Shopping Lists on the Cloud

Large Scale Distributed Systems

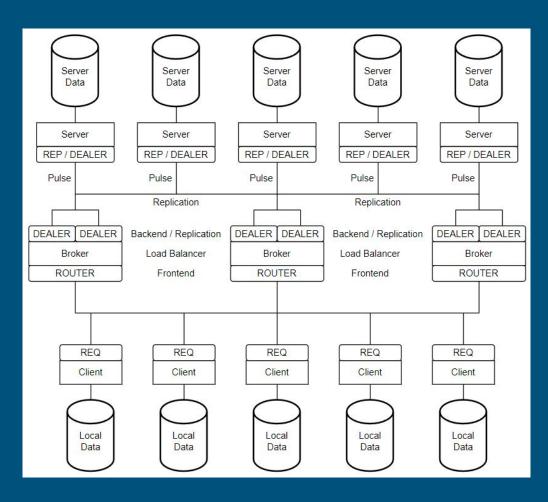
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Application Requirements

- Local-first shopping list application:
 - client-side persists data locally
 - cloud component data sharing and backup storage
 - application must be scalable
 - achieve 2 of the concepts from:
 - Consistency, Availability and Partition Tolerance (CAP)
- Shopping lists:
 - have unique IDs
 - o can be created or deleted by users
 - can be shared between users (using the ID)
- Shopping list items:
 - o can be added or removed by users
 - have quantities, which the users can update (quantity $0 \rightarrow$ item purchased)

System Architecture

- Decentralized distributed system
- 5 servers each with 20 virtual nodes
- 3 brokers
- "Pulse check" system
- **Data replication** through brokers
- Python zmq (**ZeroMQ**)
- **SQLite** database:
 - for each client
 - for each server
 - lists with removed flag
 - items with timestamps
 - knowledge of lists to be replicated
- TCP in all sockets
- Message serialization (JSON):
 - in all communications



Conflict Handling - Items

- LWW-Element-Set CRDT
- Add set added/updated items
- Remove set removed items
- Elements:
 - (name: (quantity, timestamp))
- Quantity update (add set):
 - o compare timestamps update vs item:
 - if update is more recent:
 - change quantity and timestamp of item

- Merge between CRTDs:
 - in each set, same items more recent timestamps win
 - item in both sets compare timestamps, find true state:
 - last action wins
- Major requirement:
 - precise/reliable timestamps for events comparison
 - avoid clock synchronization issues
 - universal clock in clients (NTP)

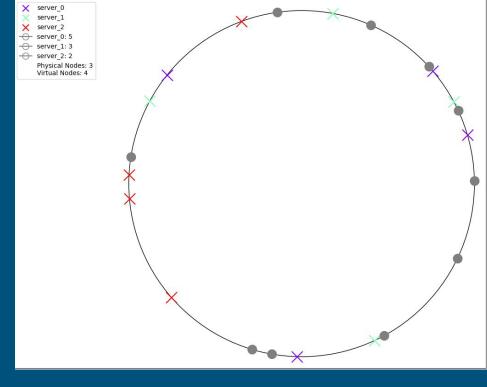
Conflict Handling - Lists

- OR-Set CRDT
- Add set created shopping lists
- Remove set deleted shopping lists
- Elements:
 - o (ID, name)
- Merge between CRDTs:
 - o server:
 - add sets union
 - remove sets union
 - o client:
 - unknown lists don't matter
 - remove sets union
 - remove intercepted with add set

- **Removal** always wins
- Data structure contains:
 - lists
 - respective items
- Merge process merges:
 - lists
 - respective items

Data Sharding

- Shopping lists distributed across servers
 - distributed system
- Each server represents a shard
- **Horizontal sharding** (by rows):
 - better scalability and load balancing
- Consistent hashing (SHA-256):
 - assignment of shopping lists to servers
 - directory-based sharding:
 - server's position on ring determines its responsibility
- Load balancing:
 - achieved through consistent hashing
 - target server (and backup nodes):
 - list ID applied to hashing ring



- Test configuration:
 - 3 Physical nodes
 - 4 Virtual Nodes
 - 10 shopping lists
 - 8 bits of salt

- All hashing done using SHA-256 (instead of 512)
 - Faster → less processing power
 - Efficient → less memory
 - Still provides good results (worse than 512)
 - 8 bits of salt used when hashing virtual nodes:
 - server_X_virtual_Y_SALT
- Physical Nodes:
 - + Resource efficiency
 - +Hotspots
- Virtual Nodes:
 - + Work distribution
 - + Dynamic scaling
 - + Spreading in the ring
- Too many Virtual Nodes:
 - + Computational overhead

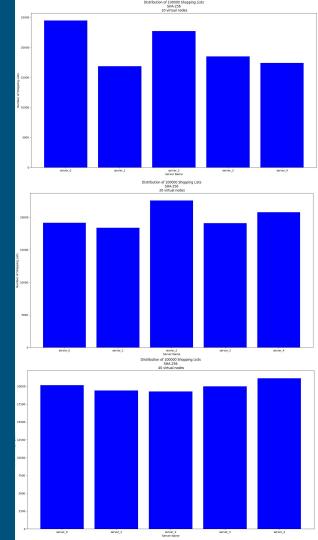
More Physical Nodes:

+ Fault tolerance

+ Complexity

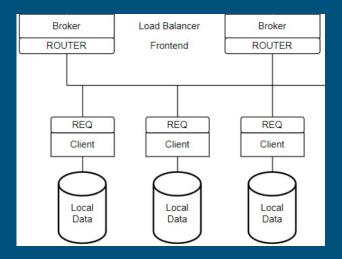
+ Hashing complexity

- Used resources:
 - 5 Physical Nodes
 - 20 Virtual Nodes



Client Logic

- Read database store in CRDT
- Write action CRDT/database locally updated
- **REQ** socket:
 - unique identity
 - connected to random broker
 - registered for polling
- Possible actions/requests:
 - access a shopping list ID
 - synchronize with cloud:
 - CRDT sent to be processed
 - CRDT merged with response CRDT
 - database update
 - GUI refresh



Communication failure:

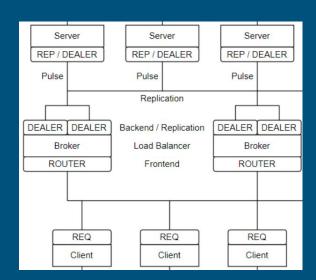
- timeout:
 - issue in broker or server communication
- connect to next broker, repeat
- all brokers dead:
 - end current request
 - maintain responsiveness

Broker Logic

- 3 sockets (polling):
 - ROUTER (frontend)
 - DEALER (backend)
 - DEALER (replication)
- Periodic "pulse check" servers online/offline status

Frontend polling

- Access to shopping list:
 - ask servers, return to client
- Sync with cloud:
 - CRDT with several lists:
 - split/distribute to servers
 - append each server response as CRDTs are sent/resolved/received
 - send it all back to client



Backend polling

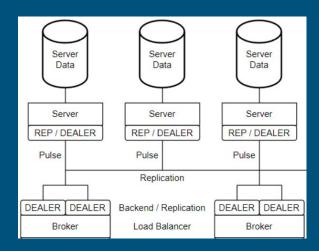
- Received CRDT from server (replicate):
 - split/distribute to servers
 - replication socket used
 - message updated as each CRDT is sent/processed/received across servers
- Reply success to original server

Server Logic

Database reading at start - data stored in CRDT

Socket in REP mode

- Response to pulse check
- Client access to list database query, list data sent
- Client sync:
 - CRDT merged with message CRDT
 - database recognize lists to be replicated
 - resolved CRDT sent
- Data received to replicate:
 - same as client sync
 - o only difference:
 - database update with lists from replication



Socket in DEALER mode

- Replication (at regular intervals) if:
 - o database has lists with updates → replicate
- CRDT to broker, get feedback:
 - if replication worked:
 - database clear info of lists to be replicated
- Communication failure:
 - same logic used by clients for brokers

System Trade-Offs

- Client-cloud sync intervals higher VS smaller:
 - bandwidth more VS less
 - overhead reduced VS increased
- Waiting timeout for broker higher VS smaller:
 - error rate decreased VS increased
 - error detection delayed VS premature
- Replication intervals higher VS smaller:
 - o data staleness increased VS decreased
 - o network resource demand less VS more
- Brokers many VS few:
 - coordination complex VS simple
 - fault tolerance high VS low

- Storing removed shopping lists:
 - **X** increased storage
 - ✓ faster/better access and recovery
- Periodic ARE YOU THERE messages:
 - increased network overhead
 - better re-routing (less downtime)
- Data replication across servers using brokers:
 - **X** potential increased latency
 - enhanced monitoring, load balancing

Conclusions

- CAP theorem AP system:
 - Availability read/write requests get responses, regardless of success/failure of other nodes
 - Partition Tolerance disruption of node segments recognized/bypassed to maintain communications
 - Consistency not all nodes have the same data at the same time, replication works towards it (eventual)
- System scalability considerations:
 - **V** collaborative editing of lists well managed data synchronization
 - ✓ clever server monitoring strategy and disruption recovery fault tolerance
 - ✓ fluctuating user traffic not very problematic load balancing
 - x network issues, bandwidth limitations detrimental speed/efficiency
 - more dependency towards brokers point of failure concerns
- Future work:
 - maybe compare data redundancy strategies
 - implement server-to-server direct communication
 - see how it affects network performance and system overall data