

## Scaling Out Key-Value Storage

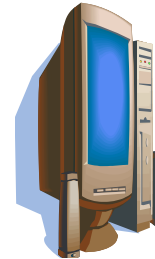


COS 418: *Distributed Systems*  
Lecture 8

Kyle Jamieson

[Selected content adapted from M. Freedman, B. Karp]

## Horizontal or vertical scalability?



Vertical Scaling



Horizontal Scaling

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## Horizontal scaling is challenging

- 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!

**Main challenge:** Coping with **constant failures**

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## Today

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

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## Scaling out: Place and partition

- **Problem 1: Data placement**
  - On which **node(s)** to place a partition?
    - Maintain mapping from data object to responsible node(s)
- **Problem 2: 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**
- **Centralized:** Cluster manager
- **Decentralized:** Deterministic hashing and algorithms

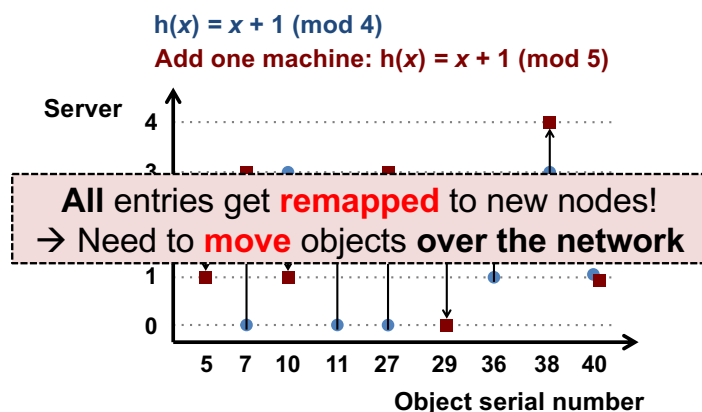
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## Modulo hashing

- 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  $i = \text{hash}(X) \bmod k$
- What happens if a server fails or joins ( $k \leftarrow k \pm 1$ )?
  - or different clients have **different estimate** of  $k$ ?

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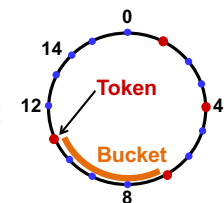
## Problem for modulo hashing: Changing number of servers



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## Consistent hashing

- Assign  $n$  **tokens** to random points on  $\text{mod } 2^k$  circle; hash key size =  $k$
- Hash object to random circle position
- Put object in **closest clockwise bucket**
  - **successor** (key)  $\rightarrow$  bucket
- **Desired features** –
  - **Balance:** No bucket has “too many” objects
  - **Smoothness:** Addition/removal of token **minimizes object movements** for other buckets



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## Consistent hashing's load balancing problem

- Each node owns  $1/n^{\text{th}}$  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:  $2/n^{\text{th}}$  of key space
    - The failure has **upset the load balance**

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## Virtual nodes

- **Idea**: Each physical node implements  $v$  **virtual** nodes
  - Each **physical node** maintains  $v > 1$  **token ids**
    - Each token id corresponds to a virtual node
- Each virtual node owns an expected  $1/(vn)^{\text{th}}$  of ID space
- **Upon a physical node's failure**,  $v$  virtual nodes fail
  - Their successors take over  $1/(vn)^{\text{th}}$  more
- **Result**: **Better load balance** with larger  $v$

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## Today

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

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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**:
  - **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”

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## How does Amazon use Dynamo?

- **Shopping cart**
- **Session info**
  - Maybe “recently visited products” *et c.*?
- **Product list**
  - Mostly read-only, replication for high read throughput

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## 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

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## 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**?

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## 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

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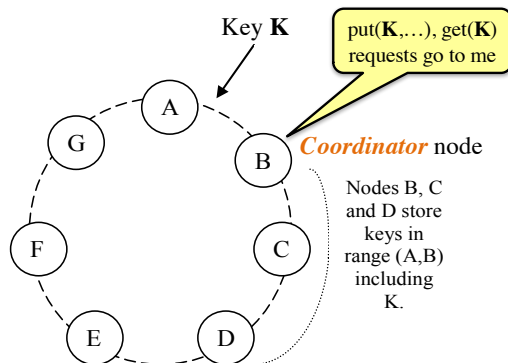
## Dynamo's techniques

- **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

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## Data placement



Each data item is **replicated** at  $N$  virtual nodes (e.g.,  $N = 3$ )

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## Data replication

- 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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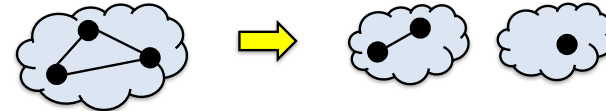
## Gossip and “lookup”

- **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”)**
    - **Reduces variability** in response times

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## Partitions force a choice between availability and consistency

- Suppose **three** replicas are partitioned into **two and one**

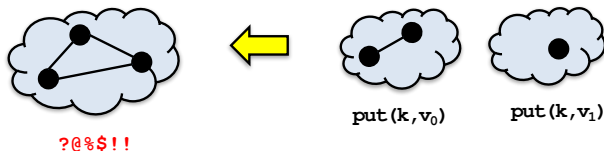


- If one replica fixed as master, no client in other partition can write
- In Paxos-based primary-backup, no client in the partition of one can write
- Traditional distributed databases emphasize consistency over availability when there are partitions

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## 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:



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## 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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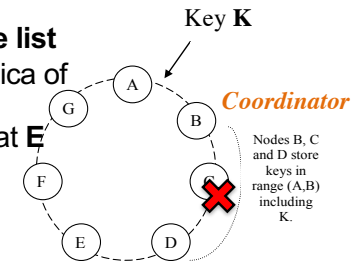
## Sloppy quorums: Hinted handoff

- Suppose coordinator **doesn't receive  $W$  replies** when replicating a put()
  - Could return failure, but remember goal of **high availability for writes...**
- **Hinted handoff:** Coordinator **tries further nodes** in preference list (**beyond first  $N$** ) if necessary
  - Indicates the **intended replica node** to recipient
  - **Recipient** will periodically try to forward to the **intended replica node**

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## Hinted handoff: Example

- Suppose **C fails**
  - **Node E** is in **preference list**
    - Needs to receive replica of the data
  - Hinted Handoff: replica at **E** points to node **C**
- When **C comes back**
  - **E** forwards the replicated data back to **C**



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## 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

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## 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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## 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?**
- If no failures, **yes:**
  - **Two writers** saw each put()
  - **Two readers** responded to each get()
  - Write and read **quorums must overlap!**

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## 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()

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## Conflicts

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

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## 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** key-value **pair**
- **Idea:** track “ancestor-descendant” relationship between different versions of data stored under the same key **k**

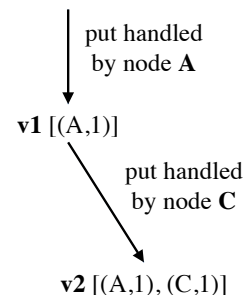
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## Version vectors: Dynamo’s mechanism

- **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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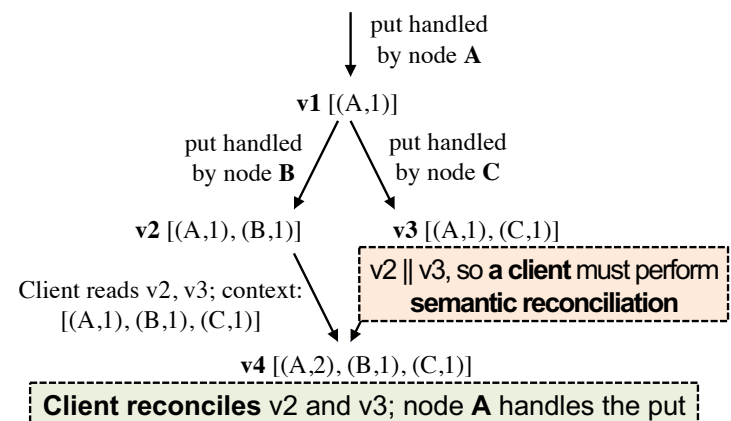
## Version vectors (auto-resolving case)



$v2 > v1$ , so Dynamo nodes **automatically drop v1**, for v2

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## Version vectors (app-resolving case)



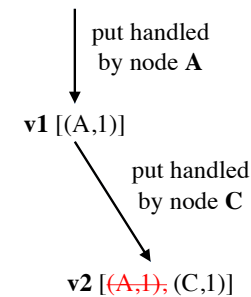
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## Trimming version vectors

- **Many nodes** may process a series of put()s to **same key**
  - Version vectors **may get long** – do they grow forever?
- No, there is a **clock truncation scheme**
  - Dynamo stores time of modification with each V.V. entry
  - When V.V. > 10 nodes long, V.V. **drops** the timestamp of the **node that least recently processed** that key

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## Impact of deleting a VV entry?



v2 || v1, so looks like application resolution is required

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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

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## Removing threats to durability

- Hinted handoff node **crashes before it can replicate data** to node in **preference list**
  - Need another way to **ensure** that each key-value pair is **replicated N times**
- **Mechanism: replica synchronization**
  - Nodes nearby on ring periodically **gossip**
    - **Compare** the (k, v) pairs they hold
    - **Copy** any missing keys the other has

How to **compare and copy** replica state **quickly and efficiently**?

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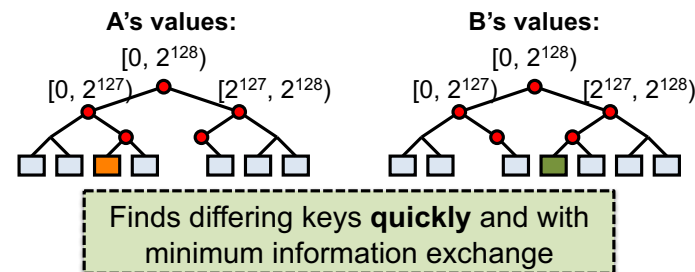
## Efficient synchronization with Merkle trees

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

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## 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**



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## 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, <b>weak durability</b> , <b>fast writes</b>
3	1	3	<b>Slow writes</b> , strong durability, fast reads
3	3	3	More likely that <b>reads see all prior writes?</b>
3	1	1	Read quorum <b>doesn't overlap</b> write quorum

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## 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

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**Wednesday topic:**  
Conflict resolution in  
eventual consistency

**Friday precept:**  
Topic TBA

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