Distributed systems

Lecture 6: Elections, consensus, and distributed transactions

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Last time

- Saw how we can build ordered multicast
 - Messages between processes in a group
 - Need to distinguish receipt and delivery
 - Several ordering options: FIFO, causal or total
- Considered distributed mutual exclusion:
 - Want to limit one process to a CS at a time
 - Central server OK; but bottleneck & SPoF
 - Token passing OK: but traffic, repair, token loss
 - Totally-Ordered Multicast: OK, but high number of messages and problems with failures

Slide from last lecture – clarification: the Lamport variant below is an example of using multicast for total ordering, rather than totally ordered multicast

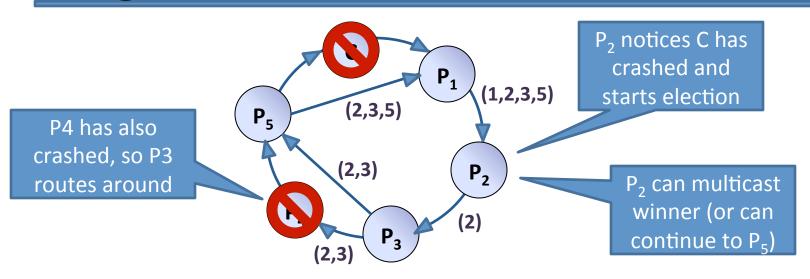
Additional Details

- Completely unstructured decentralized solution ... but:
 - Lots of messages (1 multicast + N-1 unicast)
 - Ok for most recent holder to re-enter CS without any messages
- Variant scheme (due to Lamport):
 - To enter, process P_i multicasts request(P_i, T_i) [same as before]
 - On receipt of a message, P_i replies with an ack(P_i,T_i)
 - Processes keep all requests and acks in ordered queue
 - If process P_i sees his request is earliest, can enter CS ... and when done, multicasts a release(P_i, T_i) message
 - When P_i receives release, removes P_i's request from queue
 - If P_i's request is now earliest in queue, can enter CS...
- Note that both Ricart & Agrawala and Lamport's scheme, have N points of failure: doomed if any process dies:-(

Leader Election

- Many schemes are built on the notion of having a welldefined 'leader' (master, coordinator)
 - examples seen so far include the Berkeley time synchronization protocol, and the central lock server
- An election algorithm is a dynamic scheme to choose a unique process to play a certain role
 - assume P_i contains state variable elected_i
 - when a process first joins the group, elected;
 UNDEFINED
- By the end of the election, for every P_i,
 - elected_i = P_x , where P_x is the winner of the election, or
 - elected_i = UNDEFINED, or
 - P_i has crashed or otherwise left the system

Ring-Based Election

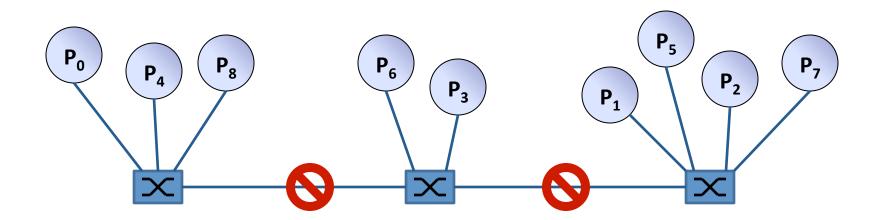


- System has coordinator who crashes
- Some process notices, and starts an election
 - Find node with highest ID who will be new leader
 - Puts its id into a message, and sends to its successor
 - On receipt, a process acks to sender (not shown), and then appends its
 id and forwards the election message
 - Finished when a process receives message containing its id

The Bully Algorithm

- Algorithm proceeds by attempting to elect the process still alive with the highest ID
 - Assume that we know the IDs of all processes
 - Assumes we can reliably detect failures by timeouts
- If process P_i sees current leader has crashed, sends election message to all processes with higher IDs, and starts a timer
 - Concurrent election initiation by multiple processes is fine
 - Processes receiving an election message reply OK to sender, and start an election of their own (if not already in progress)
 - If a process hears nothing back before timeout, it declares itself the winner, and multicasts result
- A dead process that recovers (or new process that joins) also starts an election: can ensure highest ID always elected

Problems with Elections



- Algorithms rely on timeouts to reliably detect failure
- However it is possible for networks to fail: a network partition
 - Some processes can speak to others, but not all
- Can lead to split-brain syndrome:
 - Every partition independently elects a leader → too many bosses!
- To fix, need some secondary (& tertiary?) communication scheme
 - e.g. secondary network, shared disk, serial cables, ...

Aside: Consensus

- Elections are a specific example of a more general problem: consensus
 - Given a set of n processes in a distributed system, how can we get them all to agree on something?
- Classical treatment has every process P_i propose something (a value V_i)
 - Want to arrive at some deterministic function of V_i's (e.g. 'majority' or 'maximum' will work for election)
- A correct solution to consensus must satisfy:
 - Agreement: all nodes arrive at the same answer
 - Validity: answer is one that was proposed by someone
 - Termination: all nodes eventually decide

"Consensus is impossible"

- Famous result due to Fischer, Lynch & Patterson (1985)
 - Focuses on an asynchronous network (unbounded delays) with at least one process failure
 - Shows that it is possible to get an infinite sequence of states, and hence never terminate
 - Given the Internet is an asynchronous network, then this seems to have major consequences!!
- Not really:
 - Result actually says we can't always guarantee consensus,
 not that we can never achieve consensus
 - And in practice, we can use tricks to mask failures (such as reboot, or replication), and to ignore asynchrony
 - Have seen solutions already, and will see more later

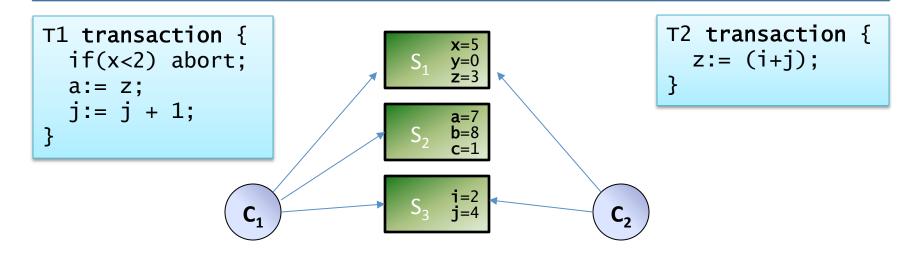
Transaction Processing Systems

- Last term looked at transactions:
 - ACID properties
 - Support for composite operations (i.e. a collection of reads and updates to a set of objects)
- A transaction is atomic ("all-or-nothing")
 - If it commits, all operations are applied
 - If it aborts, it's as if nothing ever happened
- A committed transaction moves system from one consistent state to another
- Transaction processing systems also provide:
 - isolation (between concurrent transactions)
 - durability (committed transactions survive a crash)

Distributed Transactions

- Scheme described last term was client/server
 - E.g., a program (client) accessing a database (server)
- However distributed transactions are those which span multiple transaction processing servers
- E.g. booking a complex trip from London to Vail, CO
 - Could fly LHR -> LAX -> EGE + hire a car...
 - ... or fly LHR -> ORD -> DEN + take a public bus
- Want a complete trip (i.e. atomicity)
 - Not get stuck in an airport with no onward transport!
- Must coordinate actions across multiple parties

A Model of Distributed Transactions



- Multiple servers (S₁, S₂, S₃, ...), each holding some objects which can be **read** and **written** within client transactions
- Multiple concurrent clients $(C_1, C_2, ...)$ who perform transactions that interact with one or more servers
 - e.g. T1 reads \mathbf{x} , \mathbf{z} from S_1 , writes \mathbf{a} on S_2 , and reads & writes \mathbf{j} on S_3
 - e.g. T2 reads \mathbf{i} , \mathbf{j} from S_3 , then writes \mathbf{z} on S_1
- A successful commit implies agreement at all servers

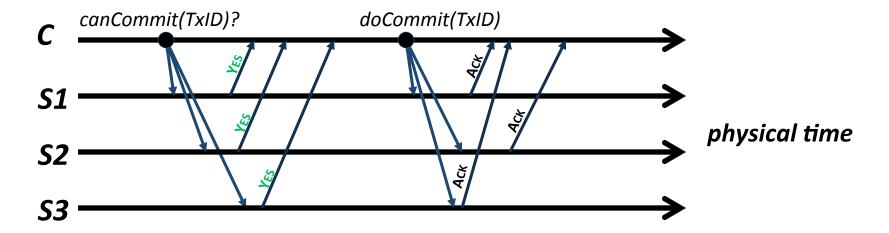
Implementing distributed transactions

- Can build on top of solution for single server:
 - e.g. use locking or shadowing to provide isolation
 - e.g. use write-ahead log for durability
- Need to coordinate to either commit or abort
 - Assume clients create unique transaction ID: TxID
 - Uses TxID in every read or write request to a server S_i
 - First time S_i sees a given TxID, it starts a tentative transaction associated with that transaction id
 - When client wants to commit, must perform atomic commit of all tentative transactions across all servers

Atomic Commit Protocols

- A naïve solution would have client simply invoke commit(TxID) on each server in turn
 - Will work only if no concurrent conflicting clients, every server commits (or aborts), and no server crashes
- To handle concurrent clients, introduce a coordinator:
 - A designated machine (can be one of the servers)
 - Clients ask coordinator to commit on their behalf... and hence coordinator can serialize concurrent commits
- To handle inconsistency/crashes, the coordinator:
 - Asks all involved servers if they could commit TxID
 - Servers S_i reply with a vote V_i = { COMMIT, ABORT }
 - If all V_i = COMMIT, coordinator multicasts doCommit(TxID)
 - Otherwise, coordinator multicasts doAbort(TxID)

Two-Phase Commit (2PC)



- This scheme is called two-phase commit (2PC):
 - First phase is voting: collect votes from all parties
 - Second phase is completion: either abort or commit
- Doesn't require ordered multicast, but needs reliability
 - If server fails to respond by timeout, treat as a vote to abort
- Once all Acks received, inform client of successful commit

2PC: Additional Details

- Client (or any server) can abort during execution: simply multicasts doAbort(TxID) to all servers
 - E.g., if client transaction throws exception or server fails
- If a server votes NO, can immediately abort locally
- If a server votes YES, it must be able to commit if subsequently asked by coordinator:
 - Before voting to commit, server will prepare by writing entries into log and flushing to disk
 - Also records all requests from & responses to coordinator
 - Hence even if crashes after voting to commit, will be able to recover on reboot

2PC: Coordinator Crashes

- Coordinator must also persistently log events:
 - Including initial message from client, requesting votes, receiving replies, and final decision made
 - Lets it reply if (restarted) client or server asks for outcome
 - Also lets coordinator recover from reboot, e.g. re-send any vote requests without responses, or reply to client
- One additional problem occurs if coordinator crashes after phase 1, but before initiating phase 2:
 - Servers will be uncertain of outcome...
 - If voted to commit, will have to continue to hold locks, etc.
- Other schemes (3PC, Paxos, ...) can deal with this

Replication

- Many distributed systems involve replication
 - Multiple copies of some object stored at different servers
 - Multiple servers capable of providing some operation(s)
- Three key advantages:
 - Load-Balancing: if have many replicas, then can spread out work from clients between them
 - Lower Latency: if replicate an object/server close to a client, will get better performance
 - Fault-Tolerance: can tolerate the failure of some replicas and still provide service
- Examples include DNS, web & file caching (& contentdistribution networks), replicated databases, ...

Replication in a Single System

- A good single-system example is RAID:
 - RAID = Redundant Array of Inexpensive Disks
 - Disks are cheap, so use several instead of just one
 - If replicate data across disks, can tolerate disk crash
 - If don't replicate data, appearance of a single larger disk
- A variety of different configurations (levels)
 - RAID 0: stripe data across disks, i.e. block 0 to disk 0, block 1 to disk 1, block 2 to disk 0, and so on
 - RAID 1: mirror (replicate) data across disks, i.e. block 0 written on disk
 0 and disk 1
 - RAID 5: parity write block 0 to disk 0, block 1 to disk 1, and (block 0 xor block 1) to disk 2
- Get improved performance since can access disks in parallel
- With RAID 1, 5 also get fault-tolerance

Distributed Data Replication

- Have some number of servers (S₁, S₂, S₃, ...)
 - Each holds a copy of all objects
- Each client C_i can access any replica (any S_i)
 - e.g. clients can choose closest, or least loaded
- If objects are **read-only**, then trivial:
 - Start with one primary server P having all data
 - If client asks S_i for an object, S_i returns a copy
 - (S_i fetches a copy from P if it doesn't already have a fresh one)
- Can easily extend to allow updates by P
 - When updating object O, send invalidate(O) to all S_i
- In essence, this is how web caching / CDNs work today
- But what if clients can perform updates?

Replication and Consistency

- Gets more challenging if clients can perform updates
- For example, imagine x has value 3 (in all replicas)
 - C1 requests write(x, 5) from S4
 - C2 requests read(x) from S3
 - What should occur?
- With strong consistency, the distributed system behaves as if there is no replication present:
 - i.e. in above, C2 should get the value 5
 - requires coordination between all servers
- With weak consistency, C2 may get 3 or 5 (or ...?)
 - Less satisfactory, but much easier to implement

Next time

- Replication and consistency
 - Strong consistency
 - Quorum-based systems
 - Weaker consistency
- Consistency, availability and partitions
- Further replication models
- Google case studies