

Horizontal scaling is chaotic

- Probability of any failure in given period = $1-(1-p)^n$
 - -p = probability a machine fails in given period
 - -n = number of machines
- For 50K machines, each with 99.99966% available
 16% of the time, data center experiences failures
- For 100K machines, failures 30% of the time!

Today

- 1. Techniques for partitioning data
- Metrics for success
- 2. Case study: Amazon Dynamo key-value store

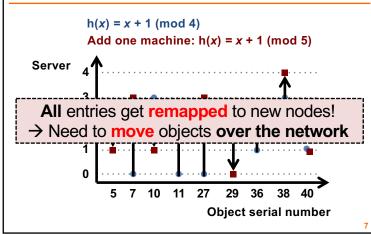
Scaling out: Partition and place

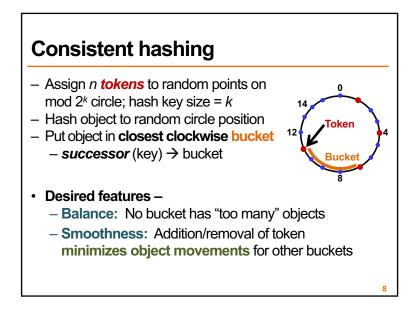
- Partition management
 - Including how to recover from node failure
 - e.g., bringing another node into partition group
 - Changes in system size, *i.e.* nodes joining/leaving
- Data placement
 - On which node(s) to place a partition?
 - Maintain mapping from data object to responsible node(s)
- Centralized: Cluster manager
- Decentralized: Deterministic hashing and algorithms

Modulo hashing

- Consider problem of data partition:
 Given object id X, choose one of k servers to use
- Suppose instead we use modulo hashing:
 Place X on server i = hash(X) mod k
- What happens if a server fails or joins (k ← k±1)?
 or different clients have different estimate of k?

Problem for modulo hashing: Changing number of servers







Each node owns 1/nth of the ID space in expectation

 Says nothing of request load per bucket

• If a node fails, its successor takes over bucket

- Smoothness goal ✓: Only localized shift, not O(n)
- But now successor owns two buckets: 2/nth of key space
 The failure has upset the load balance

Virtual nodes

- Idea: Each physical node now maintains v > 1 tokens
 Each token corresponds to a virtual node
- Each virtual node owns an expected 1/(vn)th of ID space
- Upon a physical node's failure, v successors take over, each now stores (v+1)/v × 1/nth of ID space
- · Result: Better load balance with larger v

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Dynamo: The P2P context

- Chord and DHash intended for wide-area P2P systems – Individual nodes at Internet's edge, file sharing
- Central challenges: low-latency key lookup with small forwarding state per node
- Techniques:

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- Consistent hashing to map keys to nodes
- Replication at successors for availability under failure

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Amazon's workload (in 2007)

- Tens of thousands of servers in globally-distributed data centers
- Peak load: Tens of millions of customers
- Tiered service-oriented architecture
- Stateless web page rendering servers, atop
- Stateless aggregator servers, atop
- Stateful data stores (e.g. Dynamo)
 - put(), get(): values "usually less than 1 MB"

How does Amazon use Dynamo?

Session info

Shopping cart

- Maybe "recently visited products" et c.?

Product list

- Mostly read-only, replication for high read throughput

Dynamo requirements

- Highly available writes despite failures
- Despite disks failing, network routes flapping, "data centers destroyed by tornadoes"
 - **Non-requirement:** Security, *viz.* authentication, authorization (used in a non-hostile environment)
- Low request-response latency: focus on 99.9% SLA
- Incrementally scalable as servers grow to workload
 Adding "nodes" should be seamless
- Comprehensible conflict resolution
- High availability in above sense implies conflicts

Design questions

- How is data placed and replicated?
- How are requests routed and handled in a replicated system?
- How to cope with temporary and permanent node failures?

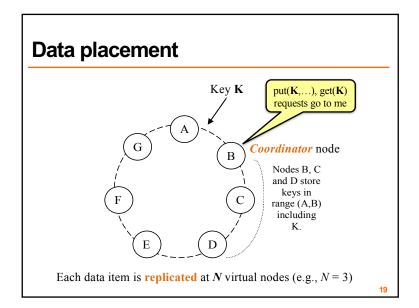
Dynamo's system interface

- Basic interface is a key-value store
 - get(k) and put(k, v)
 - Keys and values opaque to Dynamo
- get(key) → value, context
 - Returns one value or multiple conflicting values
 - Context describes version(s) of value(s)
- put(key, context, value) → "OK"
 - Context indicates which versions this version supersedes or merges

Dynamo's techniques

- Place replicated data on nodes with consistent hashing
- Maintain consistency of replicated data with vector clocks
 - Eventual consistency for replicated data: prioritize success and low latency of writes over reads
 And availability over consistency (unlike DBs)
- Efficiently synchronize replicas using Merkle trees

Key trade-offs: Response time vs. consistency vs. durability



Data replication

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- Much like in Chord: a key-value pair → key's N successors (preference list)
 - Coordinator receives a put for some key
 - Coordinator then replicates data onto nodes in the key's preference list
- Preference list size > N to account for node failures
- For robustness, the preference list skips tokens to ensure distinct physical nodes

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Gossip and "lookup"

- Gossip: Once per second, each node contacts a randomly chosen other node
 - They exchange their lists of known nodes (including virtual node IDs)
- Each node learns which others handle all key ranges
 - Result: All nodes can send directly to any key's coordinator ("zero-hop DHT")

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Reduces variability in response times

Partitions force a choice between availability and consistency

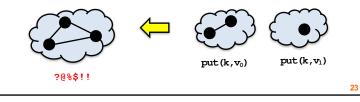
Suppose three replicas are partitioned into two and one



- · If one replica fixed as master, no client in other partition can write
- In Paxos-based primary-backup, no client in the partition of one can write
- Traditional distributed databases emphasize consistency over availability when there are partitions

Alternative: Eventual consistency

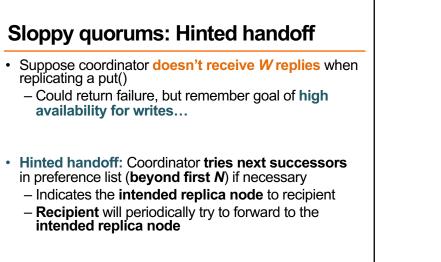
- Dynamo emphasizes availability over consistency when there are partitions
- Tell client write complete when only some replicas have stored it
- · Propagate to other replicas in background
- · Allows writes in both partitions...but risks:
 - Returning stale data
 - Write conflicts when partition heals:



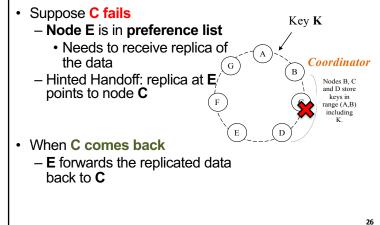
Mechanism: Sloppy quorums

- If no failure, reap consistency benefits of single master
 Else sacrifice consistency to allow progress
- Dynamo tries to store all values put() under a key on first *N* live nodes of coordinator's preference list
- BUT to speed up get() and put():
 - Coordinator returns "success" for put when W < N replicas have completed write
 - Coordinator returns "success" for get when R < N replicas have completed read

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Hinted handoff: Example



Wide-area replication

- Last ¶, § 4.6: Preference lists always contain nodes from more than one data center
 - Consequence: Data likely to survive failure of entire data center
- Blocking on writes to a remote data center would incur unacceptably high latency
 - **Compromise: W < N**, eventual consistency

Sloppy quorums and get()s

- Suppose coordinator doesn't receive R replies when processing a get()
 - Penultimate ¶, § 4.5: "*R* is the min. number of nodes that must participate in a successful read operation."
 Sounds like these get()s fail
- Why not return whatever data was found, though?
 - As we will see, consistency not guaranteed anyway...

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- Common case given in paper: N = 3; R = W = 2
 - With these values, do sloppy quorums guarantee a get() sees all prior put()s?
- If no failures, yes:
 - Two writers saw each put()
 - Two readers responded to each get()
 - Write and read quorums must overlap!

Sloppy quorums and freshness

- Common case given in paper: N = 3, R = W = 2
 With these values, do sloppy quorums guarantee
 - a get() sees all prior put()s?
- With node failures, no:
 - Two nodes in preference list go down
 put() replicated outside preference list
 - Two nodes in preference list come back up
 get() occurs before they receive prior put()

Conflicts

- Suppose N = 3, W = R = 2, nodes are named A, B, C
 - -1^{st} put(k, ...) completes on **A** and **B**
 - -2^{nd} put(k, ...) completes on **B** and **C**
 - Now get(k) arrives, completes first at \boldsymbol{A} and \boldsymbol{C}
- Conflicting results from A and C
 - Each has seen a different put(k, ...)
- · Dynamo returns both results; what does client do now?

Conflicts vs. applications

- Shopping cart:
 - Could take union of two shopping carts
 - What if second put() was result of user deleting item from cart stored in first put()?
 - Result: "resurrection" of deleted item
- Can we do better? Can Dynamo resolve cases when multiple values are found?
 - **Sometimes.** If it can't, **application** must do so.

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Version vectors (vector clocks)

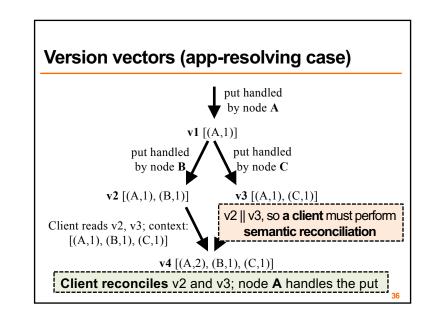
- Version vector: List of (coordinator node, counter) pairs - e.g., [(A, 1), (B, 3), ...]
- Dynamo stores a version vector with **each stored** keyvalue **pair**

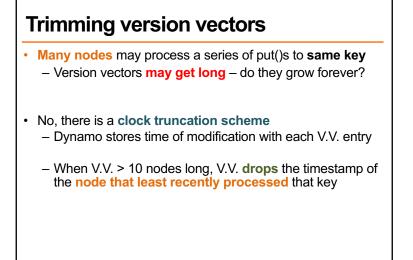
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- Idea: track "ancestor-descendant" relationship between different versions of data stored under the same key ${\bf k}$

Version vectors: Dynamo's mechanism

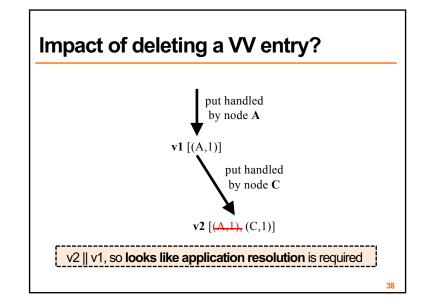
- Rule: If vector clock comparison of v1 < v2, then the first is an ancestor of the second – Dynamo can forget v1
- Each time a put() occurs, Dynamo increments the counter in the V.V. for the coordinator node
- Each time a get() occurs, Dynamo returns the V.V. for the value(s) returned (in the "context")
 - Then users must supply that context to put()s that modify the same key





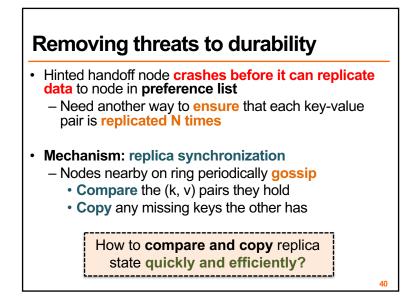
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Concurrent writes

- · What if two clients concurrently write w/o failure?
 - *e.g.* add **different items** to **same cart** at **same time**
 - Each does get-modify-put
 - They both see the same initial version
 - And they both send put() to same coordinator
- · Will coordinator create two versions with conflicting VVs?
 - We want that outcome, otherwise one was thrown away
 - Paper doesn't say, but coordinator could detect problem via put() context



Efficient synchronization with Merkle trees

- Merkle trees hierarchically summarize the key-value pairs a node holds
- One Merkle tree for each virtual node key range
 Leaf node = hash of one key's value
 - Internal node = hash of concatenation of children

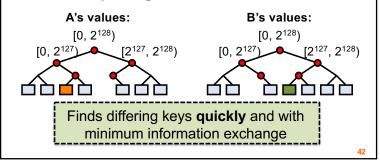
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- Compare roots; if match, values match – If they don't match, compare children
 - Iterate this process down the tree

Merkle tree reconciliation

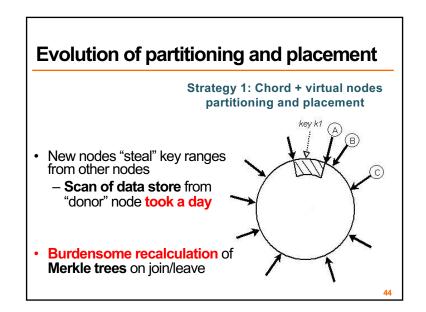
- **B** is missing orange key; **A** is missing green one
- Exchange and compare hash nodes from root downwards, pruning when hashes match

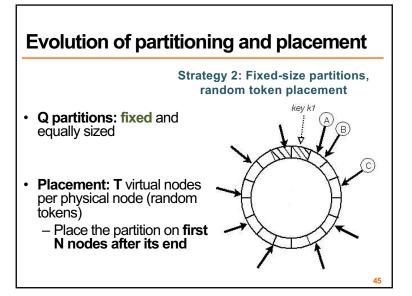


How useful is it to vary N, R, W?

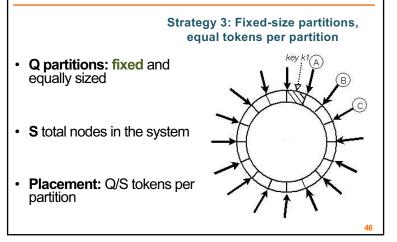
N R W Behavior

- 3 2 2 Parameters from paper: Good durability, good R/W latency
- 3 3 1 Slow reads, weak durability, fast writes
- 3 1 3 **Slow writes,** strong durability, fast reads
- 3 3 3 More likely that reads see all prior writes?
- 3 1 1 Read quorum **doesn't overlap** write quorum





Evolution of partitioning and placement



Dynamo: Take-away ideas

- Consistent hashing broadly useful for replication—not only in P2P systems
- Extreme emphasis on availability and low latency, unusually, at the cost of some inconsistency
- Eventual consistency lets writes and reads return quickly, even when partitions and failures
- Version vectors allow some conflicts to be resolved automatically; others left to application

