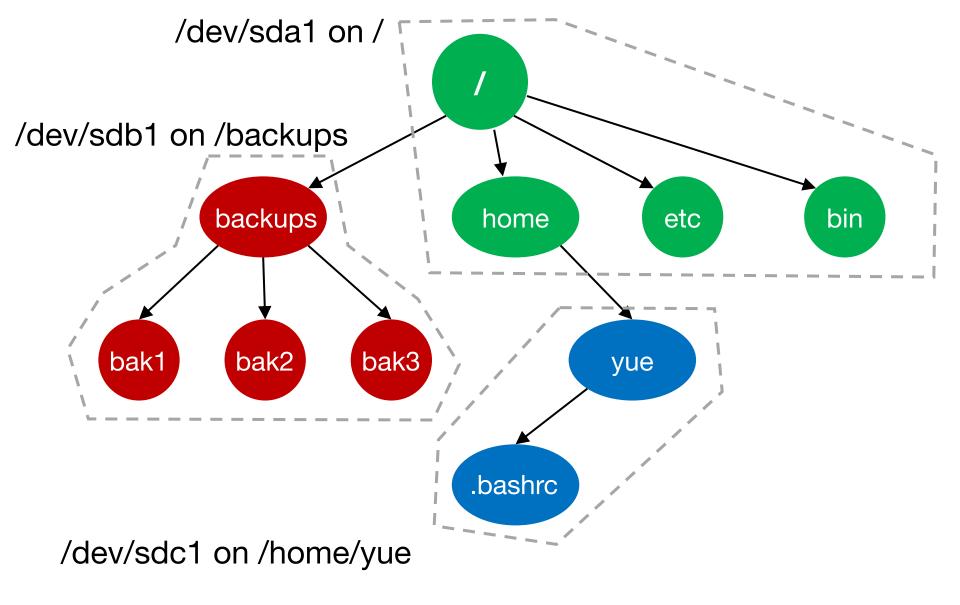




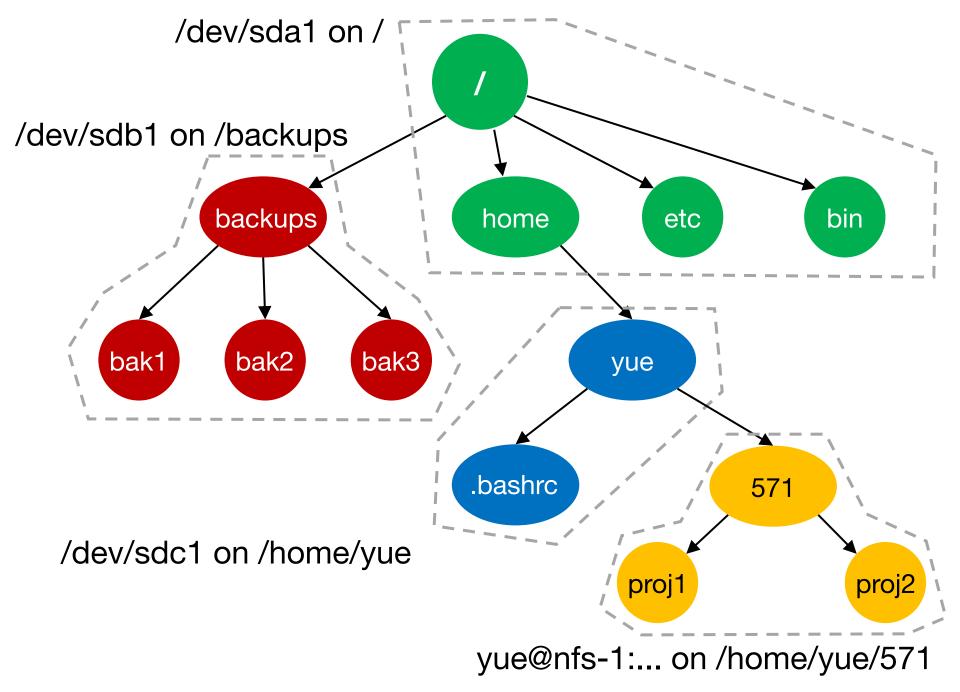


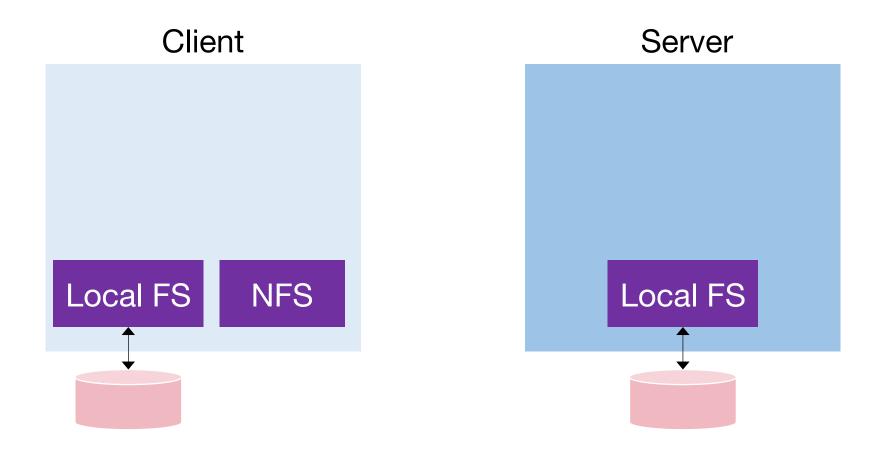


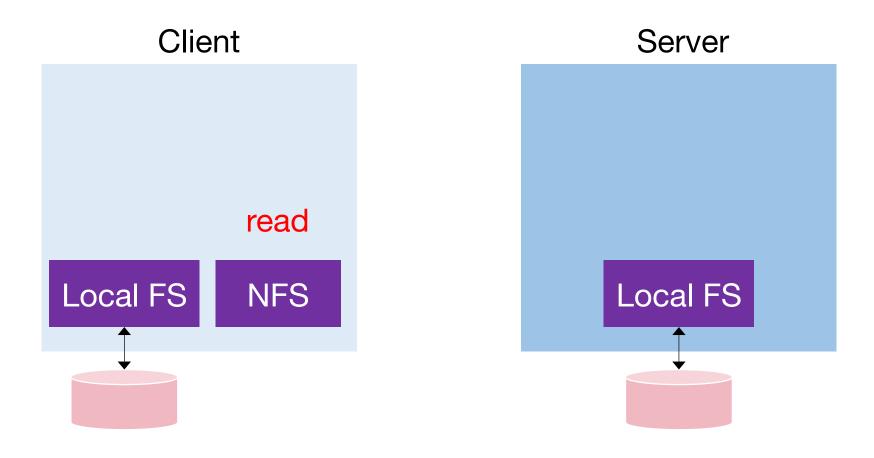


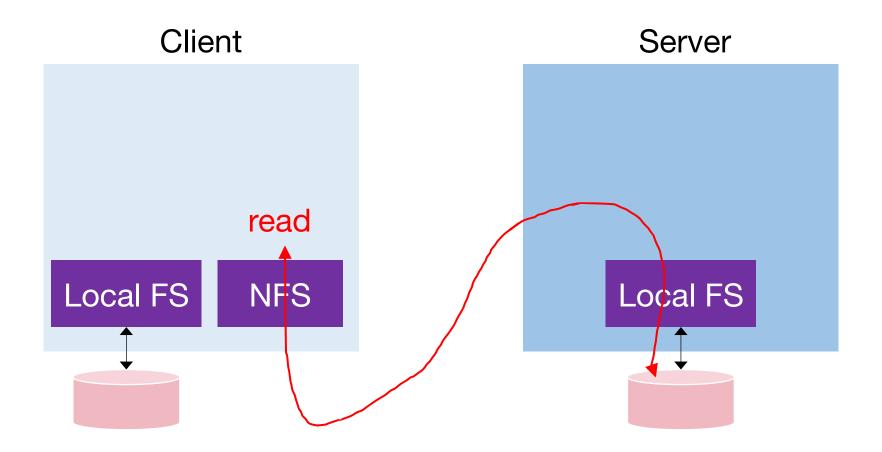


mount yue@nfs-1:... /home/yue/571









NFS agenda

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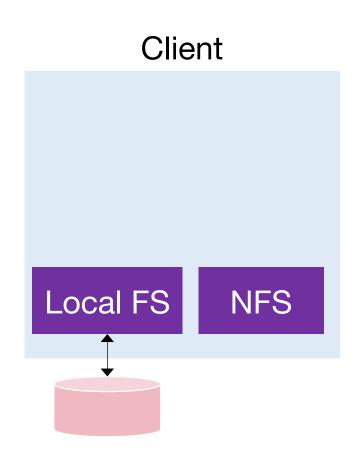
Strategy 1

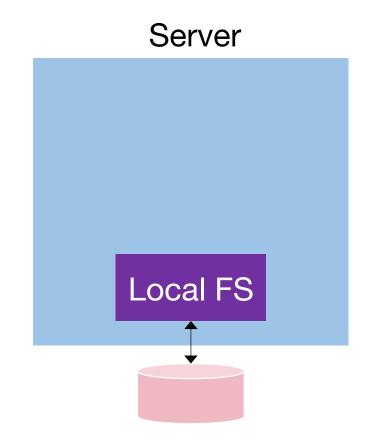
Wrap regular UNIX system calls using RPC

open() on client calls open() on server open() on server returns fd back to client

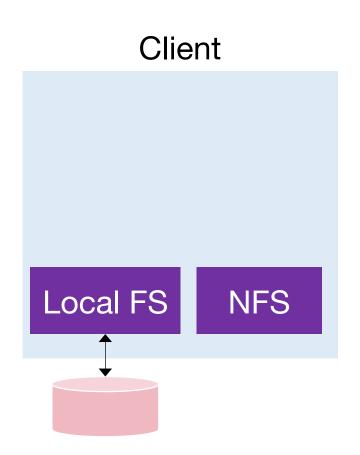
read(fd) on client calls read(fd) on server read(fd) on server returns data back to client

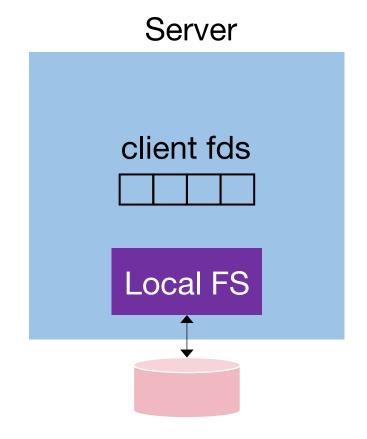
File descriptors

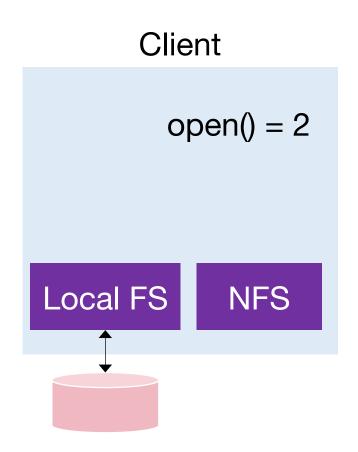


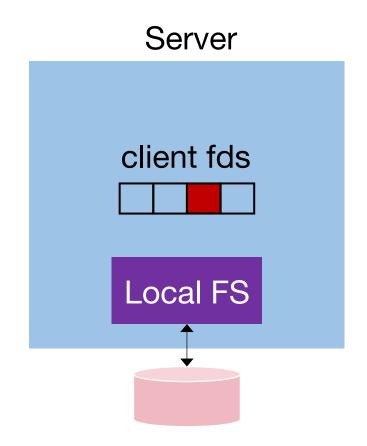


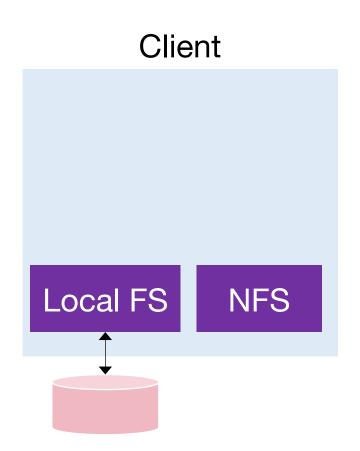
File descriptors

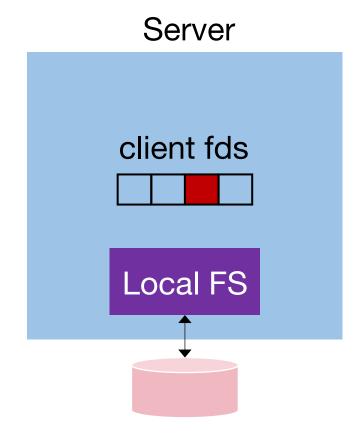


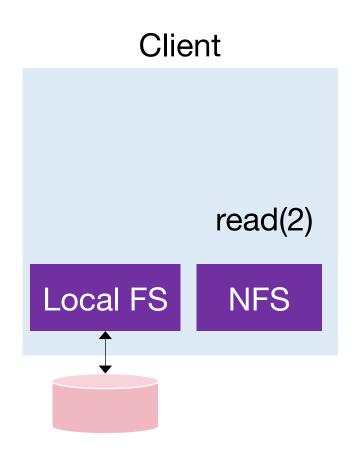


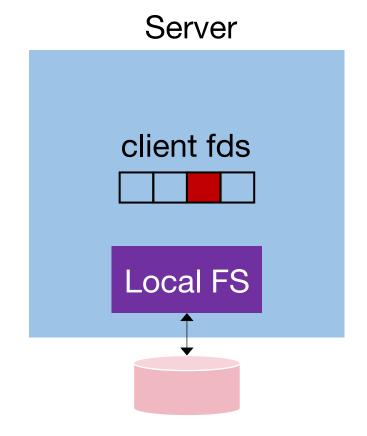


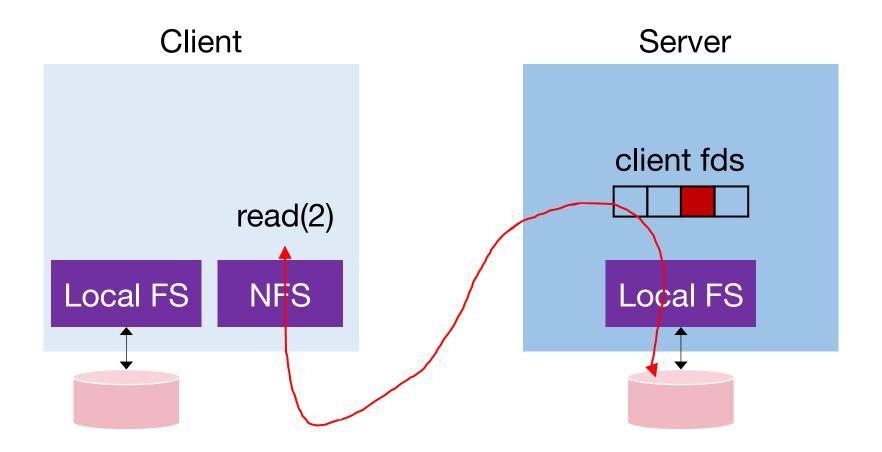












Strategy 1's problems

What about crashes?

```
int fd = open("foo", O_RDONLY);
read(fd, buf, MAX);
read(fd, buf, MAX);
...
read(fd, buf, MAX);
```

Imagine server crashes and reboots during reads...

Strategy 1's problems

What about crashes?

```
int fd = open("foo", O_RDONLY);
read(fd, buf, MAX);
read(fd, buf, MAX);
...
Nice if this just looks
read(fd, buf, MAX); like a slow read
```

Imagine server crashes and reboots during reads...

Potential solutions

- Run some crash recovery protocol upon reboot
 - Complex

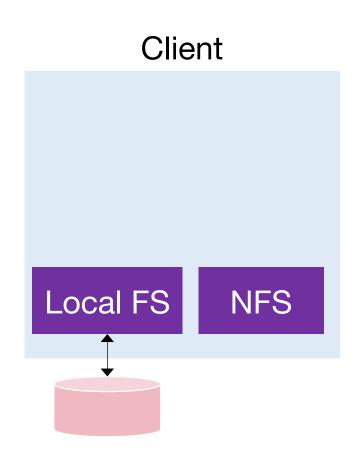
- 2. Persist fds on server disk
 - Slow
 - What if client crashes instead?

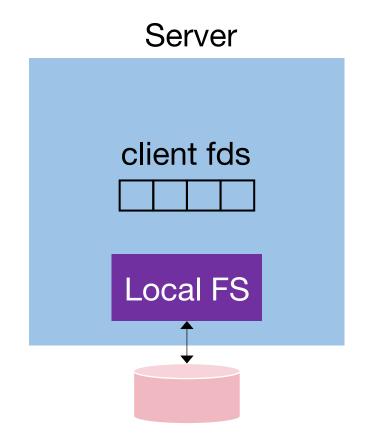
Strategy 2: Put all info in requests

- Use "stateless" protocol!
 - Server maintains no state about clients
 - Server still keeps other state, of course

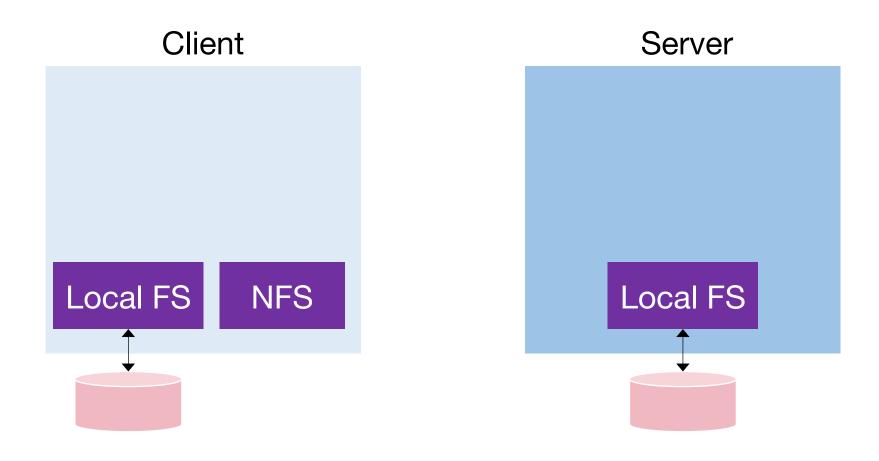
Eliminate file descriptors

Eliminate file descriptors





Eliminate file descriptors



Strategy 2: Put all info in requests

- Use "stateless" protocol!
 - Server maintains no state about clients
 - Server still keeps other state, of course

Need API change. One possibility:
 pread(char *path, buf, size, offset);
 pwrite(char *path, buf, size, offset);

 Specify path and offset each time. Server needs not remember. Pros/cons?

Strategy 2: Put all info in requests

- Use "stateless" protocol!
 - Server maintains no state about clients
 - Server still keeps other state, of course

Need API change. One possibility:
 pread(char *path, buf, size, offset);
 pwrite(char *path, buf, size, offset);

 Specify path and offset each time. Server needs not remember. Pros/cons? Too many path lookups

```
pread(char *path, buf, size, offset);
pwrite(char *path, buf, size, offset);
```

```
inode = open(char *path);
pread(inode, buf, size, offset);
pwrite(inode, buf, size, offset);
```

```
inode = open(char *path);
pread(inode, buf, size, offset);
pwrite(inode, buf, size, offset);
```

This is pretty good! Any correctness problems?

```
inode = open(char *path);
pread(inode, buf, size, offset);
pwrite(inode, buf, size, offset);
```

This is pretty good! Any correctness problems?

What if file is deleted, and inode is reused?

Strategy 4: File handles

```
fh = open(char *path);
pread(fh, buf, size, offset);
pwrite(fh, buf, size, offset);
```

File handle = <volume ID, inode #, generation #>

Aside: Append?

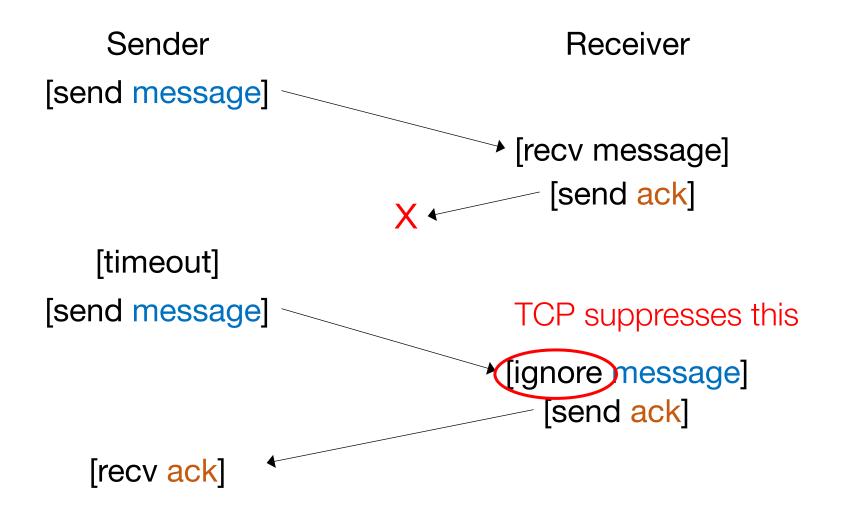
```
fh = open(char *path);
pread(fh, buf, size, offset);
pwrite(fh, buf, size, offset);
append(fh, buf, size);
```

Would append() be a good idea?

Problem: if our RPC library retries if no ACK or return, what happens when append is retried?

Solutions??

TCP remembers messages



Replica suppression is stateful

- TCP is stateful
 - If server crashes, it **forgets** what RPC's have been executed!

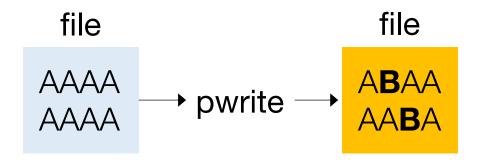
Replica suppression is stateful

- TCP is stateful
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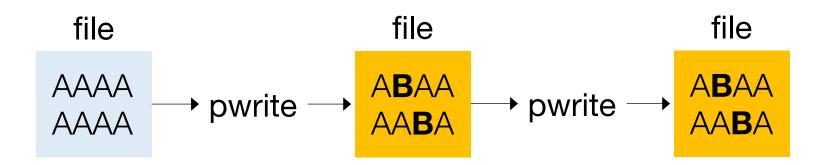
 Solution: design API so that there is no harm if executing a call more than once

- An API call that has this property is "idempotent":
 If f() is idempotent, then
 - f() has the same effect as f(); f(); ... f(); f()

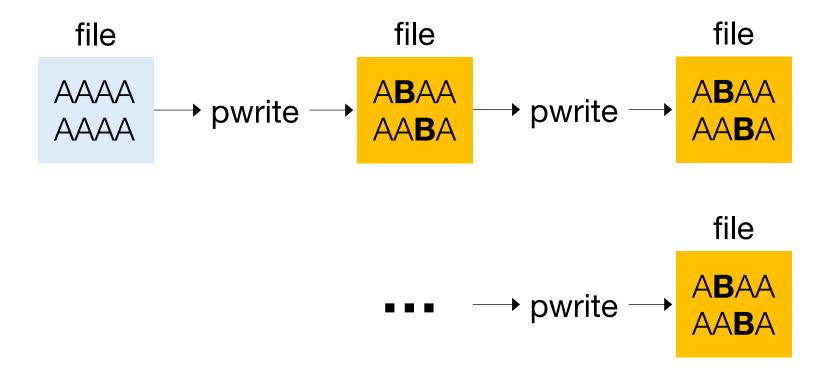
pwrite is idempotent



pwrite is idempotent

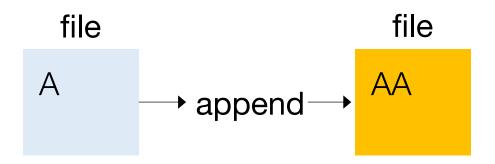


pwrite is idempotent

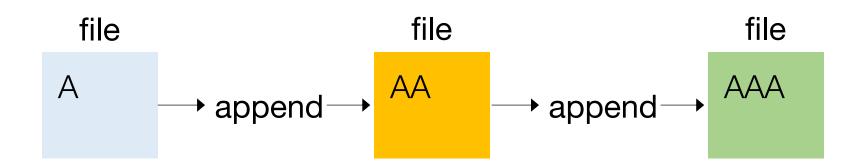


How about append?

Append is NOT idempotent



Append is NOT idempotent



Idempotence

- Idempotent
 - Any sort of read
 - pwrite

- Not idempotent
 - append

- What about these?
 - mkdir
 - creat

Strategy 4: File handles

```
fh = open(char *path);
pread(fh, buf, size, offset);
pwrite(fh, buf, size, offset);
append(fh, buf, size);
```

File handle = <volume ID, inode #, generation #>

NFS agenda

Architecture

Network API

Cache

- We can cache data in three places
 - Server memory
 - Client memory
 - Client disk

How to make sure all versions are in sync?

Client Server Client

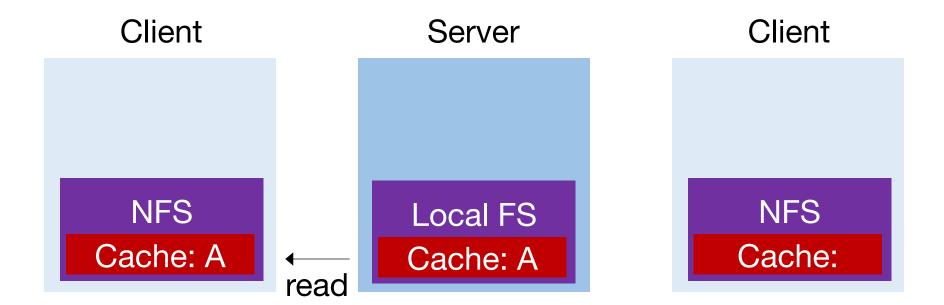
NFS
Cache: Cache: A

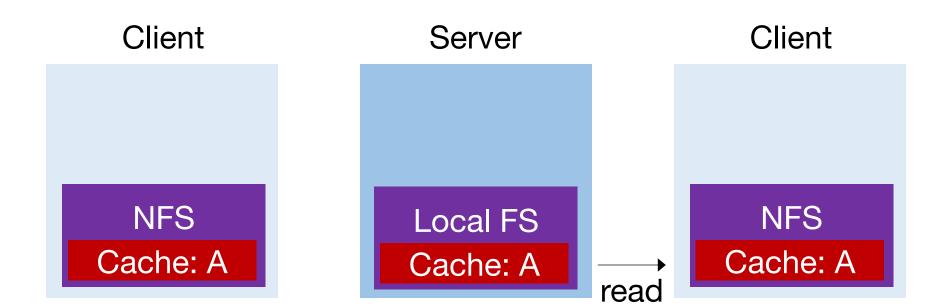
Client

Client

Client

Client





Client

Server

Client

NFS

Cache: A

Client

NFS

Cache: A

Cache: A

Client

Write

NFS
Cache: B

Client

Client

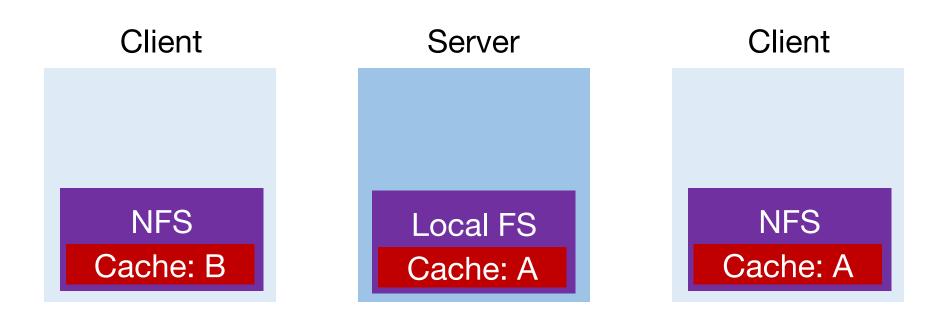
Client

Client

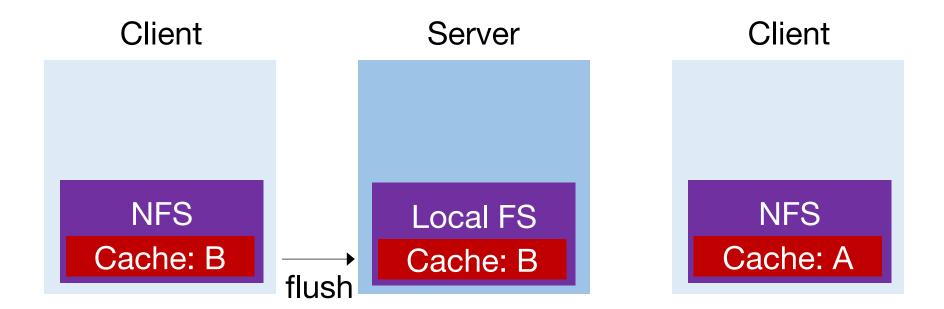
Client

Client

Client



"Update visibility" problem: server doesn't have latest



Client

Server

Client

NFS

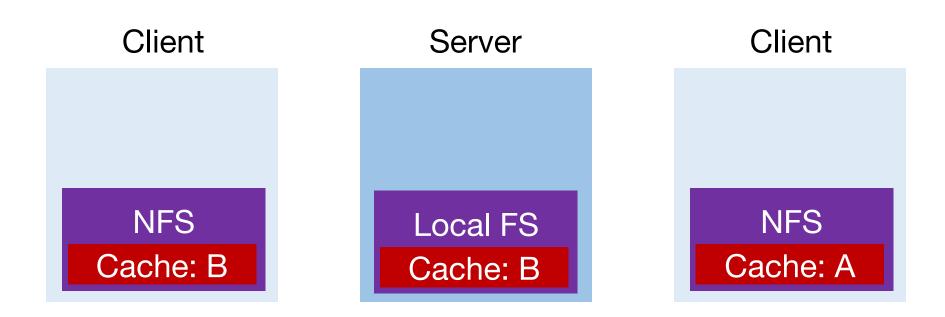
Cache: B

Client

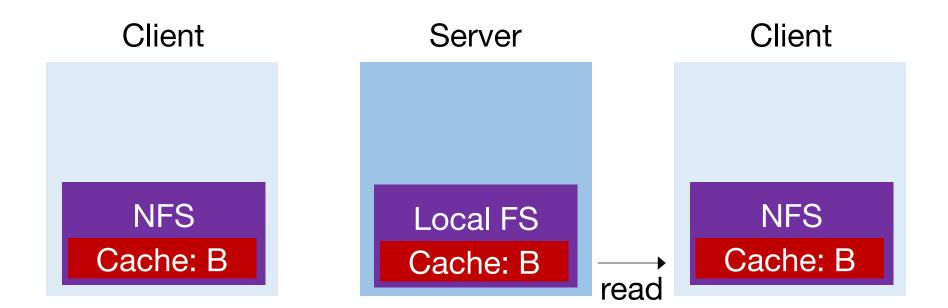
NFS

Cache: A

Client



"Stale cache" problem: client doesn't have latest



Problem 1: Update visibility

A client may buffer a write

How can server and other clients see it?

- NFS solution: flush on fd close
 - Not quite like UNIX

Problem 2: Stale cache

A client may have a cached copy that is obsolete

How can we get the latest?

Problem 2: Stale cache

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- How can we get the latest?
- If we weren't trying to be stateless, server could push out update in a timely manner

 NFS solution: clients recheck if cache is current before using it

Stale cache solution

Client caches metadata records when data was fetched

- Before it is used, client does a stat request to server
 - Gets last modified timestamp
 - Compares to cache
 - Refetches if necessary

Measure then build

 NFS developers found stat accounted for 90% of server requests

Why? Because clients frequently recheck cache

Reducing stat calls

Solution: cache results of stat calls

Why is this a terrible solution?

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Why is this a terrible solution?

 Make the stat cache entries expire after a given time (say 3 seconds)

 Why is this better than putting expirations on the regular cache?

Summary

- Robust APIs are often:
 - Stateless: servers don't remember clients
 - **Idempotent**: doing things twice never hurts

- Supporting existing specs is a lot harder than building from scratch!
- Caching is hard! Caching is harder in distributed systems, especially with crashes.

That's all...

- Please make sure to complete the online teaching evaluation form
 - Evaluation opens 04/20 and closes 04/30

Next week: final review