CS162
Operating Systems and Systems Programming Lecture 22

Distributed Decision Making (Finished), TCP/IP Networking, RPC

April 21st, 2020 Prof. John Kubiatowicz http://cs162.eecs.Berkeley.edu

Acknowledgments: Lecture slides are from the Operating Systems course taught by John Kubiatowicz at Berkeley, with few minor updates/changes. When slides are obtained from other sources, a reference will be noted on the bottom of that slide, in which case a full list of references is provided on the last slide.

Recall: Distributed Consensus Making

- Consensus problem
 - All nodes propose a value
 - Some nodes might crash and stop responding
 - Eventually, all remaining nodes decide on the same value from set of proposed values
- Distributed Decision Making
 - Choose between "true" and "false"
 - Or Choose between "commit" and "abort"
- Equally important (but often forgotten!): make it durable!
 - How do we make sure that decisions cannot be forgotten?
 - » This is the "D" of "ACID" in a regular database
 - In a global-scale system?
 - » What about erasure coding or massive replication?
 - » Like BlockChain applications!

Recall: Two-Phase Commit

- Since we can't solve the General's Paradox (i.e. simultaneous action), let's solve a related problem
 - Distributed transaction: Two machines agree to do something, or not do it, atomically
- Two-Phase Commit protocol:
 - Prepare Phase:
 - » The global coordinator requests that all participants will promise to commit or rollback the transaction
 - » Participants record promise in log, then acknowledge
 - » If anyone votes to abort, coordinator writes "Abort" in its log and tells everyone to abort; each records "Abort" in log
 - Commit Phase:
 - » After all participants respond that they are prepared, then the coordinator writes "Commit" to its log
 - » Then asks all nodes to commit; they respond with ack
 - » After receive acks, coordinator writes "Got Commit" to log
- Persistent stable log on each machine:
 - Help nodes remember what they have said that they would do
 - » If a machine crashes, when it wakes up it first checks its log to recover state of world at time of crash
 - » Log can be used to complete this process such that all machines either commit or don't commit

Two-Phase Commit: Setup

- One machine (coordinator) initiates the protocol
- It asks every machine to vote on transaction
- Two possible votes:
 - Commit
 - Abort
- Commit transaction only if unanimous approval

Two-Phase Commit: Preparing

Agree to Commit

- Machine has guaranteed that it will accept transaction
- Must be recorded in log so machine will remember this decision if it fails and restarts

Agree to Abort

- Machine has guaranteed that it will never accept this transaction
- Must be recorded in log so machine will remember this decision if it fails and restarts

Two-Phase Commit: Finishing

Commit Transaction

- Coordinator learns all machines have agreed to commit
- Record decision to commit in local log
- Apply transaction, inform voters

Abort Transaction

- Coordinator learns at least on machine has voted to abort
- Record decision to abort in local log
- Do not apply transaction, inform voters

Two-Phase Commit: Finishing

Commit Transaction

- Coordinator learns all machines have agreed to commit
- Record decision to commit in local log
- Apply transaction, inform voters

Abort Transaction

- ine exactive Coordinator learns at least on mach
- Record decision to abort in local
- voted to about to abo Do not apply transaction

Detailed Algorithm

Coordinator Algorithm

Coordinator sends **VOTE-REQ** to all workers

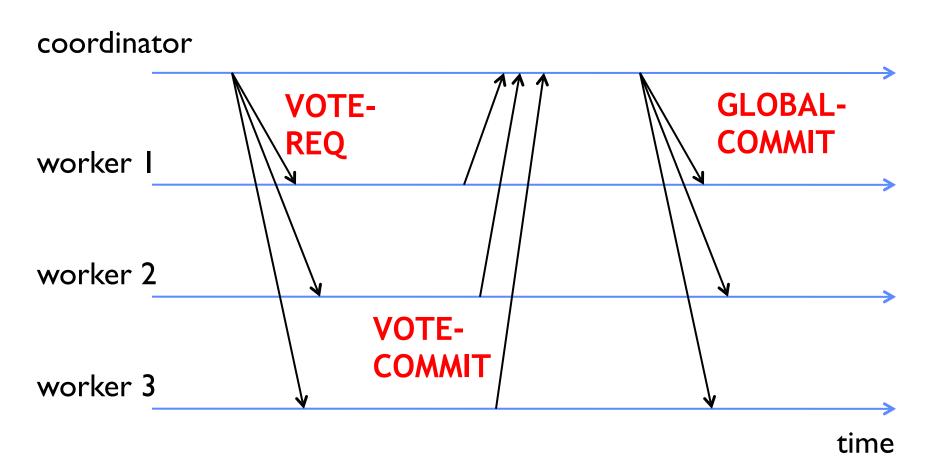
- If receive VOTE-COMMIT from all N workers, send GLOBAL-COMMIT to all workers
- If don't receive VOTE-COMMIT from all N workers, send GLOBAL-ABORT to all workers

Worker Algorithm

- Wait for VOTE-REQ from coordinator
- If ready, send VOTE-COMMIT to coordinator
 - If not ready, send **VOTE-ABORT** to coordinator
 - And immediately abort

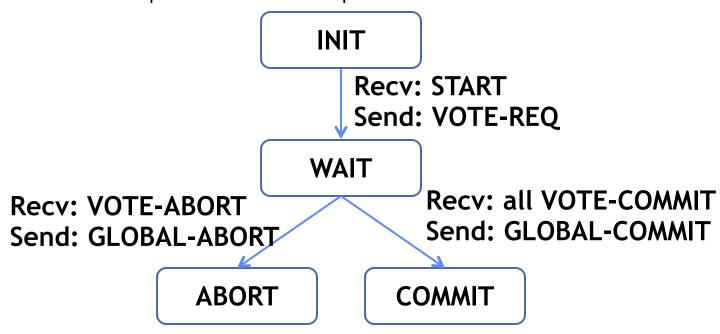
- If receive GLOBAL-COMMIT then commit
- If receive GLOBAL-ABORT then abort

Failure Free Example Execution

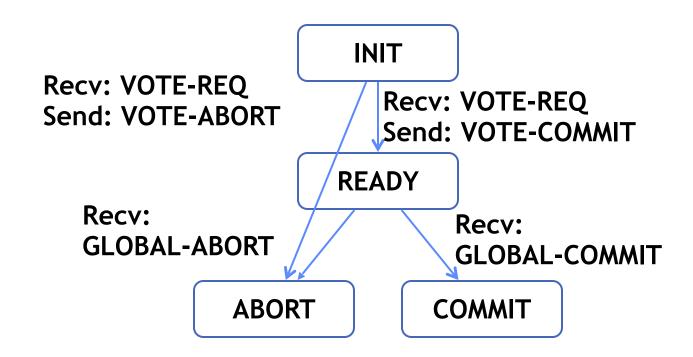


State Machine of Coordinator

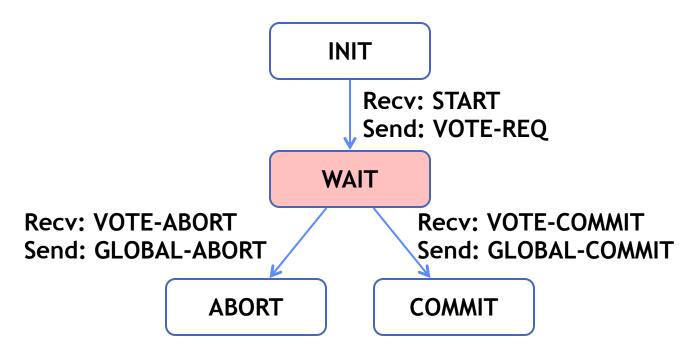
Coordinator implements simple state machine:



State Machine of Workers

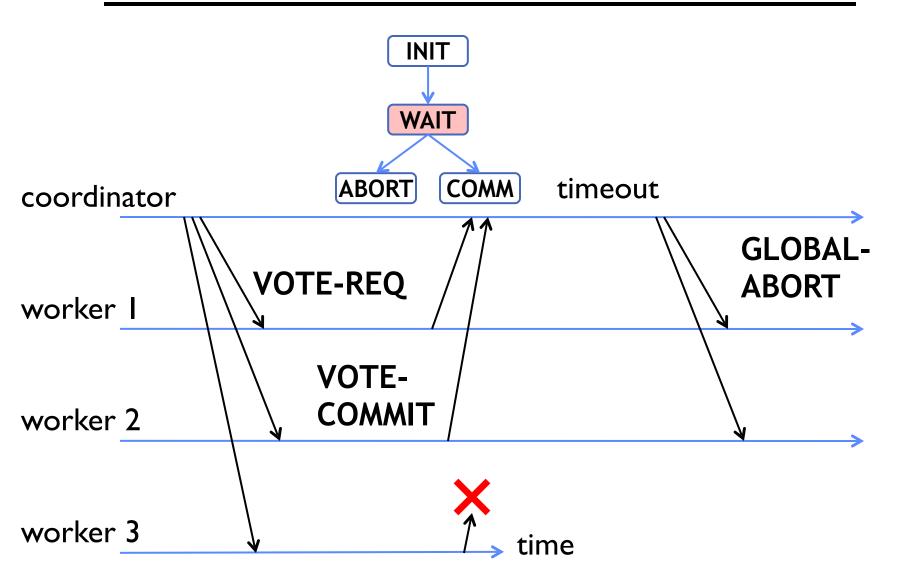


Dealing with Worker Failures

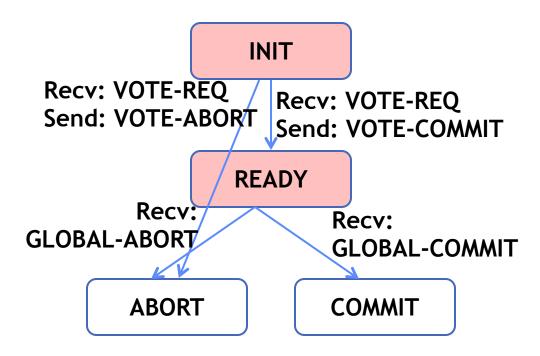


- Failure only affects states in which the coordinator is waiting for messages
- Coordinator only waits for votes in "WAIT" state
- In WAIT, if doesn't receive N votes, it times out and sends
 GLOBAL-ABORT

Example of Worker Failure

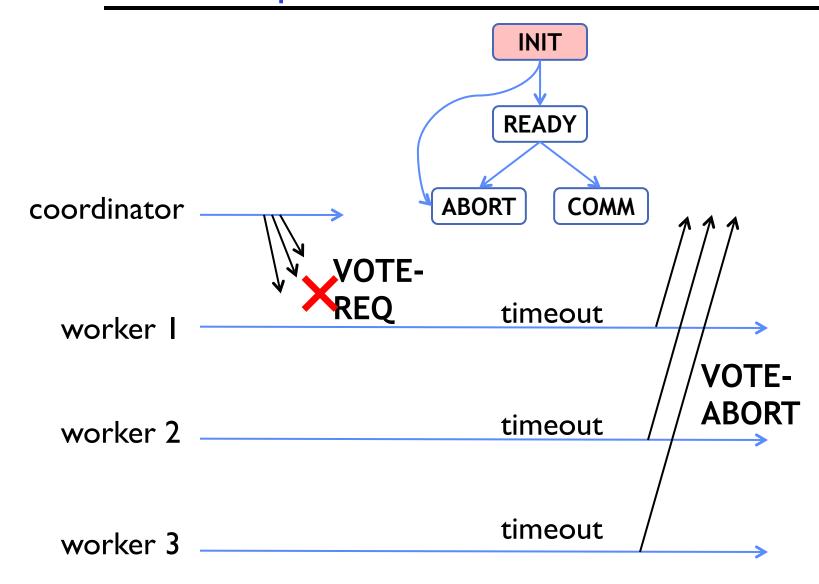


Dealing with Coordinator Failure

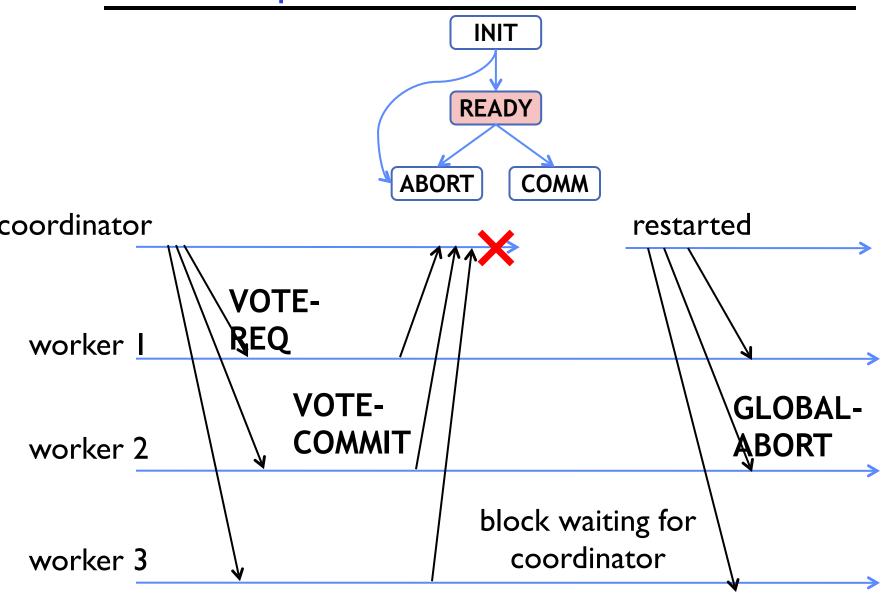


- Worker waits for VOTE-REQ in INIT
 - Worker can time out and abort (coordinator handles it)
- Worker waits for GLOBAL-* message in READY
 - If coordinator fails, workers must BLOCK waiting for coordinator to recover and send GLOBAL_* message

Example of Coordinator Failure #1



Example of Coordinator Failure #2

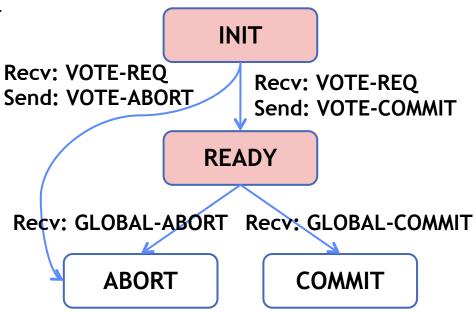


Durability

- All nodes use stable storage to store current state
 - stable storage is non-volatile storage (e.g. backed by disk) that guarantees atomic writes.
 - E.g.: SSD, NVRAM
- Upon recovery, nodes can restore state and resume:
 - Coordinator aborts in INIT, WAIT, or ABORT
 - Coordinator commits in COMMIT
 - Worker aborts in INIT, ABORT
 - Worker commits in COMMIT
 - Worker "asks" Coordinator in READY

Blocking for Coordinator to Recover

- A worker waiting for global decision can ask fellow workers about their state
 - If another worker is in ABORT or COMMIT state then coordinator must have sent GLOBAL-*
 - » Thus, worker can safely abort or commit, respectively



 If all workers are in ready, need to BLOCK (don't know if coordinator wanted to abort or commit)

Distributed Decision Making Discussion (1/2)

- Why is distributed decision making desirable?
 - Fault Tolerance!
 - A group of machines can come to a decision even if one or more of them fail during the process
 - After decision made, result recorded in multiple places
- Why is 2PC not subject to the General's paradox?
 - Because 2PC is about all nodes eventually coming to the same decision – not necessarily at the same time!
 - Allowing us to reboot and continue allows time for collecting and collating decisions

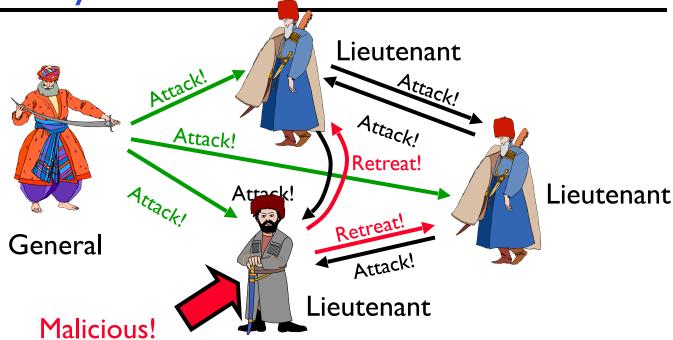
Distributed Decision Making Discussion (2/2)

- Undesirable feature of Two-Phase Commit: Blocking
 - One machine can be stalled until another site recovers:
 - » Site B writes "prepared to commit" record to its log, sends a "yes" vote to the coordinator (site A) and crashes
 - » Site A crashes
 - » Site B wakes up, check its log, and realizes that it has voted "yes" on the update. It sends a message to site A asking what happened. At this point, B cannot decide to abort, because update may have committed
 - » B is blocked until A comes back
 - A blocked site holds resources (locks on updated items, pages pinned in memory, etc) until learns fate of update

Alternatives to 2PC

- Three-Phase Commit: One more phase, allows nodes to fail or block and still make progress.
- PAXOS: An alternative used by Google and others that does not have
 2PC blocking problem
 - Develop by Leslie Lamport (Turing Award Winner)
 - No fixed leader, can choose new leader on fly, deal with failure
 - Some think this is extremely complex!
- RAFT: PAXOS alternative from John Osterhout (Stanford)
 - Simpler to describe complete protocol
- What happens if one or more of the nodes is malicious?
 - Malicious: attempting to compromise the decision making

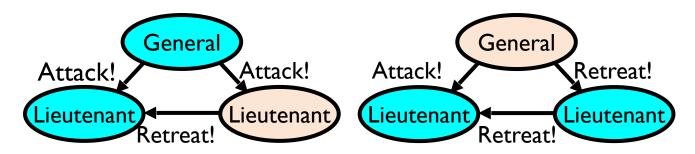
Byzantine General's Problem



- Byazantine General's Problem (n players):
 - One General and n-1 Lieutenants
 - Some number of these (f) can be insane or malicious
- The commanding general must send an order to his n-1 lieutenants such that the following Integrity Constraints apply:
 - ICI: All loyal lieutenants obey the same order
 - IC2: If the commanding general is loyal, then all loyal lieutenants obey the order he sends

Byzantine General's Problem (con't)

- Impossibility Results:
 - Cannot solve Byzantine General's Problem with n=3 because one malicious player can mess up things



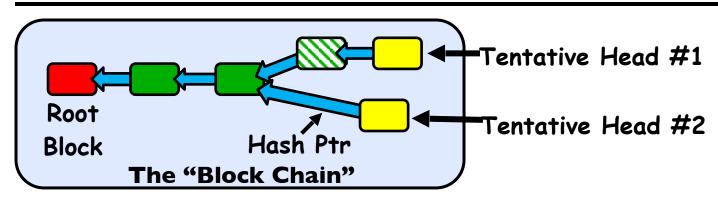
- With f faults, need n > 3f to solve problem
- Various algorithms exist to solve problem
 - Original algorithm has #messages exponential in n
 - Newer algorithms have message complexity O(n2)
 - » One from MIT, for instance (Castro and Liskov, 1999)
- Use of BFT (Byzantine Fault Tolerance) algorithm
 - Allow multiple machines to make a coordinated decision even if some subset of them (< n/3) are malicious

Request

Decision

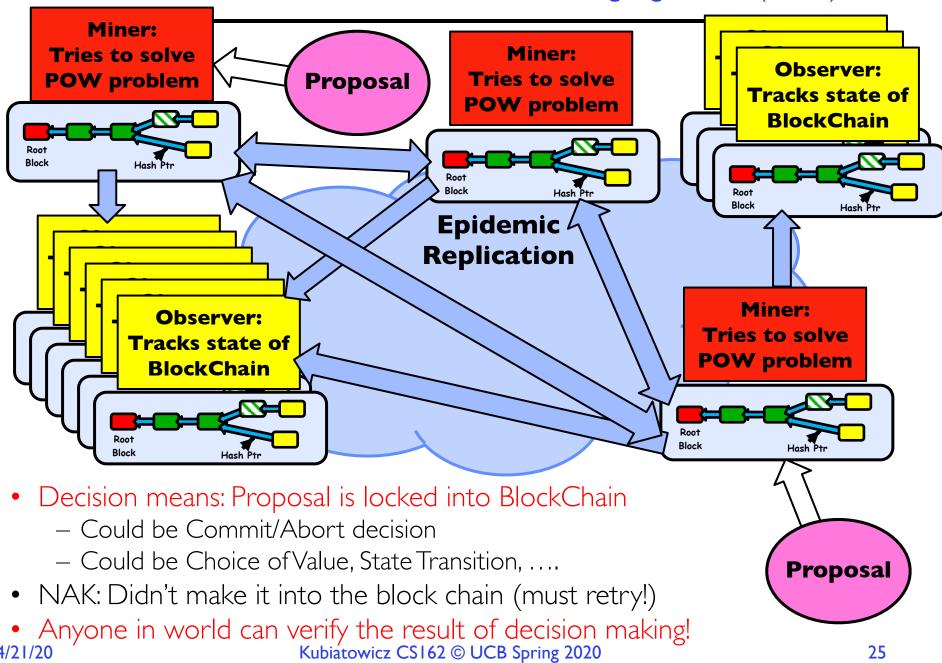
Distributed

Is a BlockChain a Distributed Decision Making Algorithm?



- BlockChain: a chain of blocks connected by hashes to root block
 - The Hash Pointers are unforgeable (assumption)
 - The Chain has no branches except perhaps for heads
 - Blocks are considered "authentic" part of chain when they have authenticity info in them
- How is the head chosen?
 - Some consensus algorithm
 - In many BlockChain algorithms (e.g. BitCoin, Ethereum), the head is chosen by solving hard problem
 - » This is the job of "miners" who try to find "nonce" info that makes hash over block have specified number of zero bits in it
 - » The result is a "Proof of Work" (POW)
 - » Selected blocks above (green) have POW in them and can be included in chains
 - Longest chain wins

Is a Blockchain a Distributed Decision Making Algorithm? (Con't)



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Remote Procedure Call (RPC)

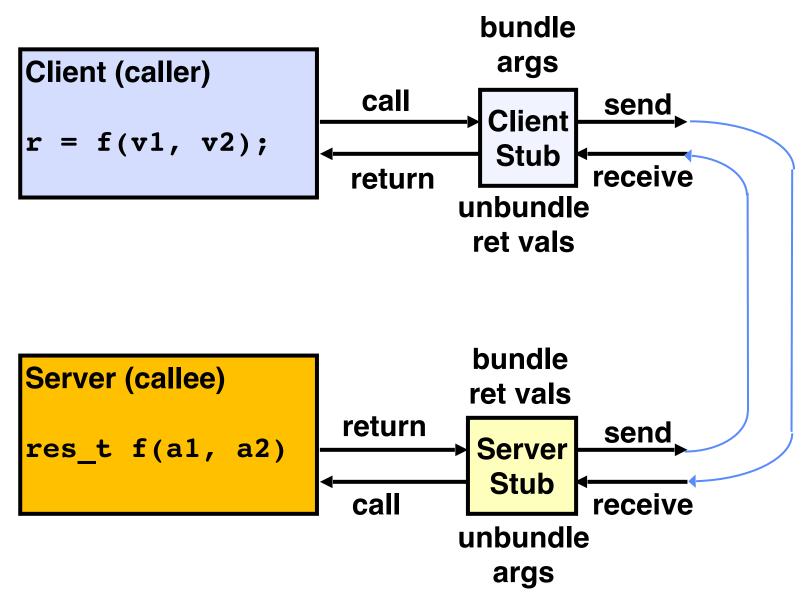
- Raw messaging is a bit too low-level for programming
 - Must wrap up information into message at source
 - Must decide what to do with message at destination
 - May need to sit and wait for multiple messages to arrive
 - And what about machines with different byte order ("BigEndian" vs "LittleEndian")
- Another option: Remote Procedure Call (RPC)
 - Calls a procedure on a remote machine
 - Client calls:

```
remoteFileSystem→Read("rutabaga");
```

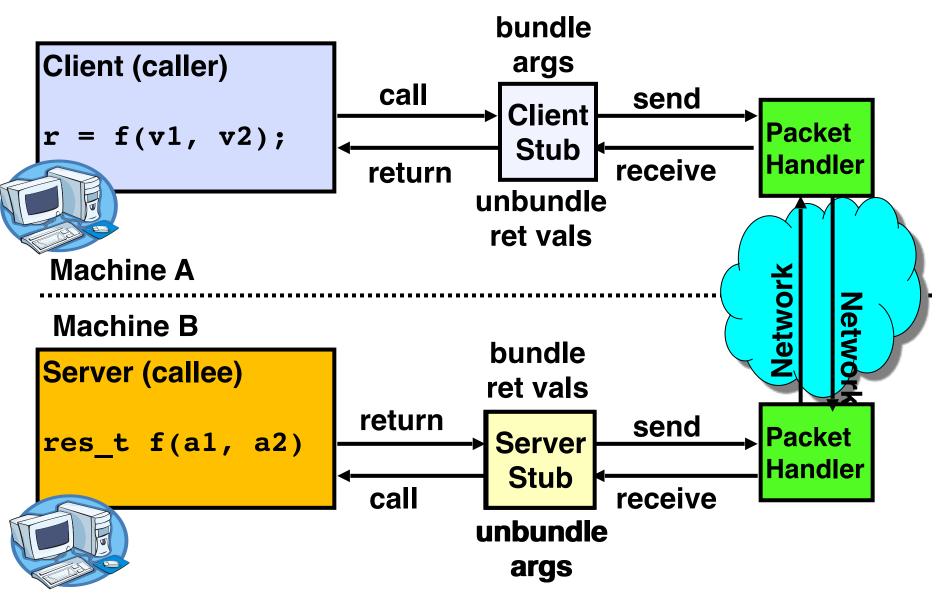
- Translated automatically into call on server:

```
fileSys→Read("rutabaga");
```

RPC Concept



RPC Information Flow



RPC Implementation

- Request-response message passing (under covers!)
- "Stub" provides glue on client/server
 - Client stub is responsible for "marshalling" arguments and "unmarshalling" the return values
 - Server-side stub is responsible for "unmarshalling" arguments and "marshalling" the return values.
- Marshalling involves (depending on system)
 - Converting values to a canonical form, serializing objects, copying arguments passed by reference, etc.

RPC Details (1/3)

- Equivalence with regular procedure call
 - Parameters ⇔ Request Message
 - Result ⇔ Reply message
 - Name of Procedure: Passed in request message
 - Return Address: mbox2 (client return mail box)
- Stub generator: Compiler that generates stubs
 - Input: interface definitions in an "interface definition language (IDL)"
 - » Contains, among other things, types of arguments/return
 - Output: stub code in the appropriate source language
 - » Code for client to pack message, send it off, wait for result, unpack result and return to caller
 - » Code for server to unpack message, call procedure, pack results, send them off

RPC Details (2/3)

- Cross-platform issues:
 - What if client/server machines are different architectures/ languages?
 - » Convert everything to/from some canonical form
 - » Tag every item with an indication of how it is encoded (avoids unnecessary conversions)
- How does client know which mbox (destination queue) to send to?
 - Need to translate name of remote service into network endpoint (Remote machine, port, possibly other info)
 - Binding: the process of converting a user-visible name into a network endpoint
 - » This is another word for "naming" at network level
 - » Static: fixed at compile time
 - » Dynamic: performed at runtime

RPC Details (3/3)

- Dynamic Binding
 - Most RPC systems use dynamic binding via name service
 - » Name service provides dynamic translation of service → mbox
 - Why dynamic binding?
 - » Access control: check who is permitted to access service
 - » Fail-over: If server fails, use a different one
- What if there are multiple servers?
 - Could give flexibility at binding time
 - » Choose unloaded server for each new client
 - Could provide same mbox (router level redirect)
 - » Choose unloaded server for each new request
 - » Only works if no state carried from one call to next
- What if multiple clients?
 - Pass pointer to client-specific return mbox in request

Problems with RPC: Non-Atomic Failures

- Different failure modes in dist. system than on a single machine
- Consider many different types of failures
 - User-level bug causes address space to crash
 - Machine failure, kernel bug causes all processes on same machine to fail
 - Some machine is compromised by malicious party
- Before RPC: whole system would crash/die
- After RPC: One machine crashes/compromised while others keep working
- Can easily result in inconsistent view of the world
 - Did my cached data get written back or not?
 - Did server do what I requested or not?
- Answer? Distributed transactions/Byzantine Commit

Problems with RPC: Performance

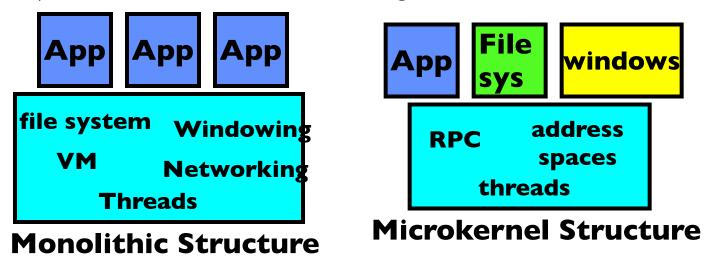
- RPC is *not* performance transparent:
 - Cost of Procedure call « same-machine RPC « network RPC
 - Overheads: Marshalling, Stubs, Kernel-Crossing, Communication
- Programmers must be aware that RPC is not free
 - Caching can help, but may make failure handling complex

Cross-Domain Communication / Location Transparency

- How do address spaces communicate with one another?
 - Shared Memory with Semaphores, monitors, etc...
 - File System
 - Pipes (I-way communication)
 - "Remote" procedure call (2-way communication)
- RPC's can be used to communicate between address spaces on different machines or the same machine
 - Services can be run wherever it's most appropriate
 - Access to local and remote services looks the same
- Examples of RPC systems:
 - CORBA (Common Object Request Broker Architecture)
 - DCOM (Distributed COM)
 - RMI (Java Remote Method Invocation)

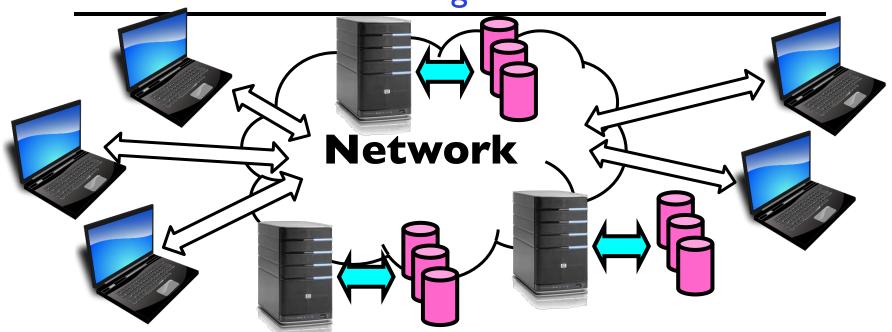
Microkernel operating systems

- Example: split kernel into application-level servers.
 - File system looks remote, even though on same machine



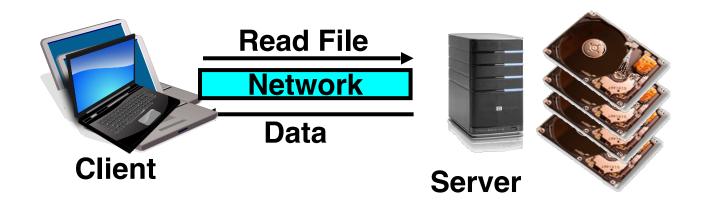
- Why split the OS into separate domains?
 - Fault isolation: bugs are more isolated (build a firewall)
 - Enforces modularity: allows incremental upgrades of pieces of software (client or server)
 - Location transparent: service can be local or remote
 - » For example in the X windowing system: Each X client can be on a separate machine from X server;

Network-Attached Storage and the CAP Theorem



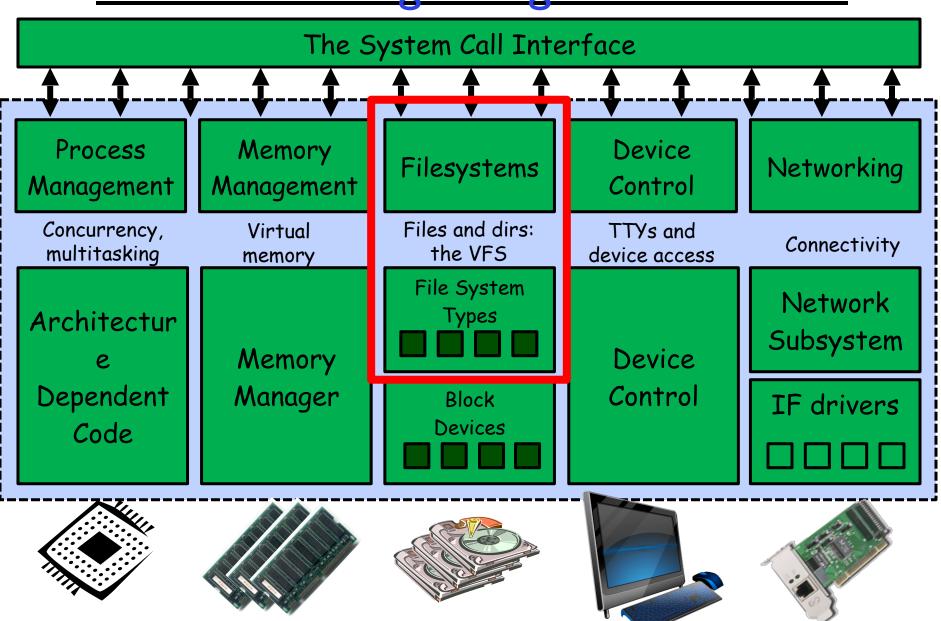
- Consistency:
 - Changes appear to everyone in the same serial order
- Availability:
 - Can get a result at any time
- Partition-Tolerance
 - System continues to work even when network becomes partitioned
- Consistency, Availability, Partition-Tolerance (CAP) Theorem: Cannot have all three at same time
 - Otherwise known as "Brewer's Theorem"

Distributed File Systems

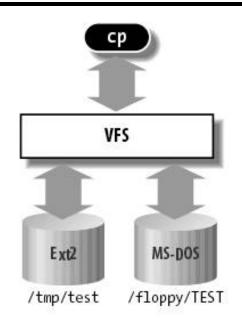


- Transparent access to files stored on a remote disk
- Mount remote files into your local file system
 - Directory in local file system refers to remote files
 - e.g., /home/oksi/162/ on laptop actually refers to / users/oski on campus file server

Enabling Design: VFS

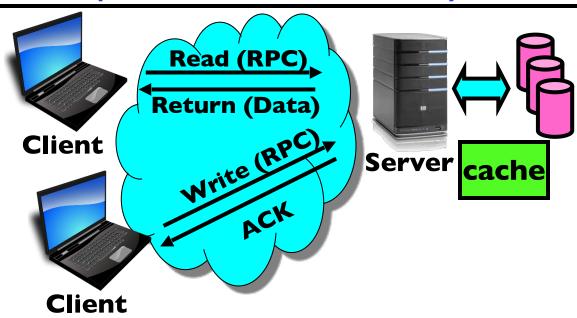


Virtual Filesystem Switch (Con't)



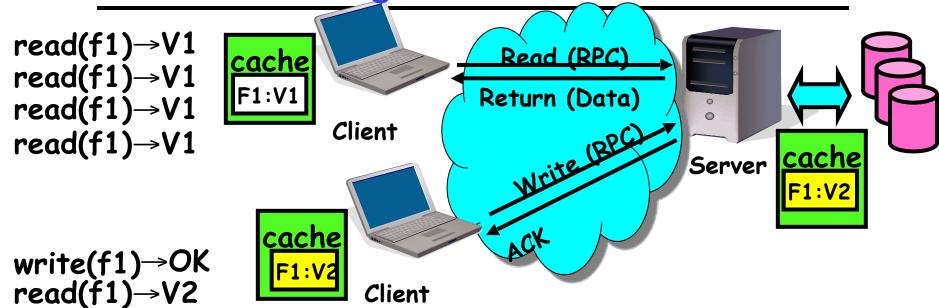
- VFS: Virtual abstraction similar to local file system
 - Provides virtual superblocks, inodes, files, etc
 - Compatible with a variety of local and remote file systems
 - » provides object-oriented way of implementing file systems
- VFS allows the same system call interface (the API) to be used for different types of file systems
 - The API is to the VFS interface, rather than any specific type of file system

Simple Distributed File System



- Remote Disk: Reads and writes forwarded to server
 - Use Remote Procedure Calls (RPC) to translate file system calls into remote requests
 - No local caching/can be caching at server-side
- Advantage: Server provides completely consistent view of file system to multiple clients
- Problems? Performance!
 - Going over network is slower than going to local memory
 - Lots of network traffic
 - Server can be a bottleneck

Use of caching to reduce network load



- Idea: Use caching to reduce network load
 - In practice: use buffer cache at source and destination
- Advantage: if open/read/write/close can be done locally, don't need to do any network traffic...fast!
- Problems:
 - Failure:
 - » Client caches have data not committed at server
 - Cache consistency!
 - » Client caches not consistent with server/each other

Dealing with Failures

- What if server crashes? Can client wait until it comes back and just continue making requests?
 - Changes in server's cache but not in disk are lost
- What if there is shared state across RPC's?
 - Client opens file, then does a seek
 - Server crashes
 - What if client wants to do another read?
- Similar problem: What if client removes a file but server crashes before acknowledgement?

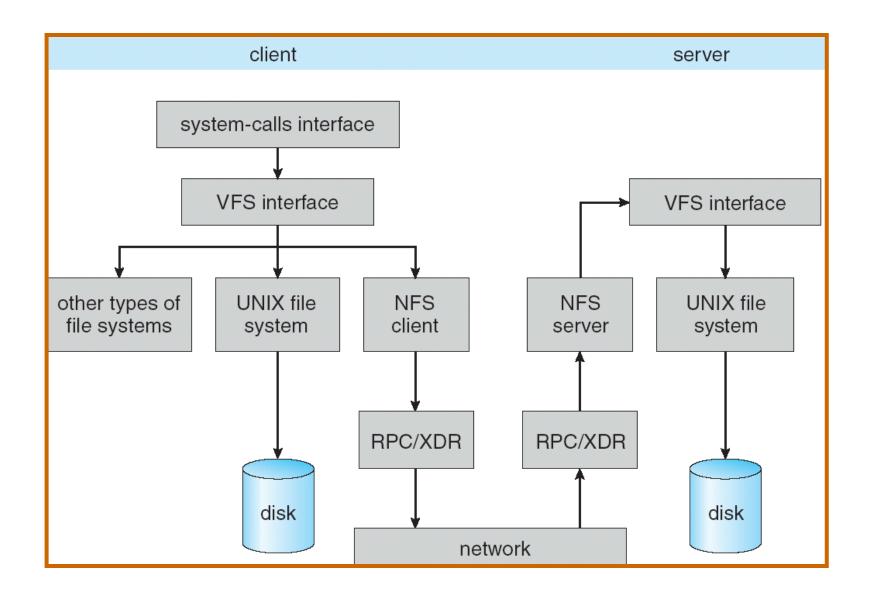
Stateless Protocol

- A protocol in which all information required to service a request is included with the request
- Even better: Idempotent Operations repeating an operation multiple times is same as executing it just once (e.g., storing to a mem addr.)
- Client: timeout expires without reply, just run the operation again (safe regardless of first attempt)
- Recall HTTP: Also a stateless protocol
 - Include cookies with request to simulate a session

Network File System (Sun)

- Defines an RPC protocol for clients to interact with a file server
 - E.g., read/write files, traverse directories, . . .
 - Stateless to simplify failure cases
- Keeps most operations idempotent
 - Even removing a file: Return advisory error second time
- Don't buffer writes on server side cache
 - Reply with acknowledgement only when modifications reflected on disk

NFS Architecture



Network File System (NFS)

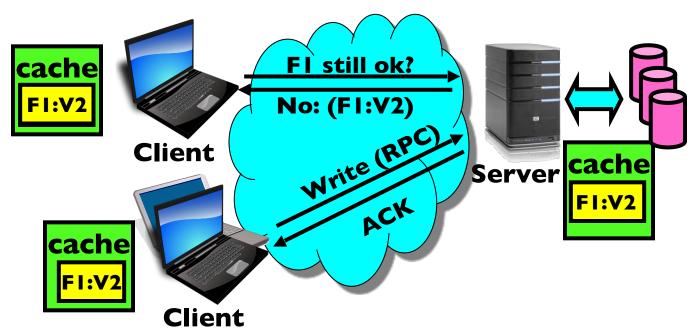
- Three Layers for NFS system
 - UNIX file-system interface: open, read, write, close calls + file descriptors
 - VFS layer: distinguishes local from remote files
 - » Calls the NFS protocol procedures for remote requests
 - NFS service layer: bottom layer of the architecture
 - » Implements the NFS protocol
- NFS Protocol: RPC for file operations on server
 - Reading/searching a directory
 - manipulating links and directories
 - accessing file attributes/reading and writing files
- Write-through caching: Modified data committed to server's disk before results are returned to the client
 - lose some of the advantages of caching
 - time to perform write() can be long
 - Need some mechanism for readers to eventually notice changes! (more on this later)

NFS Continued

- NFS servers are stateless; each request provides all arguments require for execution
 - E.g. reads include information for entire operation, such as
 ReadAt (inumber, position), not Read (openfile)
 - No need to perform network open() or close() on file each operation stands on its own
- Idempotent: Performing requests multiple times has same effect as performing it exactly once
 - Example: Server crashes between disk I/O and message send, client resend read, server does operation again
 - Example: Read and write file blocks: just re-read or re-write file block no side effects
 - Example: What about "remove"? NFS does operation twice and second time returns an advisory error
- Failure Model: Transparent to client system
 - Is this a good idea? What if you are in the middle of reading a file and server crashes?
 - Options (NFS Provides both):
 - » Hang until server comes báck up (next week?)
 - » Return an error. (Of course, most applications don't know they are talking over network)

NFS Cache consistency

- NFS protocol: weak consistency
 - Client polls server periodically to check for changes
 - » Polls server if data hasn't been checked in last 3-30 seconds (exact timeout it tunable parameter).
 - » Thus, when file is changed on one client, server is notified, but other clients use old version of file until timeout.



- What if multiple clients write to same file?
 - » In NFS, can get either version (or parts of both)
 - » Completely arbitrary!

Sequential Ordering Constraints

- What sort of cache coherence might we expect?
 - i.e. what if one CPU changes file, and before it's done, another CPU reads file?
- Example: Start with file contents = "A"

Client 1:

Client 2:

Client 3:

Read: gets A or B Write C

Read: parts of B or C

Read: parts of B or C

Time

- What would we actually want?
 - Assume we want distributed system to behave exactly the same as if all processes are running on single system
 - » If read finishes before write starts, get old copy
 - » If read starts after write finishes, get new copy
 - » Otherwise, get either new or old copy
 - For NFS:
 - » If read starts more than 30 seconds after write, get new copy; otherwise, could get partial update

Andrew File System

- Andrew File System (AFS, late 80's) → DCE DFS (commercial product)
- Callbacks: Server records who has copy of file
 - On changes, server immediately tells all with old copy
 - No polling bandwidth (continuous checking) needed
- Write through on close
 - Changes not propagated to server until close()
 - Session semantics: updates visible to other clients only after the file is closed
 - » As a result, do not get partial writes: all or nothing!
 - » Although, for processes on local machine, updates visible immediately to other programs who have file open
- In AFS, everyone who has file open sees old version
 - Don't get newer versions until reopen file

Summary (1/3)

- TCP: Reliable byte stream between two processes on different machines over Internet (read, write, flush)
 - Uses window-based acknowledgement protocol
 - Congestion-avoidance dynamically adapts sender window to account for congestion in network
- Remote Procedure Call (RPC): Call procedure on remote machine or in remote domain
 - Provides same interface as procedure
 - Automatic packing and unpacking of arguments without user programming (in stub)
 - Adapts automatically to different hardware and software architectures at remote end

Summary (2/3)

- Distributed File System:
 - Transparent access to files stored on a remote disk
 - Caching for performance
- VFS:Virtual File System layer
 - Provides mechanism which gives same system call interface for different types of file systems
- Cache Consistency: Keeping client caches consistent with one another
 - If multiple clients, some reading and some writing, how do stale cached copies get updated?
 - NFS: check periodically for changes
 - AFS: clients register callbacks to be notified by server of changes

Summary (3/3)

Key-Value Store:

- Two operations
 - » put(key, value)
 - » value = get(key)
- Challenges
 - » Scalability → serve get()'s in parallel; replicate/cache hot tuples
 - » Fault Tolerance → replication
 - » Consistency → quorum consensus to improve put() performance