Today’s topic

- How do cloud systems actually use transactions?
- Last time we saw the basic transactional model.
  - But as we saw from reviewing Brewer’s CAP theorem and the BASE methodology, transactions are sometimes too expensive and not scalable enough
- This has led to innovations on the transaction side
  - Snapshot isolation (related to serializability and ACID)
  - Business transactions (related to BASE)
This idea started with discussion about lock-based (pessimistic) concurrency control in comparison with timestamp-based concurrency control.

- With locking we incur high costs to obtain one lock at a time. In distributed settings these costs are prohibitive.
  - Deadlock is a risk, must use a deadlock avoidance scheme.
- With timestamped concurrency control, we just pick a time at which transactions will run.
  - If times are picked to be unique, progress guaranteed because some transaction will have the smallest TS and won’t abort. But others may abort and be forced to retry.
Each scheme attracted a following

- Locking is easy to design and works well if transactions do a great deal of updates/writes
- But 2PC can be costly if transactions are doing mostly reads and few writes

- In contrast, timestamp schemes work very well for read-mostly or pure-read workloads and do a lot of rollback if a workload has a mixture
Snapshots isolation

- Arose from database products that offered “multiversion” data
  - Popular in the cloud, because we sometimes don’t want to throw anything away
  - Each transaction can be seen as moving the database from a consistent state to a new consistent state

\[ T_1 \rightarrow T_2 \rightarrow T_3 \rightarrow T_5 \]

\{A=2,B=7,C=4\} \rightarrow \{B=8,D=3\} \rightarrow \{C=0\} \rightarrow \{A=25,D=99\}
A multiversion database

Instead of just keeping the value of the variables in the database, we track each revision and when the change was committed.

| A  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 025 | 0 |
| B  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| C  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| D  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 99 |

{T1: 10:02.421} {A=2,B=7,C=4} {T2: 10:03.006} {B=8,D=3} {T3: 10:04.521} {C=0} {T5: 10:08.571} {A=25,D=99}
Snapshot isolation idea

- For a read transaction, just pick a time at which the reads should be executed (ideally, a recent time corresponding to the commit of some transaction)
  - If transactions really take us from consistent state to consistent state, this will be a “safe” time to execute
  - Reads don’t change the state so execute without risk of needing to abort
- Then use locking to execute transactions that need to perform update operations
Fancier snapshot isolation

- Often used for all reads, not just read-only transactions
- Runs dynamically: Instead of picking just one time at which to run, pick a “range” of times and track it
- A single window is used even if X accesses many variables
... pick a “range” of times and track it

- E.g. transaction X might initially pick time range [0...NOW]

- As X actually accesses variables, narrow the time window of the transaction \([\max(\text{old start}, \text{new start}), \min(\text{old end}, \text{new end})]\)

  - E.g. X tries to read variable A and because A is locked for update by transaction Y, reads A=2
  - A=2 was valid from time [10:02.421, 10:08.57]

  - This narrows the window of validity for transaction X
How can a window vanish?

- Occurs if there just isn’t any point in the serialization order at which this set of reads could have happened.

- Result of an update that invalidates some past read.

- Causes transaction to abort.
Complications

- In fact, snapshot isolation doesn’t guarantee full serializability
  - An update transaction might “invalidate” a read by updating A at an unexpectedly early time
  - Unless we check the read-only transactions won’t know which ones to abort
  - Real issue: X may already have finished
- If we use s.o. for reads in read/write transactions, we get additional “bad cases”
Snapshot isolation is widely used

- Works well with multitier cloud computing infrastructures
  - Caching structures that track validity intervals for cached variables are common
  - Several papers have shown how to make snapshot isolation fully serializable, but methods haven’t been widely adopted (and may never be)

- Fits nicely with BASE: Basically available, soft state replication with eventual consistency
  - Often we don’t worry about consistency for the client
Consistency: Two “views”

- Client sees a snapshot of the database that is internally consistent and “might” be valid.

- Internally, database is genuinely serializable, but the states clients saw aren’t tracked and might sometimes become invalidated by an update.

- Inconsistency is tolerated because it yields such big speedups, although some clients see “wrong” results.
Do clients need perfect truth?

- If so, one recent idea is to “validate” at commit time
  - Many systems have a core transactional system that does updates
  - Collections of read-only cached replicas are created at the edge where clients reside
  - Read-only transactions run on these (true) replicas, with no risk of error
  - Read/write transactions track the versions read and the changes they “want” to make (intentions list)
- Then package these intended changes as ultra-fast transactions to be sent to the core system
  - It checks that these versions are still current, and if so, applies the updates, like in the Sinfonia system (discussed in class)
  - If not, transaction “aborts” and must be retried
- Effect is to soak up as much hard work as possible at the edge
A picture of how this works

1. Update transaction runs on cache first.
2. Simplified transaction lists versions to validate, then values to write for updates.
3. If successful, Core reports commit.

Read only transaction can safely execute on cache.
Core issue: How much contention?

- Root challenge is to understand
  - How many updates will occur
  - How often those updates conflict with concurrent reads or with concurrent updates

- In most of today’s really massive cloud applications, either contention is very rare, in which case transactional database solutions work, or we end up cutting corners and relaxing consistency
Tradeoff: Scale versus consistency

- With a core system we can impose strong consistency, but doing so limits scalability
  - It needs to “validate” every update
  - At some point it will get overloaded

- But if we don’t use a core system we can’t guarantee consistency
  - We may be able to design the application to tolerate small inconsistencies. Many web systems work this way
Are there other options?

- How does this approach compare with scalable replication using Paxos or Virtual Synchrony?

- In those systems the “contention” related to the order in which multicasts were delivered

  - Virtual synchrony strives to find ways of weakening required ordering to gain performance
  
  - Paxos is like serializability: One size fits all. But this is precisely why Brewer ended up proposing CAP!
Business transactions

- The Web Services standards introduces (yet) another innovation in the space

- They define a standard transactional API for cloud computing, and this is widely supported by transactional products of all kinds

- But they also define what are called “business transactions”
Think of Expedia

- You book a trip to Costa Rica
  - Flight down involves two separate carriers
  - Fourteen nights in a total of three hotels
  - Rental car for six days, bus tours for the rest
  - Two rainforest tours, one with “zip line experience”
  - Dinner reservation for two on your friend’s birthday at the Inka Grill restaurant in San Jose
  - Travel insurance covering stomach ailments (costs extra)
  - Special “babysit your dog” service in Ithaca
Should this be one transaction?

- Traditionally the transactional community would have argued that cases like these are precisely what transactions were invented for.

- In practice... it makes little sense to use transactions:
  - Multiple services, perhaps with very distinct APIs (e.g. may just need to phone the Inka Grill directly).
  - Many ways to roll back if something goes wrong, like just cancelling the car reservation.
Concept of a business transaction

- Instead of a single transaction, models something like this as a whole series of separate transactions
  - Maybe in a few cases done as true transactions
  - But others might be done in business-specific ways

- The standard assumes that each has its own specialized rollback technology available

- It also requires a “reliable message queuing” system
Reliable message queuing

- Basically, email for programs
  - Like with normal email, can send messages to addresses and they will be held until read/deleted
  - Spooler is assumed to be highly available and reliable
  - Generally has some kind of multi-stage structure: spools messages near the sender until handed off to the server, and only deleted once safely logged
How this works

- Application “sends” a set of requests, like one email each
- Spooler accepts the set and executes them one by one, restarting any that are disrupted by crashes
- Handling of other kinds of failures ("Sorry sir, the restaurant is fully booked that night") is under programmatic control
  - You need to add details to tell the system what to do
  - It won’t know that the Mexicali Cafe is a fallback
We create a sequence of transactions and of the associated undo actions for each

- Spool the series of transactions, linked by a business-transaction-identifier
- As each is executed, the undo action is spooled but in a “disabled” state
- On commit of the final transaction in the sequence, the undo actions are deleted
- On abort, the undo actions are enabled and run as a kind of reverse business transaction
If our reservations go part-way through but then the dog-sitter step fails, we end up leaving the world in a kind of inconsistent state.

But soon after we run the undo actions and this reverses the problems we created.

Even if someone failed to get a reservation at Inka Grill because of your temporarily booked table, they won’t be so surprised when they try again in a few days and now a table is free.
“Consistency is much overrated”

- We hear this a lot lately

- But you also need to wonder... what about
  - Medical care systems that run on the Internet?
  - Google’s self-driving cars?
  - The smart power grid
If eBay (BASE) ran the power grid

- With BASE, control system could have “two voices”
- In physical infrastructure settings, consequences can be very costly

“Switch on the 50KV Canadian bus”

“Canadian 50KV bus going offline”
The big problem

- Scalable consistency is hard!
  - Not impossible... but harder than weak consistency, or no consistency.

- Today’s most profitable web ventures manage quite well with weak models like BASE
  - Run a lot of stuff in parallel
  - Replicate data when you get a chance, but no rush
  - Sweep any errors under the rug
Not everyone is focused on the same property

- Some care mostly about scale and performance
- Some need really rapid response times
- Some genuinely do need consistency, but even then the definition could include different notions of ordering and durability
- Some need dynamic membership and others don’t

No one-size-fits-all options here! But today’s cloud is optimized for CAP, NoSQL, BASE…
What happens tomorrow?

- Nobody can compete with the cloud “price point”
  - In modern technology, the cheapest solution always wins
  - It becomes the only option available
  - So everything migrates to the winner

- We’ve seen this again and again

- The cloud will win. You guys will build the winning solutions, and they will be cloud based!
Why is it hard to cloudify high assurance?

- Let’s look at Isis²

- A cloud-based high assurance story...

- Can we view it as a blueprint for cloud-scale resiliency of a kind the masses might adopt?
High assurance: Different perspectives

- A single platform has many kinds of “users”

  - Programmer: Depends on platform properties but treats implementation as a black box.
  - End user: Seeks confidence that the system is safe and that if it goes offline, a warning will appear.
  - Protocol designer: Uses formal specification and logic to prove implementation of protocols correct.
  - Datacenter operator: Requires scalability, elasticity, and guarantees that applications won’t disrupt shared resources.

- Each brings different objectives and requires different methods.
Examples of these perspectives

- The end-user (the doctor) wants the system to be trustworthy. Means different things for different use-scenarios.

- The developer (you) needs a way to reason about applications you build. “My code will work because…”

- The tool builder (me, or Leslie) needs to prove the protocols in Isis2 or Paxos correct. “Paxos is safe because…”

- The cloud computing vendor wants scalability without hassles. Doesn’t want instability or other issues.
We’ve seen several high assurance “stories”

- Paxos
- Virtual synchrony
- Transactions

In each case the cloud community says “too expensive” and even proves theorems like CAP

- But while “just say no” is easy, results are sometimes harmful.
- Must we accept a low-assurance cloud?

Applications that need high assurance are coming