

Distributed Storage Systems part 1

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Distributed Systems and Cloud Computing

This part of the course (5 slots)

Distributed Storage Systems

- CAP theorem and Amazon Dynamo
- Apache Cassandra

Distributed Systems Coordination

- Apache Zookeeper
- Lab on Zookeeper

Cloud Computing summary



General Info

No course notes/book

Slides will be verbose

- List of recommended and optional readings
 - > At the end of the slides
 - On the course webpage
 - ** http://michiard.github.io/DISC-CLOUD-COURSE/
 - See also old course webpage
 - <u>http://www.eurecom.fr/~michiard/teaching/clouds.html</u>



Today

- Distributed Storage systems part 1
 - > CAP theorem
 - Amazon Dynamo



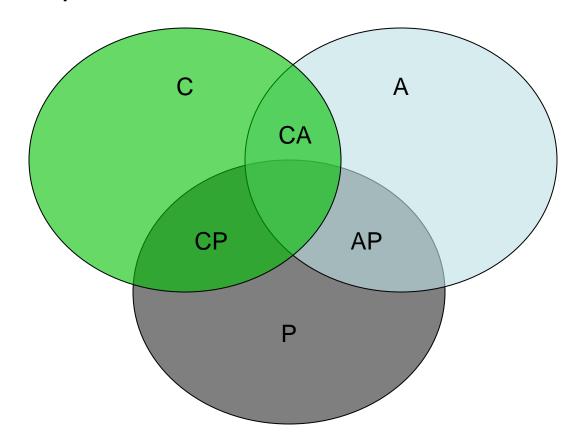
CAP Theorem

- Probably the most cited distributed systems theorem these days
- Relates the following 3 properties
 - C: Consistency
 - One-copy semantics, linearizability, atomicity, total-order
 - Every operation must appear to take effect in a single indivisible point in time between its invocation and response
 - > A: Availability
 - Every client's request is served (receives a response) unless a client fails (despite a strict subset of server nodes failing)
 - P: Partition-tolerance
 - A system functions properly even if the network is allowed to lose arbitrarily many messages sent from one node to another



CAP Theorem

- In the folklore interpretation, the theorem says
 - > C, A, P: pick two!





Be careful with CA

Sacrificing P (partition tolerance)

Negating

➤ A system functions properly even if the network is allowed to lose arbitrarily many messages sent from one node to another

Yields

- ➤ A system does not function properly even if the network is allowed to lose arbitrarily many messages sent from one node to another
 - This boils down to sacrificing C or A (the system does not work)
- ➤ Or... (see next slide)



Be careful with CA

Negating P

A system function properly if the network is not allowed to lose arbitrarily many messages

However, in practice

One cannot choose whether the network will lose messages (this either happens or not)

One can argue that not "arbitrarily" many messages will be lost

- But "a lot" of them might be (before a network repairs)
- In the meantime either C or A is sacrificed



CAP in practice

In practical distributed systems

- Partitions may occur
- This is not under your control (as a system designer)

Designer's choice

- You choose whether you want your system in C or A when/if (temporary) partitions occur
- Note: You may choose neither of C or A, but this is not a very smart option

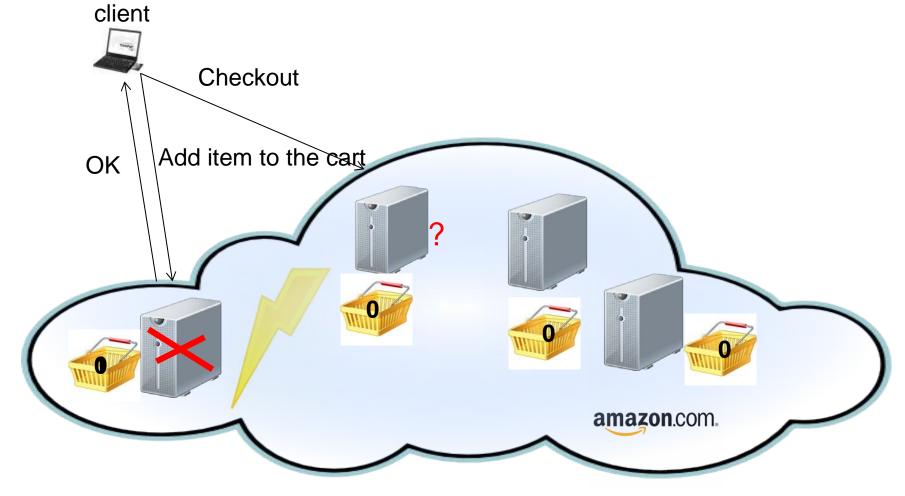
Summary

- > Practical distributed systems are either in CP or AP
- > A given system may shift between CP/AP
 - tunable consistency



CAP proof (illustration)

We cannot have a distributed system in CAP



CAP Theorem

- First stated by Eric Brewer (Berkeley) at the PODC 2000 keynote
- Formally proved by Gilbert and Lynch, 2002
 - ➤ Brewer's conjecture and the feasibility of consistent, available, partition-tolerant web services. <u>SIGACT News</u> 33(2): 51-59 (2002)
- NB: As with all impossibility results mind the assumptions
 - May do nice stuff with different assumptions
- For DistAlgo students
 - Yes, CAP is a "younger sibling" of the FLP impossibility

Gilbert/Lynch theorems

Theorem 1

It is impossible in the **asynchronous** network model to implement a read/write data object that guarantees

- Availability
- Atomic consistency

in all <u>fair</u> executions (including those in which messages are lost)

asynchronous networks: no clocks, message delays unbounded



Gilbert/Lynch theorems

Theorem 2

It is impossible in the **partially synchronous** network model to implement a read/write data object that guarantees

- Availability
- Atomic consistency

in all executions (including those in which messages are lost)

partially synchronous networks: bounds on:

- a) time it takes to deliver messages that are not lost and
- b) message processing time,

exist and are known, but process clocks are not synchronized



Gilbert/Lynch tCA

- t-connected Consistency, Availability and Partition tolerance can be combined
- t-connected Consistency (roughly)
 - > w/o partitions the system is consistent
 - In the presence of partitions stale data may be returned (C may be violated)
 - Once a partition heals, there is a time limit on how long it takes for consistency to return

 Could define t-connected Availability in a similar way



CAP: Summary

 The basic distributed systems/cloud computing theorem stating the tradeoffs among different system properties

- In practice, partitions do occur
 - Pick C or A
 - Can have a tunable system one that sometimes prefers C sometimes A
- The choice (C vs. A) heavily depends on what your application/business logic is



CAP: some choices

CP

- BigTable, Hbase, MongoDB, Redis, MemCacheDB, Scalaris, etc.
- (sometimes classified in CA) Paxos, Zookeeper, RDBMSs, etc.

AP

Amazon Dynamo, CouchDB, Cassandra, SimpleDB, Riak, Voldemort, etc.



Amazon Dynamo

Amazon Web Services (AWS)

- [Vogels09] At the foundation of Amazon's cloud computing are infrastructure services such as
 - Amazon's S3 (Simple Storage Service), SimpleDB, and EC2 (Elastic Compute Cloud)
 - These provide the resources for constructing Internetscale computing platforms and a great variety of applications.
- The requirements placed on these infrastructure services are very strict; need to
 - Score high in security, scalability, availability, performance, and cost-effectiveness, and
 - > Serve millions of customers worldwide, continuously.



AWS

Observation

- Vogels does not emphasize consistency
- > AWS is in AP, sacrificing consistency

AWS follows BASE philosophy

BASE (vs ACID)

- Basically Available
- Soft state
- Eventually consistent



Why Amazon favors availability over consistency?

"even the slightest outage has significant financial consequences and impacts customer trust"

- Surely, consistency violations may as well have financial consequences and impact customer trust
 - > But not in (a majority of) Amazon's services
 - ➤ NB: Billing is a separate story



Amazon Dynamo

Not exactly part of the AWS offering

however, Dynamo and similar Amazon technologies are used to power parts of AWS (e.g., S3)

Dynamo powers internal Amazon services

Hundreds of them!

Shopping cart, Customer session management, Product catalog, Recommendations, Order fullfillment, Bestseller lists, Sales rank, Fraud detection, etc.

So what is Amazon Dynamo?

- A highly available key-value storage system
- > Favors high availability over consistency under failures



Key-value store

- put(key, object)
- get(key)
 - > We talk also about writes/reads (the same here as put/get)
- In Dynamo case, the put API is put(key, context, object)
 - where context holds some critical metadata (will discuss this in more details)
- Amazon services (see previous slide)
 - Predominantly do not need transactional capabilities of RDBMs
 - Only need primary-key access to data!
- Dynamo: stores relatively small objects (typically <1MB)



Amazon Dynamo: Features

- High performance (low latency)
- Highly scalable (hundreds of server nodes)
- "Always-on" available (especially for writes)
- Partition/Fault-tolerant
- Eventually consistent
- Dynamo uses several techniques to achieve these features
 - Which also comprise a nice subset of a general distributed system toolbox



Amazon Dynamo: Key Techniques

Consistent hashing [Karger97]

For data partitioning, replication and load balancing

Sloppy Quorums

- Boosts availability in presence of failures
- might result in inconsistent versions of keys (data)

Vector clocks [Fidge88/Mantern88]

For tracking causal dependencies among different versions of the same key (data)

Gossip-based group membership protocol

For maintaining information about alive nodes

Anti-entropy protocol using hash/Merkle trees

Background synchronization of divergent replicas



Amazon SOA platform

Runs on commodity hardware

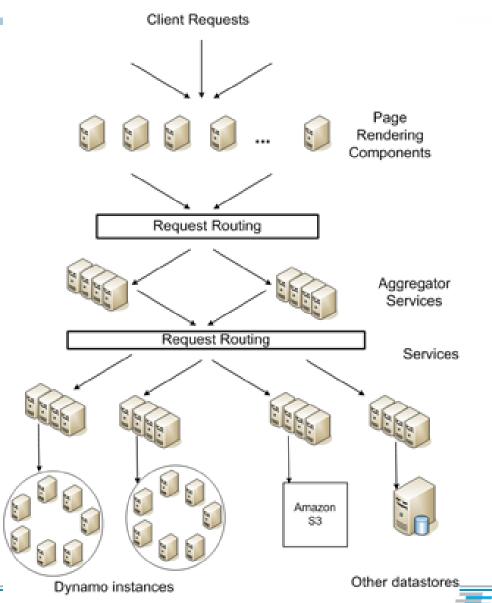
NB: This is low-end server class rather than low-end PC

Stringent Latency requirements

- Measured at 99.9%
- Part of SLAa

Every service runs its own Dynamo instance

- Only internal services use Dynamo
- No Byzantine nodes



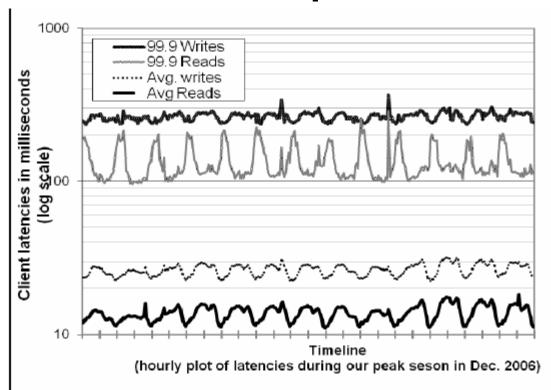
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SLAs and three nines

Sample SLA

➤ A service XYZ guarantees to provide a response within 300 ms for 99.9% of requests for a peak load of 500 req/s

Amazon focuses on 99.9 percentile





Dynamo design decisions

"always-writable" data store

Think shopping cart: must be able to add/remove items

If unable to replicate the changes?

- Replication is needed for fault/disaster tolerance
- > Allow creations multiple versions of data (vector clocks)
- Reconcile and resolve conflicts during reads

How/who should reconcile

- > Application: depending on e.g., business logic
 - Complicates programmer's life, flexible
- > Dynamo: deterministically, e.g., "last-write" wins
 - Simpler, less flexible, might loose some value wrt. Business logic



Dynamo architecture



Dynamo architecture

Scalable and robust components for

➤ Load balancing, membership/fault detection, failure recovery, replica synchronization, overload handling, state transfer, concurrency, job scheduling, request marshalling, request routing, system monitoring and alarming, configuration management

We focus on techniques for

> Partitioning, replication, versioning, membership, failure-handling, scaling



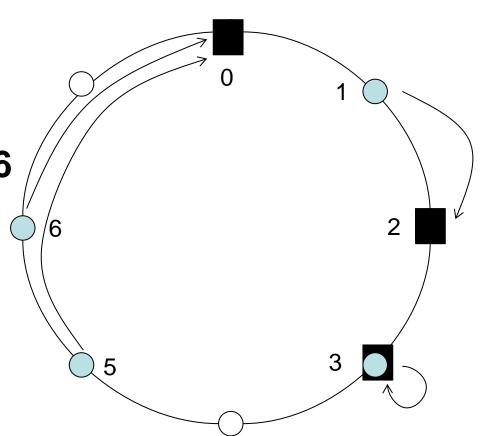
Partitioning using consistent hashing

- Dynamo dynamically partitions a set of keys over a set of storage nodes
 - Used also in many DHTs (e.g., Chord)
- Hashes (MD5, can use SHA-1,...) of keys (resp., node IP) give key (resp., node) m-bit identifiers
- Consistent hashing
 - > Identifiers are ordered in an identifier circle
- Partitioning
 - > A key is assigned to the closest successor node id
 - \triangleright i.e., key k is assigned to the first node with id $\ge k$
 - or if such a node does not exist to the node with smallest id (circle)



Consistent hashing: Example

- m=3: 3-bit namespace
- 3 nodes (0,2,3)
- 4 keys (1,3,5,6)
- Node 0 stores keys 5,6
- Node 2 stores key 1
- Node 3 stores key 3





Consistent hashing

Designed to let nodes enter and leave the network with minimal disruption

Key to incremental scalability

Maintainance

- ➤ When node *n* joins
 - certain keys previously assigned to *n*'s successor now become assigned to *n*.
- When node n leaves
 - all of n's assigned keys are reassigned to n's successor.



Consistent hashing: Properties

Assume N nodes and K keys. Then (with high probability) [Karger97]

- \triangleright Each node is responsible for at most (1+ ε)K/N keys
- When N+1st node joins/leaves, O(K/N) keys change hands (optimal)

• ε=O(logN)

 \triangleright Can have $\epsilon \rightarrow 0$ with "virtual" nodes

"Virtual" nodes

- Each physical node mapped multiple times to the circle
 *Load balancing!
- Dynamo employs virtual nodes also in order to leverage heterogeneity among physical nodes



Replication

To achieve high availability and durability

> Each data item (key) replicated at N nodes

➤ N is configurable per Dynamo instance

Assume N=3

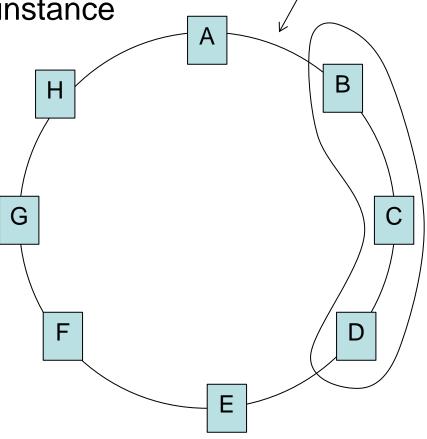
- ➤ For key k, B is the 1st successor node (coordinator)
- ➤ B replicates k to N-1 further successor nodes (C and D)

B, C and D

> are *preference list* for k

Virtual nodes

Same physical nodes skipped in a preference list



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Data versioning

Replication performed after a response is sent to a client

- This is called asynchronous replication (not to be confused with the state machine replication in the asynchronous network model)
- May result in inconsistencies under partitions
 - Read does not return the last value. Eventual consistency!

But operations should not be lost

- "add to cart" should not be rejected but also not forgotten
- ➤ If "add to cart" is performed when latest version is not available it is performed on an older version
- > We may have different versions of a key/value pair



Data versioning

- Once a partition heals versions are merged
 - The goal is not to lose any "add to cart"
- Most of the time there will be no partitions and the system will be consistent
 - New versions subsume all previous ones
- It is vital to understand that the application must know that different versions might exist
 - ➤ This is the Achilles' heel of eventual consistency (more difficult to reason about, program with)
- Key data versioning technique: Vector clocks
 - Capture causality between different versions of an object

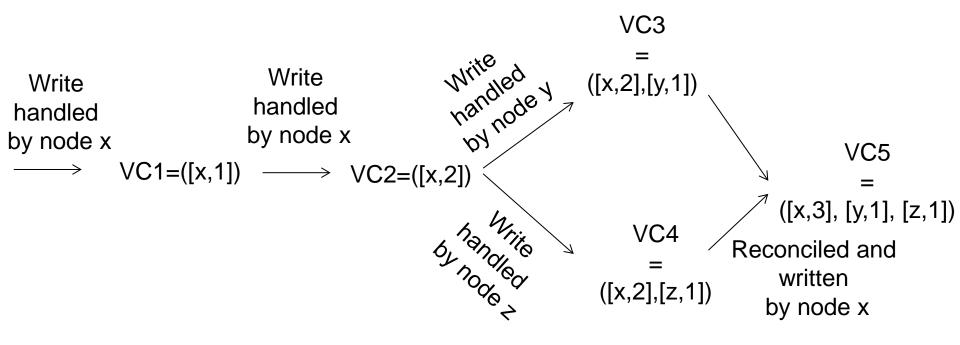


Vector clocks in Dynamo

- Each write to a key k is associated with a vector clock VC(k)
- VC(k) is an array (map) of integers
 - ➤ In theory: one entry VC(k)[i] for each node i
- When node i handles a write of key k it increments VC(k)[i]
 - VCs are included in the context of the put call
- In practice:
 - VC(k) will not have many entries (only nodes from the preference list should normally have entries), and
 - Dynamo truncates entries if more than a threshold (say 10)



Vector clocks in Dynamo



NB: one VC per key



Number of different versions (#DV)

- These are the evidence of consistency violations (#DV>1)
- 24h experiment on the shopping cart
 - > #DV=1: 99.94% of requests (all but 1 in cca 1700 req)
 - > #DV=2: 0.00057% of requests
 - > #DV=3: 0.00047% of requests
 - **>** . . .
- Attributed to busy robots (automated client programs)
 - Rarely visible to humans



Handling puts and gets (failure-free case)

- Any Dynamo storage node can receive get/put request for any key. This node is selected by
 - Generic load balancer
 - By a client library that immediately goes to coordinator nodes in a preference list
- If the request comes from the load balancer
 - > Node serves the request only if in preference list
 - Otherwise, the node routes the request to the first node in preference list
- Each node has routing info to all other nodes
 - ▶ 0-hop DHT
 - Not the most scalable, but latency is critical



Handling puts and gets

Extended preference list

N nodes from preference list + some additional nodes (following the circle) to account for failures

Failure-free case

Nodes from preference list are involved in get/put

Failures

First N alive nodes from extended preference list are involved



Dynamo's quorums

Two configurable parameters

- > R number of nodes that need to participate in a get
- > W number of nodes that need to participate in a write
- R + W > N (a quorum system)
- Handling put (by coordinator) // rough sketch
 Generate new VC, Write new version locally
 Send value, VC to N selected nodes from preference list
 Wait for W-1
- Handling get (by coordinator) // rough sketch
 Send READ to N selected nodes from preference list
 Wait for R
 Select highest versions per VC, return all such versions (causally unrelated)
 Reconcile/merge different versions
 Writeback reconciled version



Of choices of R, W

R, W smaller than N

- To decrease latency
- Slowest replica dictates the latency

- W=1
 - Always-available for writes
 - Yields R=N (reads pay the penalty)

Most often in Dynamo (W,R,N)=(2,2,3)



Handling failures

N selected nodes are the first N healthy nodes

- Might change from request to request
- Hence these quorums are "Sloppy" quorums

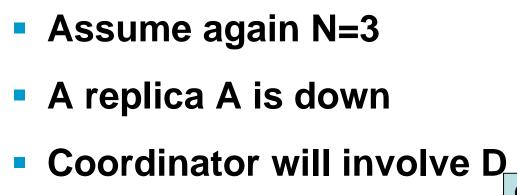
"Sloppy" vs. strict quorums

- "sloppy" allow availability under a much wider range of partitions (failures) but sacrifice consistency
- Also, important to handle failures of an entire data center
 - Power outages, cooling failures, network failures, disasters
 - Preference list accounts for this (nodes spread across data centers)



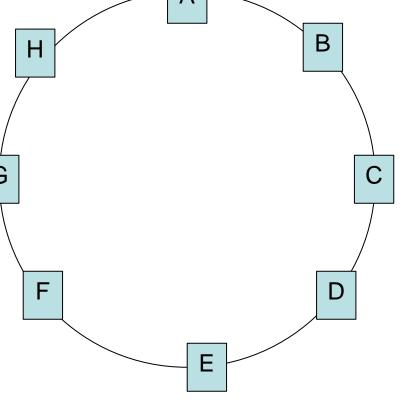
Handling temporary failures: hinted handoff

 If a replica in the preference list is down then another replica is created on a new node



With a hint that this D substitutes A until A comes back again

When D gets info A is back up it hands back the data to A



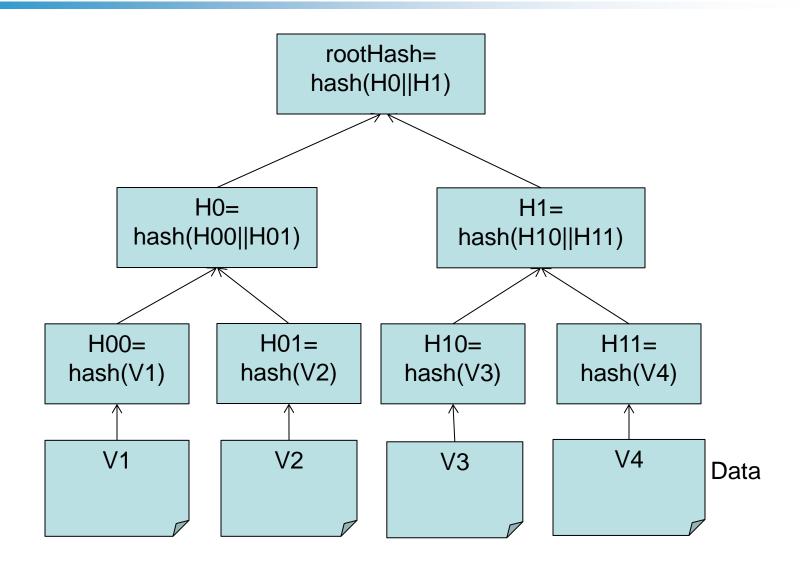


Anti-entropy synchronization using hash/Merkle trees

- Each Dynamo node keeps a Merkle tree for each of its key ranges
 - > Remember, one key range per virtual node
- Compares the root of the tree with replicas
 - ➤ If equal, all keys in a range are equal (replicas in sync)
 - > If not equal
 - Traverse the branches of the tree to pinpoint the children that differ
 - The process continues to all leaves
 - Synchronize on those keys that differ



Merkle trees



Membership

- Node outages temporary
 - Not considered as permanent leaves
- Dynamo relies on administrator explicitly declaring joins/leaves on any Dynamo node
 - This triggers membership changes (with the aid of seeds)
- Membership info are also eventually consistent
 - propagated by background gossip protocol
 - Node contacts a random node every 1s
 - 2 nodes reconcile the membership info
 - This gossip used also for exchanging partitioning/placement metadata



Failure detection

- Unreliable failure detection (FD)
 - Used, e.g., to refresh the healthy node info in the extended preference list
- With steady load node A will find out if node B is unavailable
 - > E.g., if B does not respond to A's messages
 - But this is clearly unreliable, B might be partitioned not faulty
 - Then, A periodically checks on B to see if B recovers
- In the absence of traffic A might not find out B is unavailable
 - But this info anyway does not matter w/o traffic
 - Dynamo has in-band FD, rather than a dedicated component

Dynamo: Summary

- An eventually consistent highly available key value store
 - > AP in the CAP space
- Focuses on low latency, SLAs
 - Very low latency writes, reconciliation in reads
- Key techniques used in many other distributed systems
 - Consistent hashing, (sloppy) quorum-based replication, vector clocks, gossip-based membership, Merkle-tree based synchronization



Further reading (recommended)

Seth Gilbert, Nancy A. Lynch: Brewer's conjecture and the feasibility of consistent, available, partition-tolerant web services. SIGACT News 33(2): 51-59 (2002)

DeCandia et al. Dynamo: Amazon's highly available key-value store.

SOSP 2007: 205-220 (2007)



Further Reading (optional)

- Eric A. Brewer: Pushing the CAP: Strategies for Consistency and Availability. <u>IEEE Computer 45(2)</u>: 23-29 (2012)
- Seth Gilbert, Nancy A. Lynch: Perspectives on the CAP Theorem. IEEE Computer 45(2): 30-36 (2012)

- Marko Vukolić: Quorum Systems with Applications to Storage and Consensus. Morgan&Claypool (2012)
- <u>lon Stoica</u> et al: Chord: a scalable peer-to-peer lookup protocol for internet applications. <u>IEEE/ACM Trans. Netw. 11</u>(1): 17-32 (2003)

