#### CS 417 – DISTRIBUTED SYSTEMS

# Week 8: Distributed Transactions Part 2: Three-Phase Commit and the CAP Theorem

Paul Krzyzanowski

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Notes

# **Three-Phase Commit Protocol**

# What's wrong with the 2PC protocol?

Biggest problem: it's a blocking protocol with failure modes that require all systems to recover eventually

- If the coordinator crashes, participants have no idea whether to commit or abort
  - A recovery coordinator helps
- If a coordinator AND a participant crashes
  - The system has no way of knowing the result of the transaction
  - It might have committed at the crashed participant hence all others must block

# The protocol cannot pessimistically abort because some participants may have already committed

When a participant gets a commit/abort message, it does not know if every other participant was informed of the result

### **Three-Phase Commit Protocol**

- Same setup as the two-phase commit protocol:
  - Coordinator & Participants
- Add timeouts to each phase that result in an abort

- Propagate the result of the commit/abort vote to each participant <u>before</u> telling them to act on it
  - This will allow us to recover the state of the transaction from any participant and give more options for aborting

### **Three-Phase Commit Protocol**

#### Split the second phase of 2PC into two parts:

#### 2a. "Precommit" (prepare to commit) phase

- Send Prepare message to all participants when it received a yes from all participants in phase 1
- Participants can prepare to commit but cannot do anything that cannot be undone
- Participants reply with an acknowledgement
- <u>Purpose</u>: let every participant know the state of the result of the vote so that state can be recovered if anyone dies

#### 2b. "Commit" phase (same as in 2PC)

- If coordinator gets ACKs for all *prepare* messages
  - It will send a *commit* message to all participants
- Else it will abort send an *abort* message to all participants

### Three-Phase Commit Protocol: Phase 1

#### Phase 1: Voting phase

Coordinator sends *CanCommit?* queries to participants & gets responses Purpose: Find out if everyone agrees to commit

- [!] If the coordinator gets a *timeout* from any participant or any "No" replies are received
  - Send an *abort* to all participants
- [!] If a participant times out waiting for a request from the coordinator
  - It aborts itself (assume coordinator crashed)
- Else continue to phase 2

#### We can abort if the participant and/or coordinator dies

### **Three-Phase Commit Protocol**

#### Phase 2: Prepare to commit phase

- Send a *prepare* message to all participants
- Get OK messages from all participants
  - We need to hear from <u>all</u> before proceeding so we can be sure the state of the protocol can be properly recovered if the coordinator dies
- Purpose: let all participants know the decision to commit
- [!] If a participant times out: assume it crashed; send *abort* to all participants

#### Phase 3: *Finalize phase*

- Send commit messages to participants and get responses from all
- [!] If participant times out: contact any other participant and move to that state (commit or abort)
- [!] If coordinator times out: that's ok we know what to do

### **3PC Recovery**

#### If the coordinator crashes

A recovery node can query the state from any available participant

#### Possible states that the participant may report:

#### **Already committed**

- That means that every other participant has received a Prepare to Commit
- Some participants may have committed
  - ⇒ Send *Commit* message to all participants (just in case they didn't get it)

#### Not committed but received a Prepare message

- That means that all participants agreed to commit; some may have committed
- Send Prepare to Commit message to all participants (just in case they didn't get it)
- Wait for everyone to acknowledge; then commit

#### Not yet received a Prepare message

- This means no participant has committed; some may have agreed
- Transaction can be aborted or the commit protocol can be restarted

### **3PC Weaknesses**

- May have problems when the network gets partitioned
  - Partition A: nodes that received *Prepare* message
    - Recovery coordinator for A: allows commit
  - Partition B: nodes that did not receive *Prepare* message
    - Recovery coordinator for B: aborts
  - Either of these actions are legitimate as a whole



- But when the network merges back, the system will be inconsistent
- Not good when a crashed coordinator recovers
  - It needs to find out that someone else took over and stay quiet
  - Otherwise, it will mess up the protocol, leading to an inconsistent state

### 3PC coordinator recovery problem

Suppose a coordinator sent a Prepare message to all participants

- All participants acknowledged the message
- BUT the coordinator died before it got all acknowledgements
- A recovery coordinator queries a participant
  - It continues with the commit: Sends Prepare, gets ACKs, sends Commit
- Around the same time...the original coordinator recovers
  - Realizes it is still missing some replies from the Prepare
  - Gets timeouts from some and decides to send an Abort to all participants
- Some processes may commit while others abort!
- 3PC works well when servers crash (fail-stop model)
- But ...
  - 3PC is not resilient against fail-recover environments
  - 3PC is not resilient against network partitions
  - Also, 3PC involves an extra round of messages vs.  $2PC \rightarrow extra latency!$

# **Consensus-based Commit**

# What about Raft? Didn't it give us consensus?

- Consensus-based protocols (Raft, Paxos) are designed to be resilient against network partitions
- But consensus protocols are designed to solve a different problem!
  - Majority agreement makes sense in replicated state machines, not in distributed transactions, where each sub-transaction has different responsibilities
- What does Raft/Paxos consensus offer?
  - Total ordering of proposals (replicated log)
  - Fault tolerance: a proposal is accepted only if a majority of nodes accept it
    - This allows recovery of the decision even if some nodes die & others come up
  - Is provably resilient in asynchronous networks
- For a two-phase commit protocol to use a consensus algorithm:
  Turn the coordinator into a fault-tolerant replicated state machine
  - Use replicated nodes to avoid blocking if the coordinator fails
  - Run a consensus algorithm on the commit/abort decision of **EACH** participant

# What do we want to do with a consensus protocol?

- Each participant must get its chosen value can\_commit or must\_abort
  - accepted by the majority of replicated nodes

- Transaction Leader
  - Chosen via an election algorithm
  - Coordinates the commit algorithm
  - Not a single point of failure we can elect a new one; Raft nodes store state

### How do we do it?

- Some participant decides to begin to commit
  - Sends a message to the Transaction Leader
- Transaction Leader: Sends a *prepare* message to each participant
- Each participant now sends a can\_commit or must\_abort message to its instance of the consensus protocol
  - All participants share the elected Transaction Leader
  - "Can\_commit" or "Must\_abort" is sent to majority of followers
  - Result is sent to the leader
- Transaction Leader tracks all instances of the commit protocol
  - Commit iff every participant's instance of the consensus protocol chooses "can\_commit"
  - Tell each participant to **commit** or **abort**

#### Consensus-based fault-tolerant coordinator

#### The cast:

- One instance of Raft per participant (N participants)
- Set of 2F+1 nodes and a leader play the role of the coordinator
  - We can withstand the failure of F nodes
  - Leader = node elected to be in charge & run the protocol



- A leader will get at least F+1 messages for each instance
- Commit iff every participant's instance of Raft chooses can commit
- · Raft commit acts like 2PC if only one node

#### Virtual Synchrony vs. Transactions vs. Raft

#### Virtual Synchrony

- Fast & scalable
- Atomic multicast of messages to the entire group designed for state machine replication
- Focuses on group membership management & atomic multicasts
- Does not handle partitions!

#### Two-Phase & Three-Phase Commit

- Most expensive requires extensive use of stable storage
- 2PC is efficient in terms of # of messages designed for transactional activities
- Not suitable for high-speed or continuous messaging

#### Raft or Paxos Consensus

- General purpose fault-tolerant consensus algorithm *designed for state machine replication*
- Not designed for transactions: relies on a majority of systems being up; no concept of abort
- Performance usually limited need to get majority acceptance and Raft requires stable storage
- Useful for fault-tolerant log replication & elections
- Using consensus-based commit overcomes dead coordinator and network partition problems of 2PC and 3PC
  - But the transaction coordinator at each participant will be a replicated state machine high overhead
- Need mechanisms to restore state on *abort*

# Scaling & Consistency

#### Reliance on multiple systems affects availability

- One database with 99.9% availability
  - 8 hours, 45 minutes, 35 seconds downtime per year
- If a transaction uses 2PC protocol and requires two databases, each with a 99.9% availability:
  - Total availability = (0.999\*0.999) = 99.8%
  - 17 hours, 31 minutes, 12 seconds downtime per year
- If a transaction requires 5 databases:
  - Total availability = 99.5%
  - 1 day, 19 hours, 48 minutes, 0 seconds downtime per year

### Scaling Transactions

- Transactions require locking part of the database so that everyone sees consistent data at all times
  - Good on a small scale
    - Low transaction volumes: getting multiple databases consistent is easy
  - Difficult to do efficiently on a huge scale
- Add replication processes can read any replica
  - But all replicas must be locked during updates to ensure consistency
- Risks of not locking:
  - Users run the risk of seeing stale data
  - The "I" of ACID may be violated
    - E.g., two users might try to buy the last book on Amazon

The delays to achieve consistency can hurt business

- Amazon: 0.1 second increase in response time costs 1% of sales
- Google: 0.5 second increase in latency causes traffic to drop by 20%
- Latency is due to lots of factors
  - OS & software architecture, computing hardware, tight vs. loose coupling, network links, geographic distribution, ...
  - We're only looking at the problems caused by the tight software coupling due to achieving the ACID model

http://highscalability.com/latency-everywhere-and-it-costs-you-sales-how-crush-it http://www.julianbrowne.com/article/viewer/brewers-cap-theorem

### Eric Brewer's CAP Theorem

Three core requirements in a shared data system:

- 1. Atomic, Isolated Consistency
  - Operations must appear totally ordered and each is isolated
- 2. Availability
  - Every request received by a non-failed node must result in a response
- 3. Partition Tolerance: tolerance to network partitioning Messages between nodes may be lost

No set of failures less than total failure is allowed to cause the system to respond incorrectly

# **CAP Theorem:** when there is a network partition, you cannot guarantee both availability & consistency

Commonly (not totally accurately) stated as you can have at most two of the three: C, A, or P

# **Example:** Partition

time



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# Giving up one of {C, A, P}

#### Ensure partitions never occur

- Put everything on one machine or a cluster in one rack: high availability clustering
- Use two-phase commit or three phase commit
- Scaling suffers
- Give up availability [system is consistent & can handle partitioning]
  - Lock data: have services wait until data is consistent
  - Classic ACID distributed databases (also 2PC)
  - Response time suffers

We <u>really</u> want partition tolerance & high availability for a distributed system!

- Give up consistency [system is available & can handle partitioning]
  - Eventually consistent data
  - Use expirations/leases, queued messages for updates
  - Often not as bad as it sounds!
  - Examples: DNS, web caching, Amazon Dynamo, Cassandra, CouchDB

### Partitions will occur

- With distributed systems, we expect partitions to occur
  - Maybe not a true partition but high latency can act like a partition
  - This is a property of the distributed environment
  - The CAP theorem says we have a tradeoff between availability & consistency
- But we want availability and consistency
  - We get availability via replication
  - We get consistency with atomic updates
    - 1. Lock all copies before an update
    - 2. Propagate updates
    - 3. Unlock
- We can choose high availability: allow reads before all nodes are updated (avoid locking)

... or choose consistency: enforce proper locking of nodes for updates

#### **Eventual Consistency Model**

- Traditional database systems want ACID
  - But scalability is a problem (lots of transactions in a distributed environment)
- Give up Consistent and Isolated

in exchange for high availability and high performance

- Get rid of locking in exchange for multiple versions
- Incremental replication
- BASE = Basically Available Soft-state Eventual Consistency

#### Eventual consistency model:

If no updates are made to a data item, *eventually* all accesses to that item will return the last updated value

# ACID vs. BASE

#### ACID

- Strong consistency
- Isolation
- Focus on commit
- Nested transactions
- Availability can suffer
- Pessimistic access to data (locking)

From Eric Brewer's PODC Keynote, July 2000 http://www.cs.berkeley.edu/~brewer/cs262b-2004/PODC-keynote.pdf

#### BASE

- Weak (eventual) consistency: stale data at times
- High availability
- Best effort approach
- Optimistic access to data
- Simpler model (but harder for app developer)
- Faster

# A place for BASE

- ACID is neither dead nor useless
  - Many environments require it
  - It's safer the framework handles ACID for you
- BASE has become common for large-scale web apps where replication & fault tolerance is crucial
  - eBay, Twitter, Amazon
  - Eventually consistent model not always surprising to users
    - Cellphone usage data
    - Banking transactions (e.g., fund transfer activity showing up on statement)
    - Posting of frequent flyer miles

But ... the app developer has to worry about update conflicts and reading stale data ... and programmers often write buggy code

# The End

# The End