

Object Storage on CRAQ

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So far, we have seen...

Three Replication Techniques!

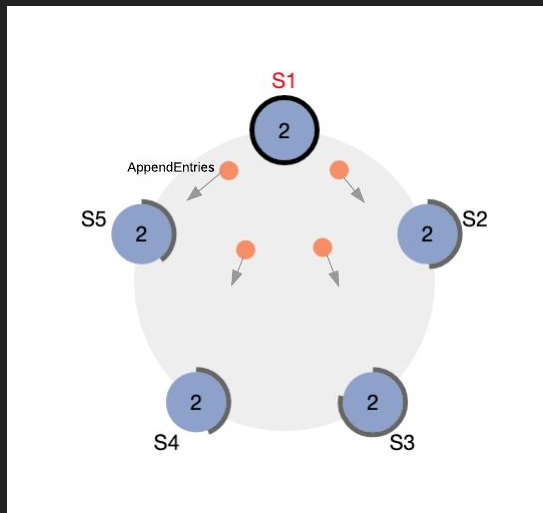
Can you name them?

1. Primary backup replication (**GFS**, **VMware FT**)
2. Consensus Protocol (**Raft**, **Zookeeper**)
3. Chain Replication (**CR**, **CRAQ**)

Motivation



Motivation



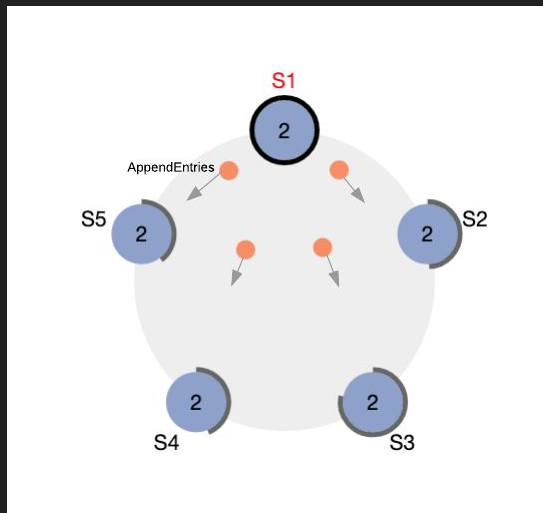
Raft / Paxos

Clients have to send **reads** to the **leader**

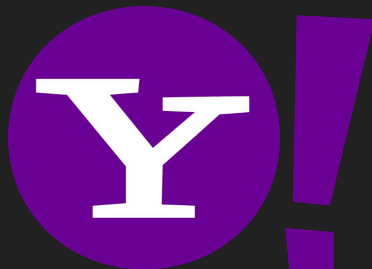
But no opportunity to **divide the read load** over the followers.

How can we improve?

Case Study 1:

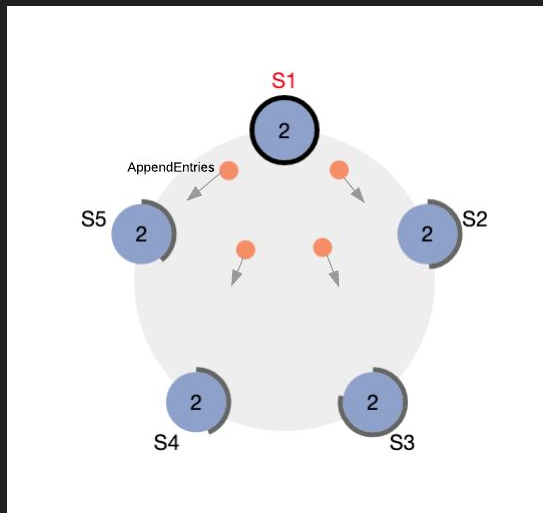


Raft / Paxos



**“Let’s change the definition
of correctness!”**

Case Study 1:



Raft / Paxos



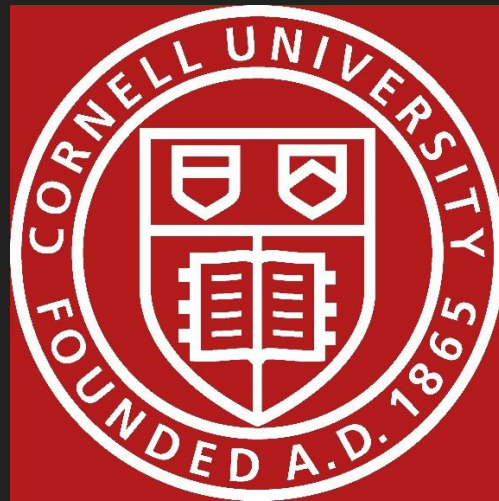
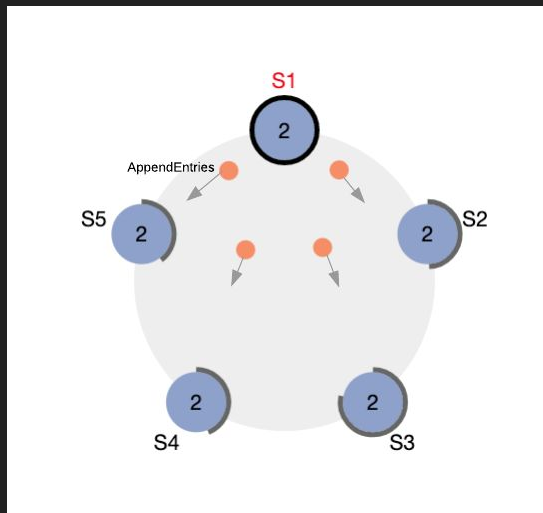
Zookeeper

Linearizable **Writes**

FIFO client order

But leader is still a
bottleneck for write
throughputs.

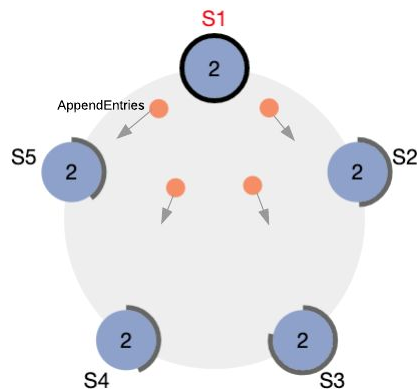
Case Study 2:



We got a new idea!

Raft / Paxos

Case Study 2:



Raft / Paxos



Zookeeper

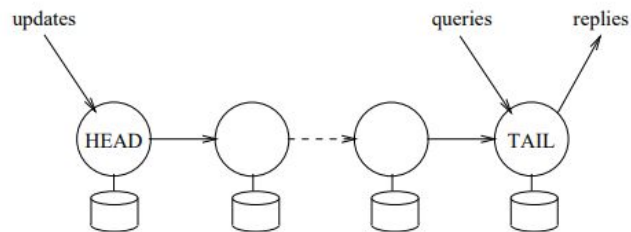


Figure 2: A chain.

**Chain
Replication**

Can we improve read perf to CR as
Zookeeper did to Raft?? 🤔🤔

Case Study 3:

Object Storage on CRAQ

High-throughput chain replication for read-mostly workloads

Jeff Terrace and Michael J. Freedman
Princeton University

Abstract

Massive storage systems typically replicate and partition data over many potentially-faulty components to provide both reliability and scalability. Yet many commercially-deployed systems, especially those designed for interactive use by customers, sacrifice stronger consistency properties in the desire for greater availability and higher throughput.

This paper describes the design, implementation, and evaluation of CRAQ, a distributed object-storage system that challenges this inflexible tradeoff. Our basic approach, an improvement on Chain Replication, maintains strong consistency while greatly improving read throughput. By distributing load across all object replicas, CRAQ scales linearly with chain size without increasing consistency coordination. At the same time, it exposes non-committed operations for weaker consistency guarantees when this suffices for some applications, which is especially useful under periods of high system churn. This paper explores additional design and implementation con-

Object-based systems are more attractive than their file-system counterparts when applications have certain requirements. Object stores are better suited for flat namespaces, such as in key-value databases, as opposed to hierarchical directory structures. Object stores simplify the process of supporting whole-object modifications. And, they typically only need to reason about the ordering of modifications to *a specific object*, as opposed to the entire storage system; it is significantly cheaper to provide consistency guarantees per object instead of across all operations and/or objects.

When building storage systems that underlie their myriad applications, commercial sites place the need for high performance and availability at the forefront. Data is replicated to withstand the failure of individual nodes or even entire datacenters, whether from planned maintenance or unplanned failure. Indeed, the news media is rife with examples of datacenters going offline, taking down entire websites in the process [26]. This strong focus on availability and performance—especially as such properties are being codified in tight SLA requirements [4, 24]—



Google Cloud Platform

What are the claims from CRAQ

1. Strong consistency! 🙌 (unlike Zookeeper)
2. Greatly improving read throughput
3. This means CRAQ is strictly more powerful than CR

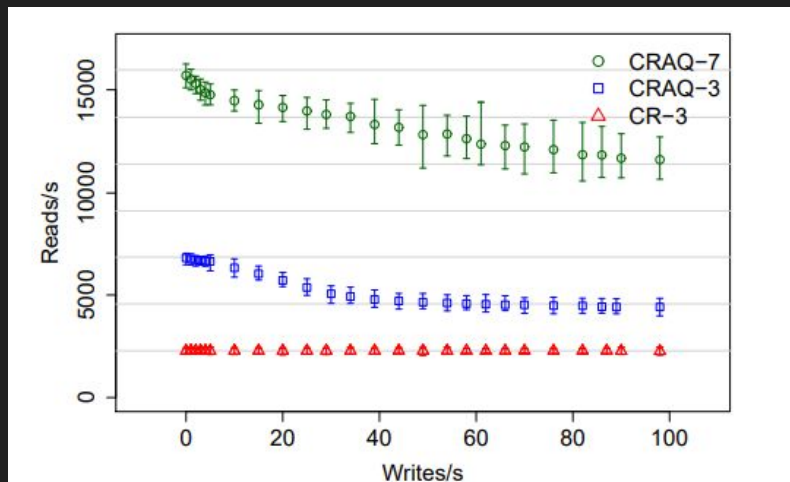


Figure 7: Read throughput as writes increase (5KB object).

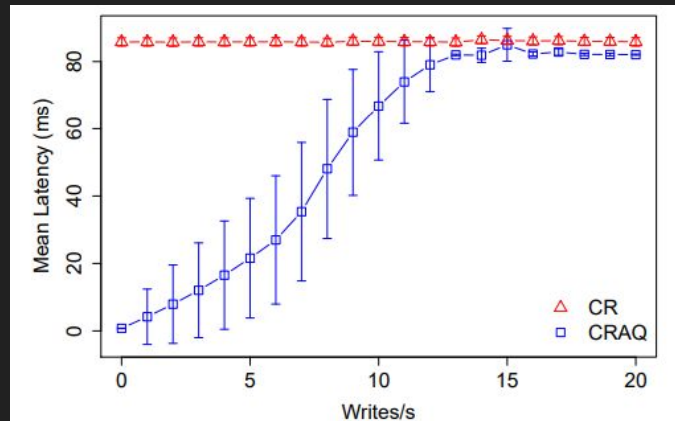






Figure 13: CR and CRAQ's read latency to a local client when the tail is in a distant datacenter separated by an RTT of 80ms and the write rate of a 500-byte object is varied.

What is object storage?

1. Best suited for **unstructured data** (files , music , photos , movies )
2. Support 2 primitive operations
 - a. **Read** (return data block stored under object name)
 - b. **Write** (change the state of a single object)
3. Flat namespaces (key-value)
4. Supporting whole object modifications
5. Significantly cheaper to provide **consistency**

Strong vs Eventual consistency

1. Strong consistency = **linearizability** = a read to an object always sees the latest written value
2. Eventual consistency:
 - a. Writes are **linearizable**
 - b. Reads can return stale data
 - c. When all replicas receive the write, read can never return older version
 - d. Monotonic read consistency for one session.

Some of the most well-known key-value databases

Dynamo: Amazon's Highly Available Key-value Store

Giuseppe DeCandia, Deniz Hastorun, Madan Jampani, Gunavardhan Kakulapati, Avinash Lakshman, Alex Pilchin, Swaminathan Sivasubramanian, Peter Vosshall and Werner Vogels

Amazon.com

ABSTRACT

Reliability at massive scale is one of the biggest challenges we face at Amazon.com, one of the largest e-commerce operations in the world; even the slightest outage has significant financial consequences and impacts customer trust. The Amazon.com platform, which provides services for many web sites worldwide, is implemented on top of an infrastructure of tens of thousands of servers and network components located in many datacenters around the world. At this scale, small and large component failures occur continuously and the way persistent state is managed in the face of these failures drives the reliability of the platform and the software systems.

This paper presents the design of a highly available key-value store that Amazon's core services use to provide a high level of availability and reliability under certain failure scenarios. The design is based on a software

versioning and application-assisted replication strategy that treats failure as a normal event and provides a novel interface for

One of the lessons our organization has learned from Amazon's platform is that the reliability of a distributed system is dependent on how its components are designed. Amazon uses a highly distributed, fault-tolerant, and oriented architecture in its environment. This paper describes the design of a key-value database and shares many design and implementation strategies therewith. Cassandra does not support a full relational data model; instead, it provides clients with a simple data model that supports dynamic control over data layout and format. Cassandra system was designed to run on cheap commodity hardware and handle high write throughput while not sacrificing read efficiency.

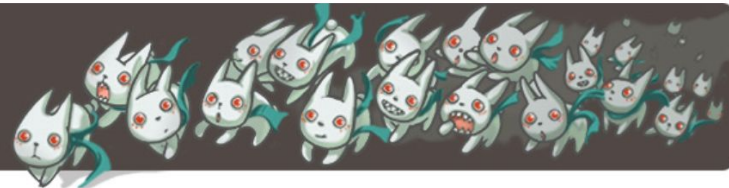
Cassandra - A Decentralized Storage System

that enables users to search through their email inbox. At Facebook this meant the system had to be able to handle a very high write throughput, billions of writes per day, and also a high read throughput. Since users are searching for information in a highly distributed system, the search centers were located in many different datacenters around the world. Facebook's search was launched in 2004 and today it has kept up with the growth of the company's user base within Facebook.

This paper is structured as follows. Section 2 talks about related work, some of which has been very influential on our design. Section 3 presents the data model in more detail. Section 4 presents the overview of the client API. Section 5 presents the details of the implementation. Section 6 discusses the challenges of implementing a distributed system like Cassandra and the lessons learned from the experience.

Cassandra

Sacrifice strong consistency due to tight SLA requirements



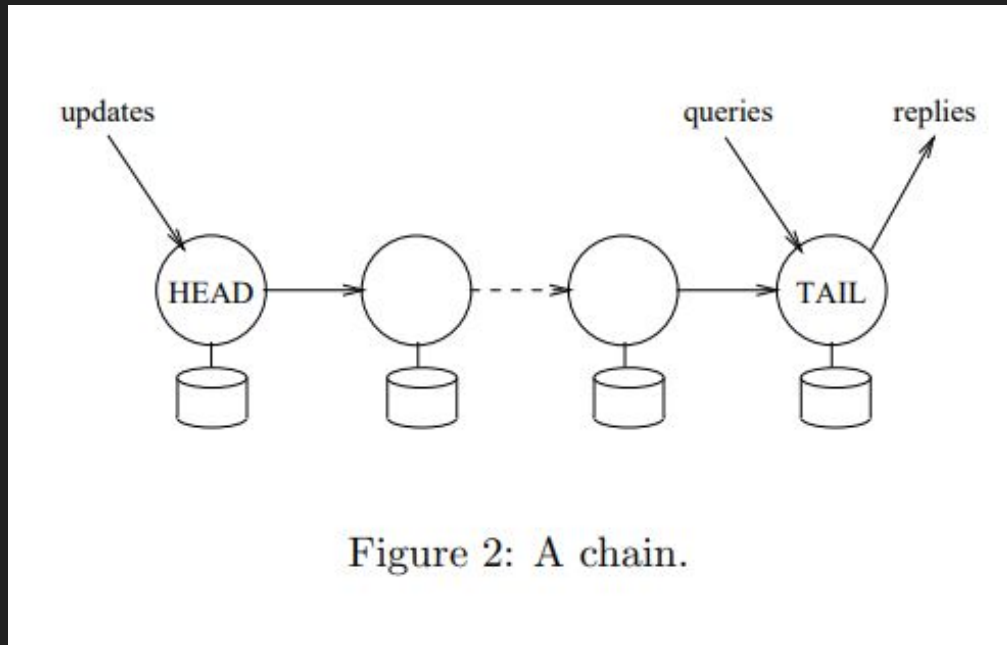
Chain Replication (CR)

Strong consistency ✓

Good throughput ✓

Easy recovery ✓

Tail hotspots ✗



Chain Replication with Apportioned Queries (CRAQ) ✨ ✨

Strong consistency ✓

Good throughput ✓

Easy recovery ✓

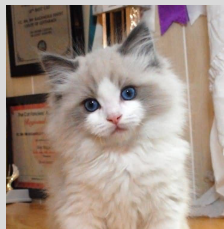
Lower latency ✓

High throughput for read ✓



Basic System Model

Let's start with a
node storing a Cat
pic in a chain.



Version: 1, clean

Basic System Model

Receives a new
version of a cat pic.

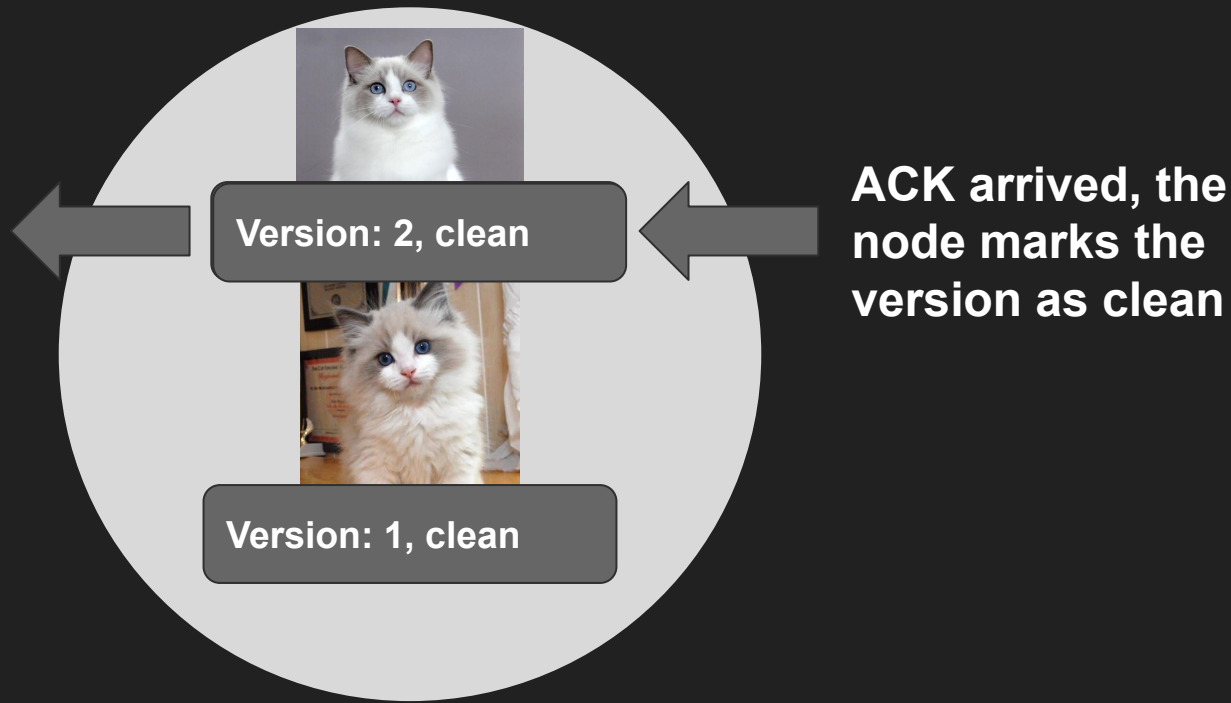
Non-tail: marks dirty

Tail:

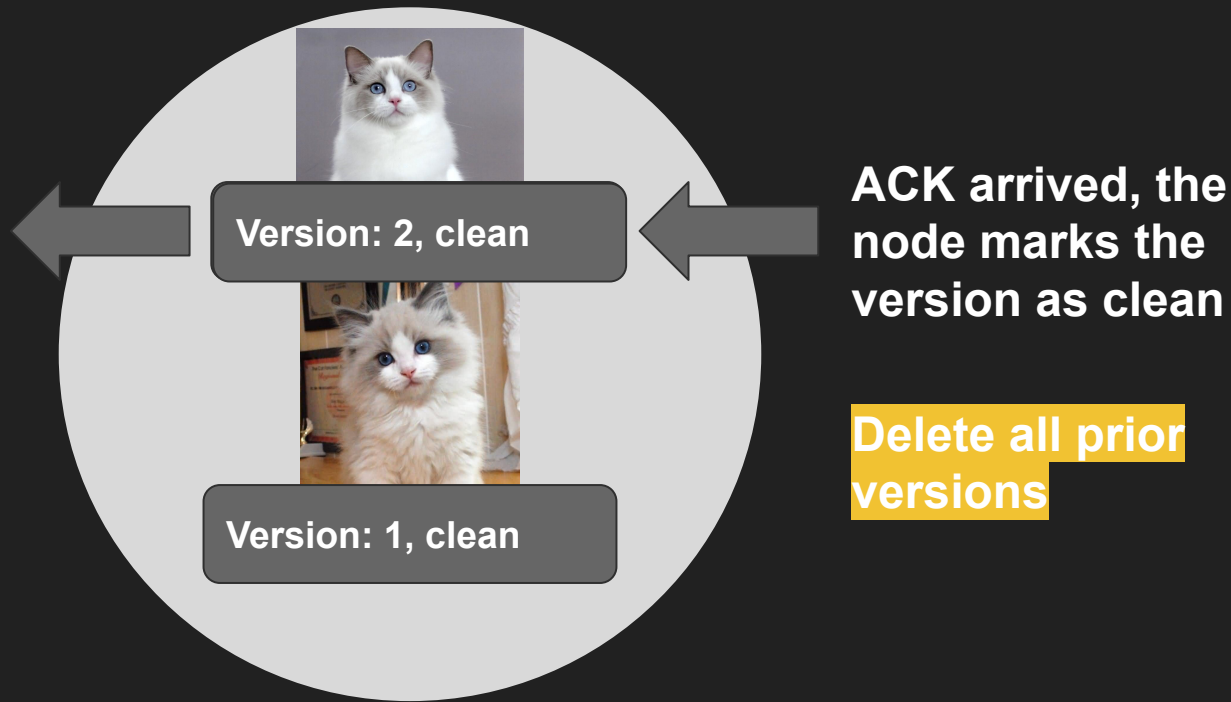
1. marks clean,
2. commit,
3. send **ACK** to all other nodes



Basic System Model



Basic System Model



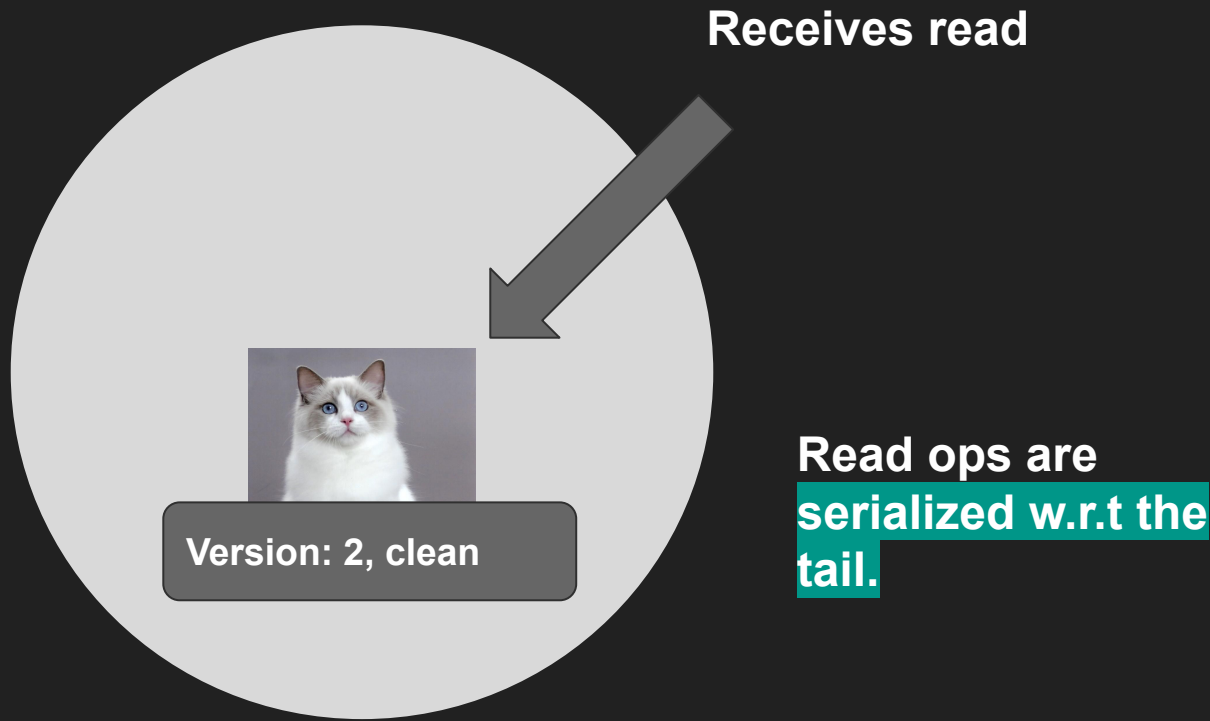
Basic System Model



Basic System Model

If the latest version is clean: return the cat picture

If dirty: The node makes a **version query**, return that version of the cat picture.



Basic System Model (from paper)

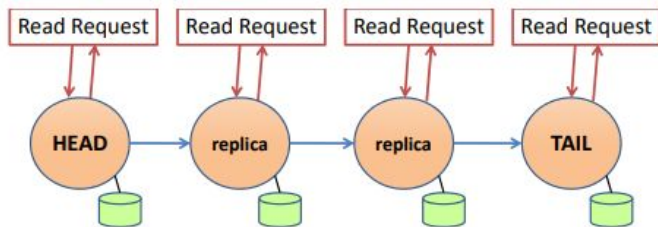


Figure 2: Reads to clean objects in CRAQ can be completely handled by any node in the system.

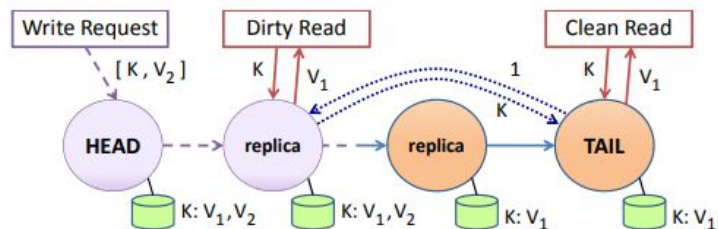


Figure 3: Reads to dirty objects in CRAQ can be received by any node, but require small version requests (dotted blue line) to the chain tail to properly serialize operations.

How does **CRAQ** improve throughput?

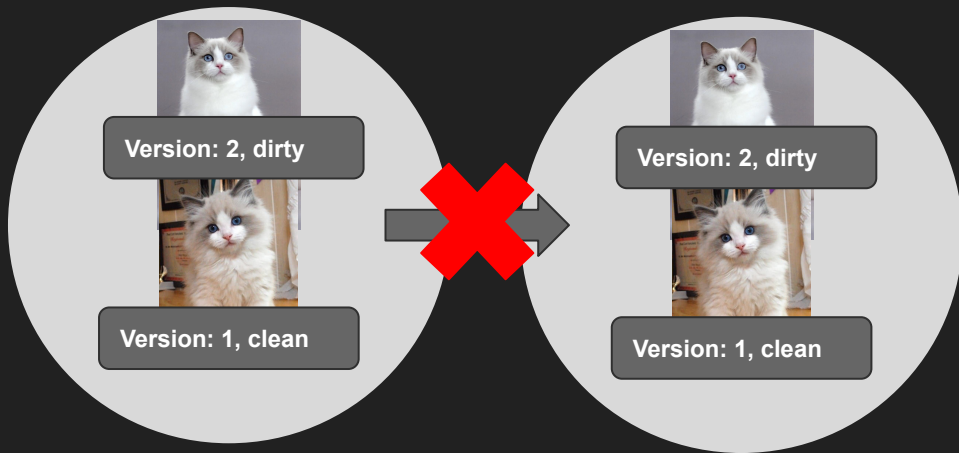
Read-Mostly Workloads: most reads handled by C-1 non-tail nodes. Throughput scales linearly with chain size C

Write-Heavy Workloads: non-tail nodes make lighter-weight version queries for reads. Still a big win over **CR**.

One could optimize read throughput by having the tail node only handle version queries. 🤖🤖

Split-brain problem

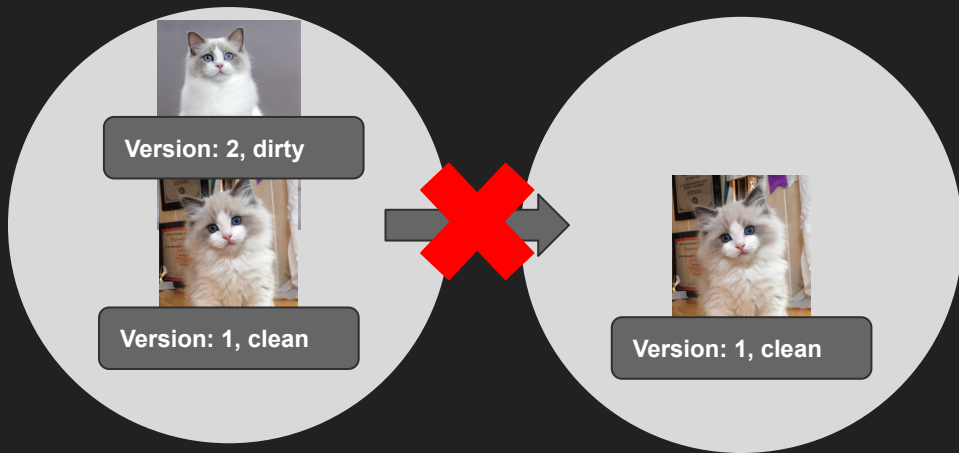
If the head node and the node next to head cannot talk to each other:



Can the node next to head claim itself as the **new head**?

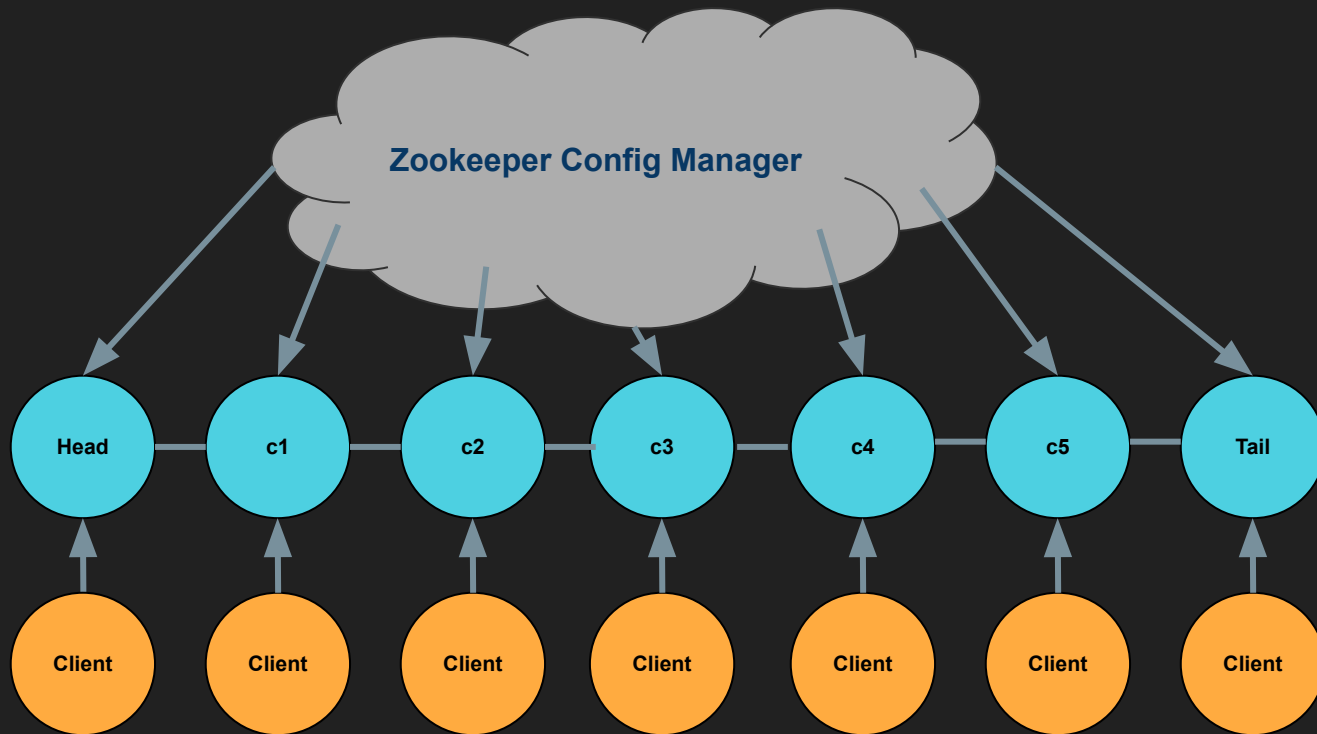
Split-brain problem

If the head node and the node next to head cannot talk to each other:



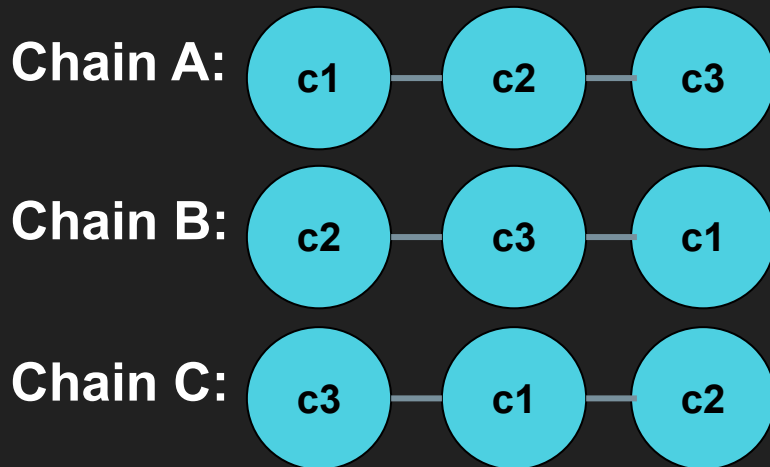
No!

Split-brain problem



Other solution to reduce tail's load

Multi-chain:



Topics not covered

Scaling CRAQ:

1. Implicit data centers & global chain size
2. Explicit data centers & global chain size
3. Explicit data center chain sizes

Extensions:

1. Mini-transactions on CRAQ
2. Handling memberships changes (using Zookeeper)

Questions?