

Distributed systems

Lecture 6: Elections, distributed transactions, and replication

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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 & **Single Point of Failure (SPoF)**
 - Token passing OK: but traffic, repair, token loss
 - Totally-Ordered Multicast: OK, but high number of messages and problems with failures

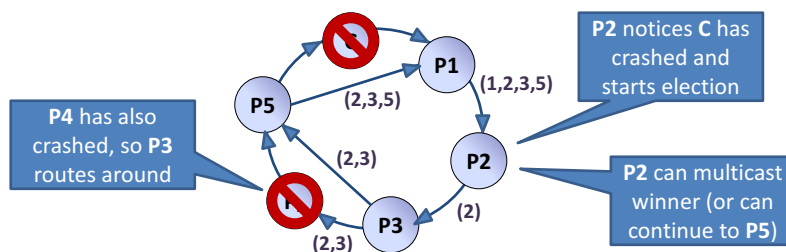
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Leader election

- Many schemes are built on the notion of having a well-defined '**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 **elect** $_i$
 - when a process first joins the group, **elect** $_i$ = UNDEFINED
- By the end of the election, for every P_i ,
 - **elect** $_i$ = P_x , where P_x is the winner of the election, or
 - **elect** $_i$ = UNDEFINED, or
 - P_i has crashed or otherwise left the system
- Common theme: **live node with the highest ID wins**

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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

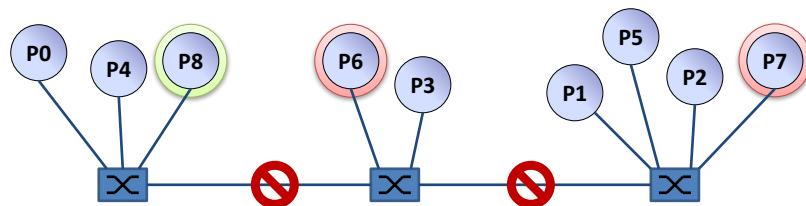
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The Bully Algorithm

- 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

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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, ...
- This is important because we want to implement distributed algorithms dependent on **the invariant that the leader is unique**

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Aside on 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

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“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

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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)
- Q: Can we bring the {scalability, fault tolerance, ...} benefits of distributed systems to transaction processing?

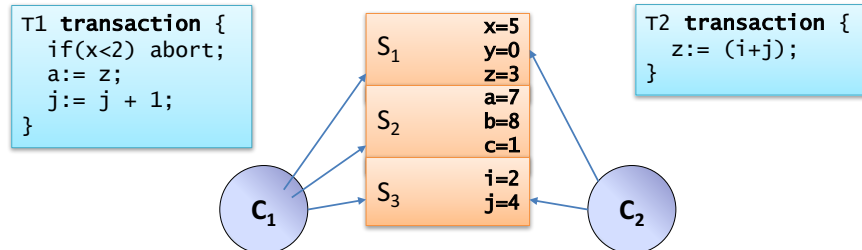
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Distributed transactions

- 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

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A model of distributed transactions



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

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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

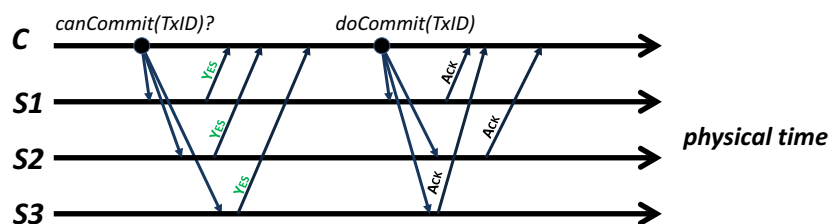
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Atomic commit protocols

- 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 $V_i = \{ \text{COMMIT}, \text{ABORT} \}$
 - If all $V_i = \text{COMMIT}$, coordinator multicasts **doCommit(TxID)**
 - Otherwise, coordinator multicasts **doAbort(TxID)**

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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

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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

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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

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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 (& content-distribution networks), replicated databases, ...

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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

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Distributed data replication

- Have some number of servers (S_1, S_2, S_3, \dots)
 - 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?**

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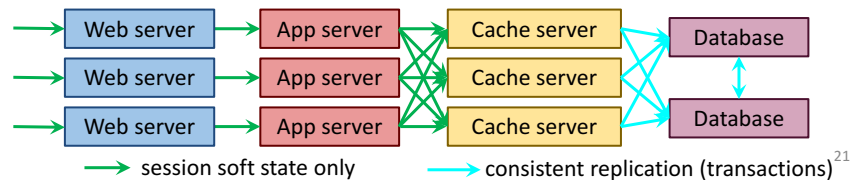
Replication and consistency

- 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

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Replication for fault tolerance

- Replication for services, not just data objects
- Easiest is for a **stateless services**:
 - Simply duplicate functionality over k machines
 - Clients use any (e.g. closest), fail over to another
- Very few totally stateless services
 - But e.g. many web apps have per-session soft state
 - State generated per-client, lost when client leaves
- For example: multi-tier web farms (Facebook, ...):



Passive replication

- A solution for stateful services is **primary/backup**:
 - Backup server takes over in case of failure
- Based around **persistent logs** and **system checkpoints**:
 - Periodically (or continuously) checkpoint primary
 - If detect failure, start backup from checkpoint
- A few variants trade-off fail-over time:
 - **Cold-standby**: backup server must start service (software), load checkpoint & parse logs
 - **Warm-standby**: backup server has software running in anticipation – just needs to load primary state
 - **Hot-standby**: backup server mirrors primary work, but output is discarded; on failure, enable output

Active replication

- **Alternative:** have k replicas running at **all** times
- Front-end server acts as an **ordering node**:
 - Receives requests from client and forwards them to all replicas using **totally ordered multicast**
 - Replicas each perform operation and respond to front-end
 - Front-end gathers responses, and replies to client
- Typically require replicas to be “**state machines**”:
 - i.e. act deterministically based on input
 - Idea is that all replicas operate ‘in lock step’
- **Active replication** is expensive (in terms of resources)...
 - ... and not really worth it in the common case.
 - However valuable if consider **Byzantine failures**

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Summary + next time

- Leader elections + distributed consensus
- Distributed transactions + atomic commit protocols
- Replication + consistency
- (More) replication and consistency
 - Strong consistency
 - Quorum-based systems
 - Weaker consistency
- Consistency, availability and partitions
- Further replication models
- Start of Google case studies

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