

# Distributed Systems

## CS6421

**Scaling the Web**

Prof. Tim Wood

# Practice / Projects

You will learn more by trying to build something real!

If you want to get involved in research, this is your chance!

- I will be accepting students into a 3 credit Research course for the spring... but you need to do a cloud/NFV project and it needs to be done well! Impress me!

If you don't do a project, you need to write a technical “blog” post explaining a cloud technology

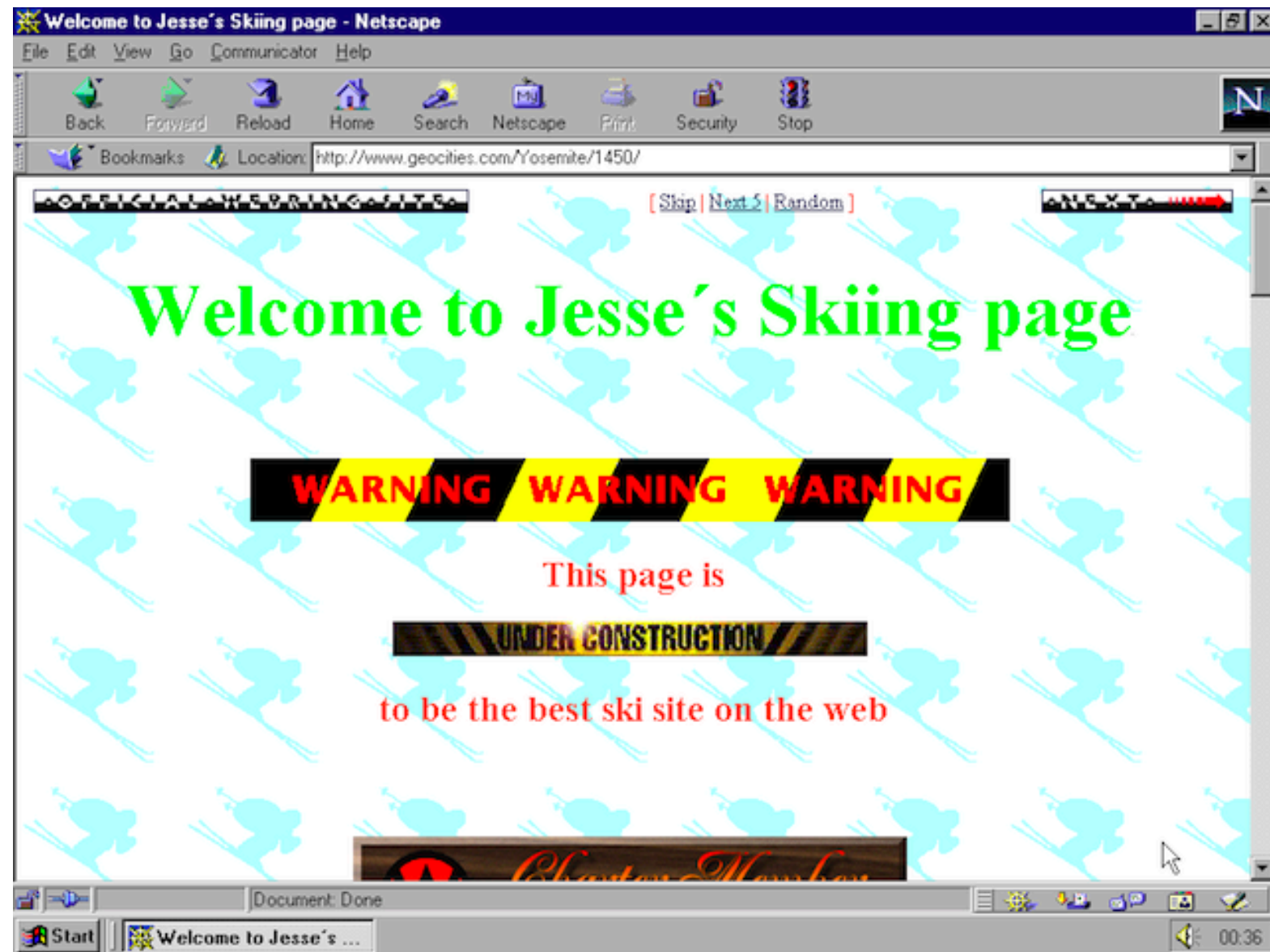
# Antique Web Servers

## Serve static content

- Read a file from disk and send it back to the client
- images, HTML

## Dynamic Content

- CGI Bin
- executes a program
- Not very safe or convenient for development...



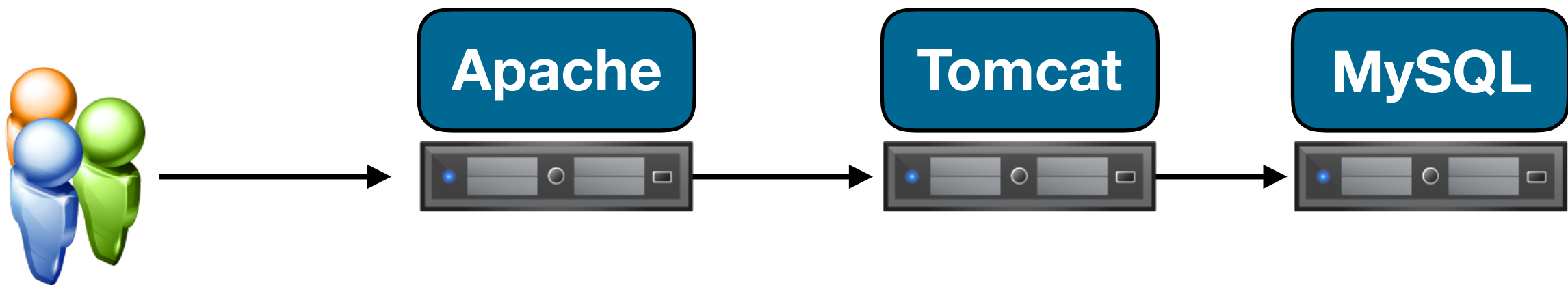
# 3-tier Web Applications

LAMP = Linux, Apache, MySQL, PHP

Separation of duties:

- Front-end web server for static content (Apache, lighttpd, nginx)
- Application tier for dynamic logic (PHP, Tomcat, node.js)
- Database back-end holds state (MySQL, MongoDB, Postgres)

Why divide up in this way?



# Stateful vs Stateless

The multi-tier architecture is based largely around whether a tier needs to worry about state

Front-end - totally **stateless**

- There is no data that must be maintained by the server to handle subsequent requests

Application tier - maintains **per-connection state**

- There is some temporary data related to each user, e.g., my shopping cart
- May not be critical for reliability - might just store in memory

Database tier - global state

- Maintains the global data that application tier might need
- Persists state and ensures it is consistent



# N-Tier Web Applications

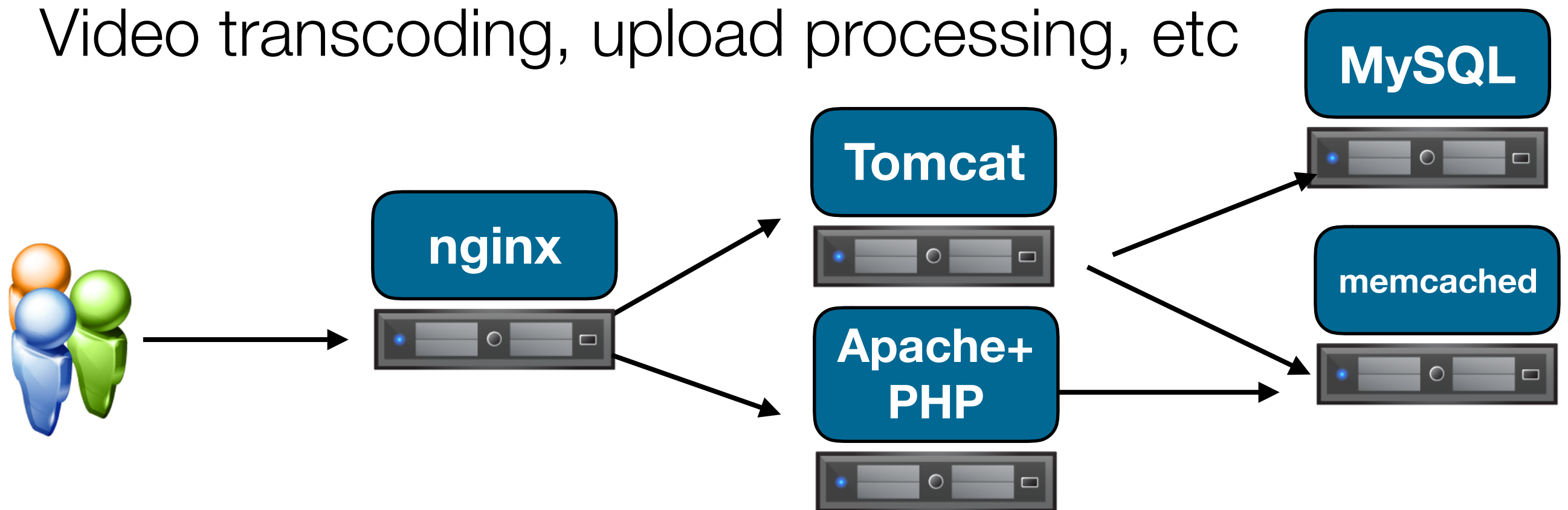
Sometimes 3 tiers isn't quite right

Database is often a bottleneck

- Add a cache! (stateful, but not persistent)

Authentication or other security services could be another tier

Video transcoding, upload processing, etc

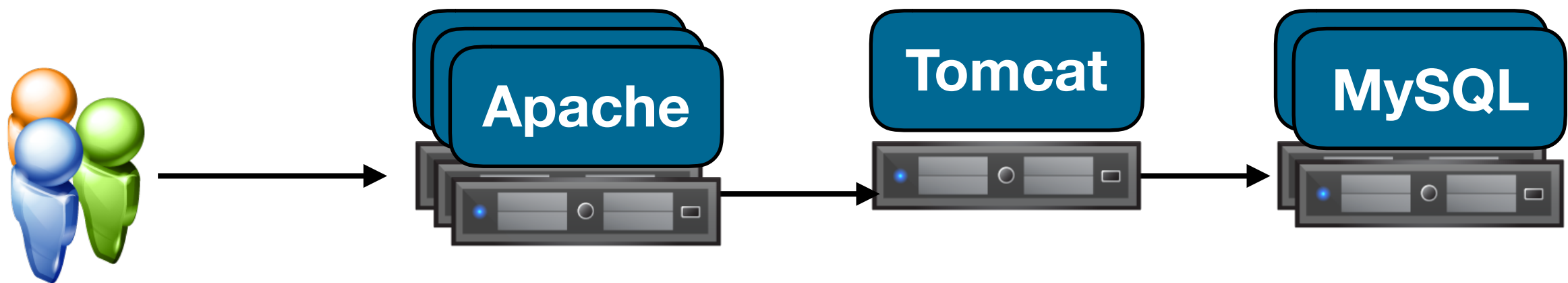


# Replicated N-Tier

Replicate the portions of the system that are likely to become overloaded

How easy to scale...?

- Apache serving static content
- Tomcat Java application managing user shopping carts
- MySQL cluster storing products and completed orders



Tune number of replicas based on demand at each tier

# Wikipedia: Big scale, cheap

5th busiest site in the world (according to alexa.com)

Runs on about ~ **1000** servers? (700 in 2012)

All open source software:

- PHP, MariaDB, Squid proxy, memcached, Ubuntu

Goals:

- Store lots of content (6TB of text data as of 2018)
- Make available worldwide
- Do this as cheaply as possible
- Relatively weak consistency guarantees

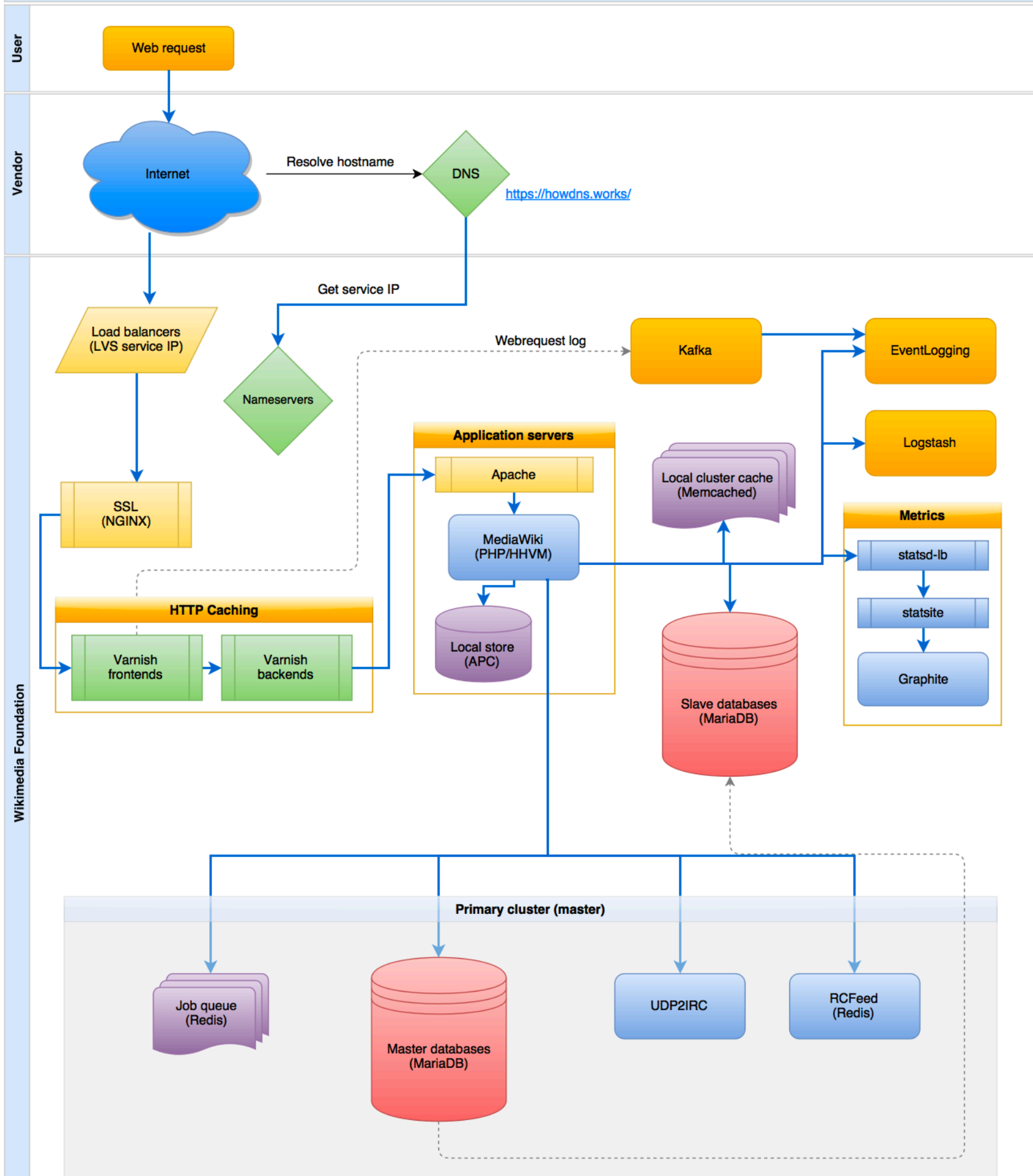
Stats: <https://grafana.wikimedia.org>



# Wikipedia

## Networking and application infrastructure

Revision: October 2015

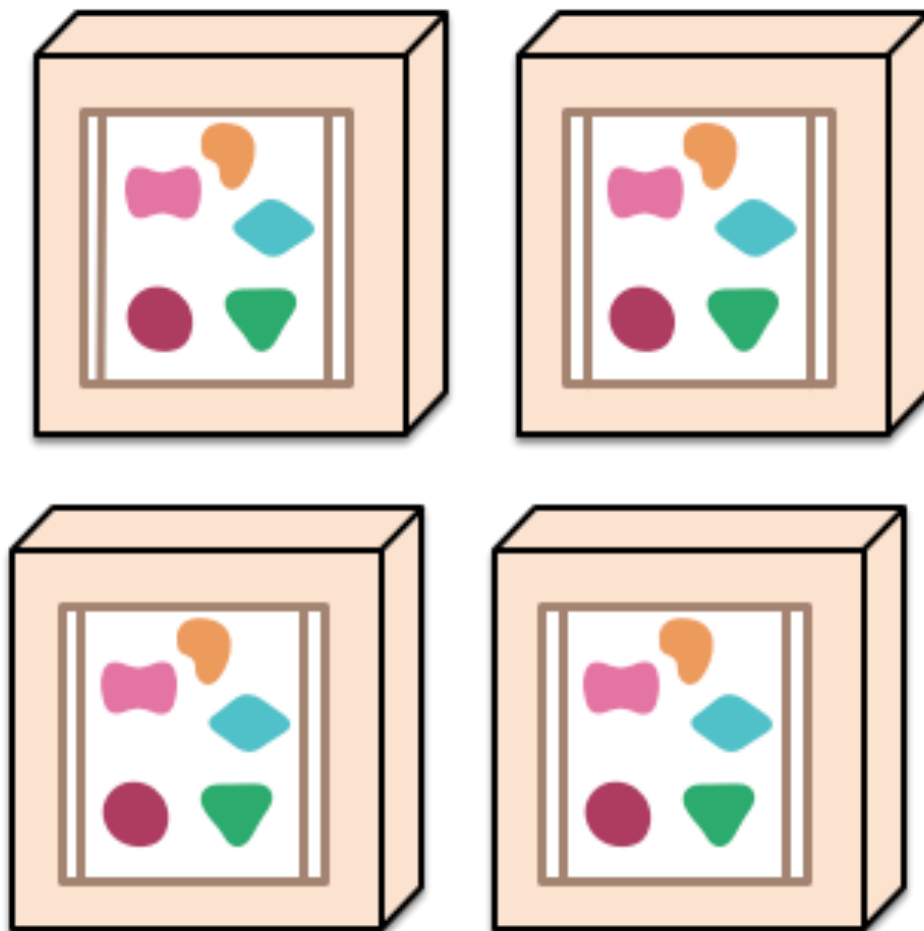


# Application Tier

*A monolithic application puts all its functionality into a single process...*



*... and scales by replicating the monolith on multiple servers*



<http://martinfowler.com/articles/microservices.html>

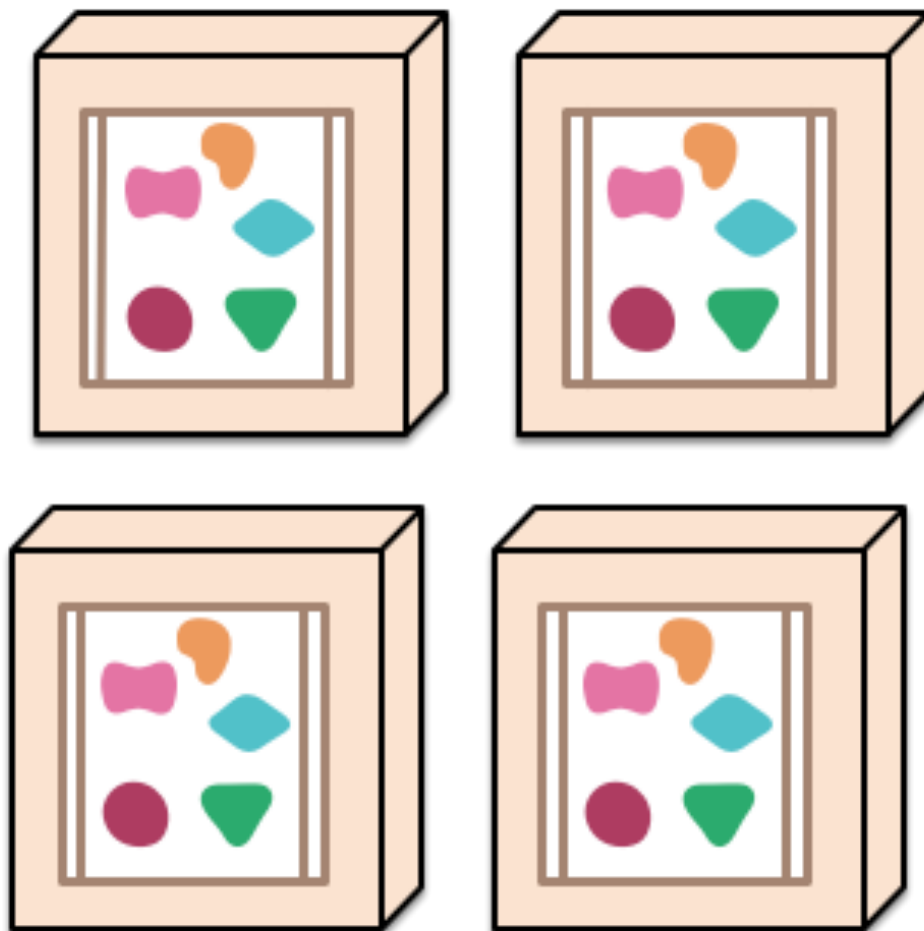
**Problems with  
Monolithic  
approach?**

# Microservices

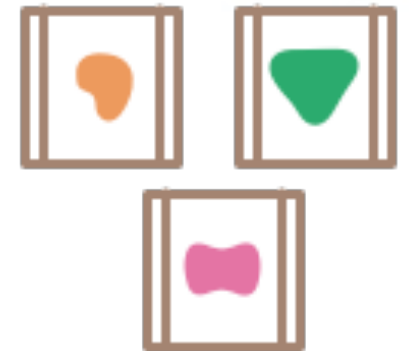
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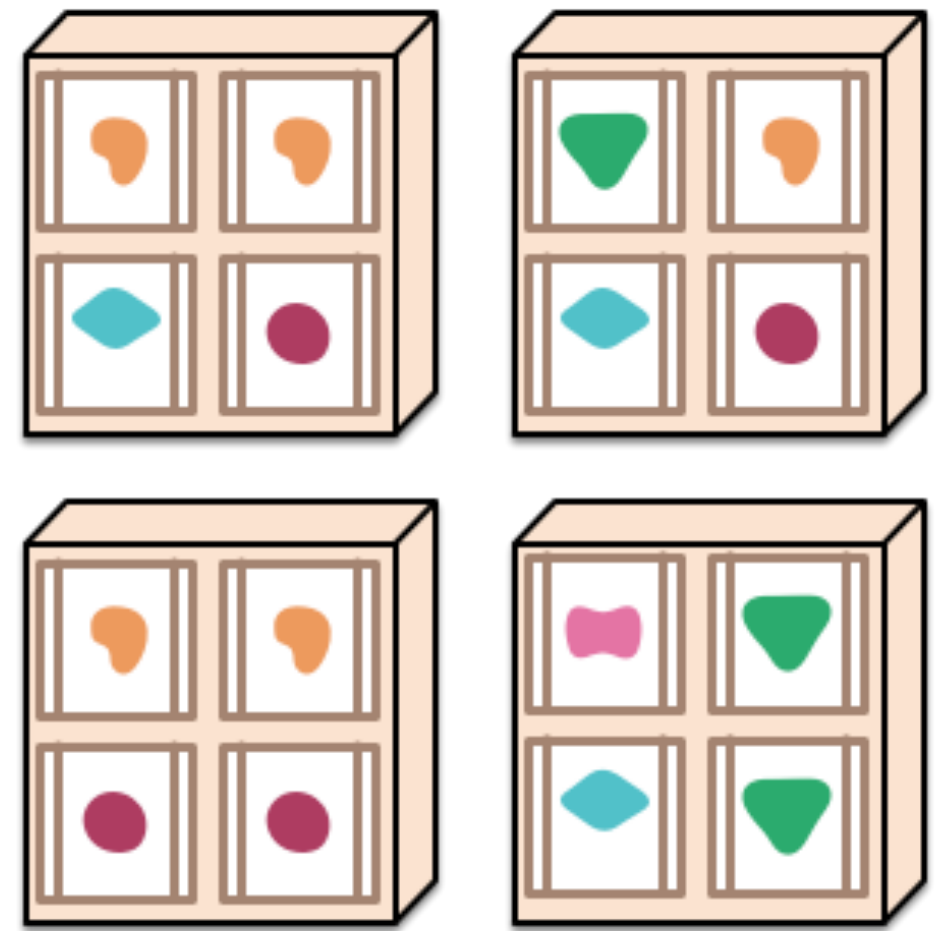
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*A microservices architecture puts each element of functionality into a separate service...*



*... and scales by distributing these services across servers, replicating as needed.*

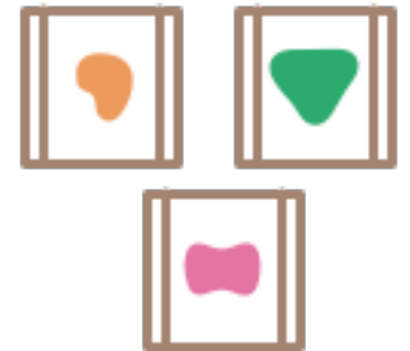


Read more: <https://martinfowler.com/articles/microservices.html>

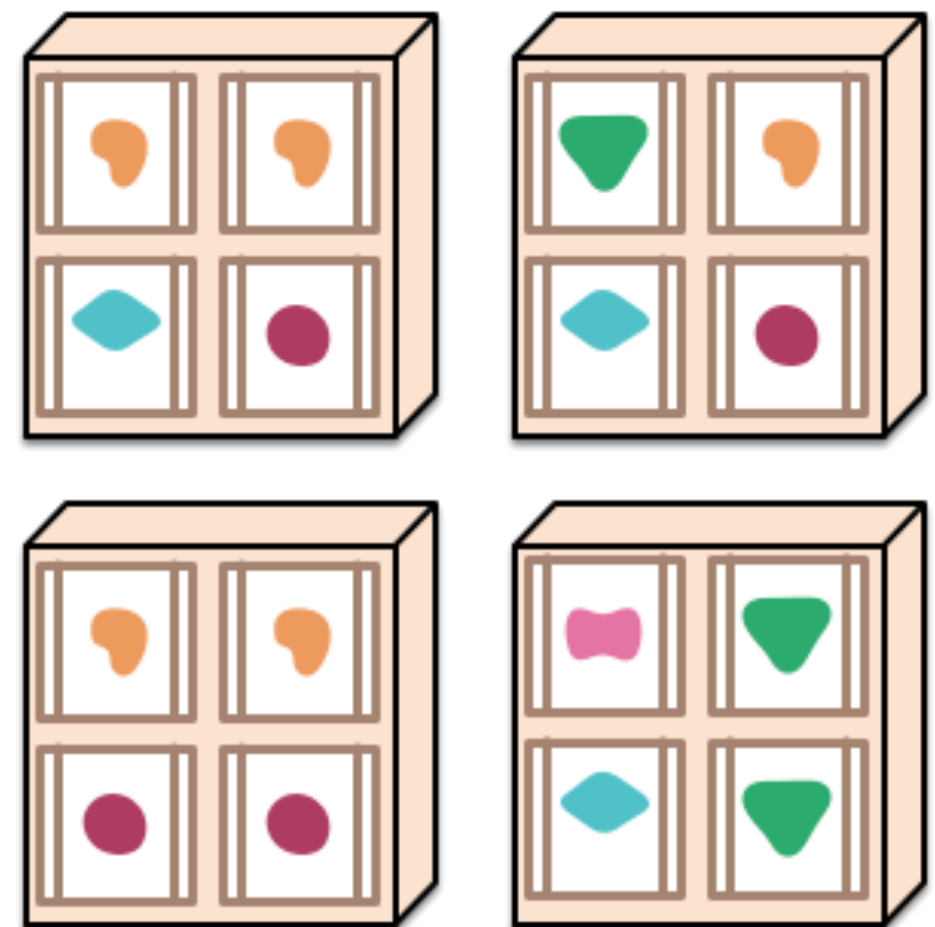
# Microservices

## Challenges with Microservices approach?

*A microservices architecture puts  
each element of functionality into a  
separate service...*



*... and scales by distributing these services  
across servers, replicating as needed.*



# Microservices Challenges

**Discovery:** how to find a service you want?

**Scalability:** how to replicate services for speed?

**Openness:** how to agree on a message protocol?

**Fault tolerance:** how to handle failed services?

All distributed systems face these challenges, microservices just increases the scale and diversity...



# Netflix

26th most popular website according to Alexa

Zero of their own servers

- All infrastructure is on AWS (2016-2018)
- Recently starting to build out their own Content Delivery Network

**NETFLIX** is **15%**  
of the total downstream volume of traffic  
across the entire internet

# Netflix

One of the first to really push microservices

- Known for their DevOps
- Fast paced, frequent updates, must always be available

700+ microservices

Deployed across  
10,000s of VMs and  
containers

## Netflix ecosystem

100s of microservices

1000s of daily production changes

10,000s of instances

100,000s of customer interactions per minute

1,000,000s of customers

1,000,000,000s of metrics

10,000,000,000 hours of streamed

**10s of operations engineers**

Netflix tech talk: <https://www.youtube.com/watch?v=CZ3wluvmHeM>

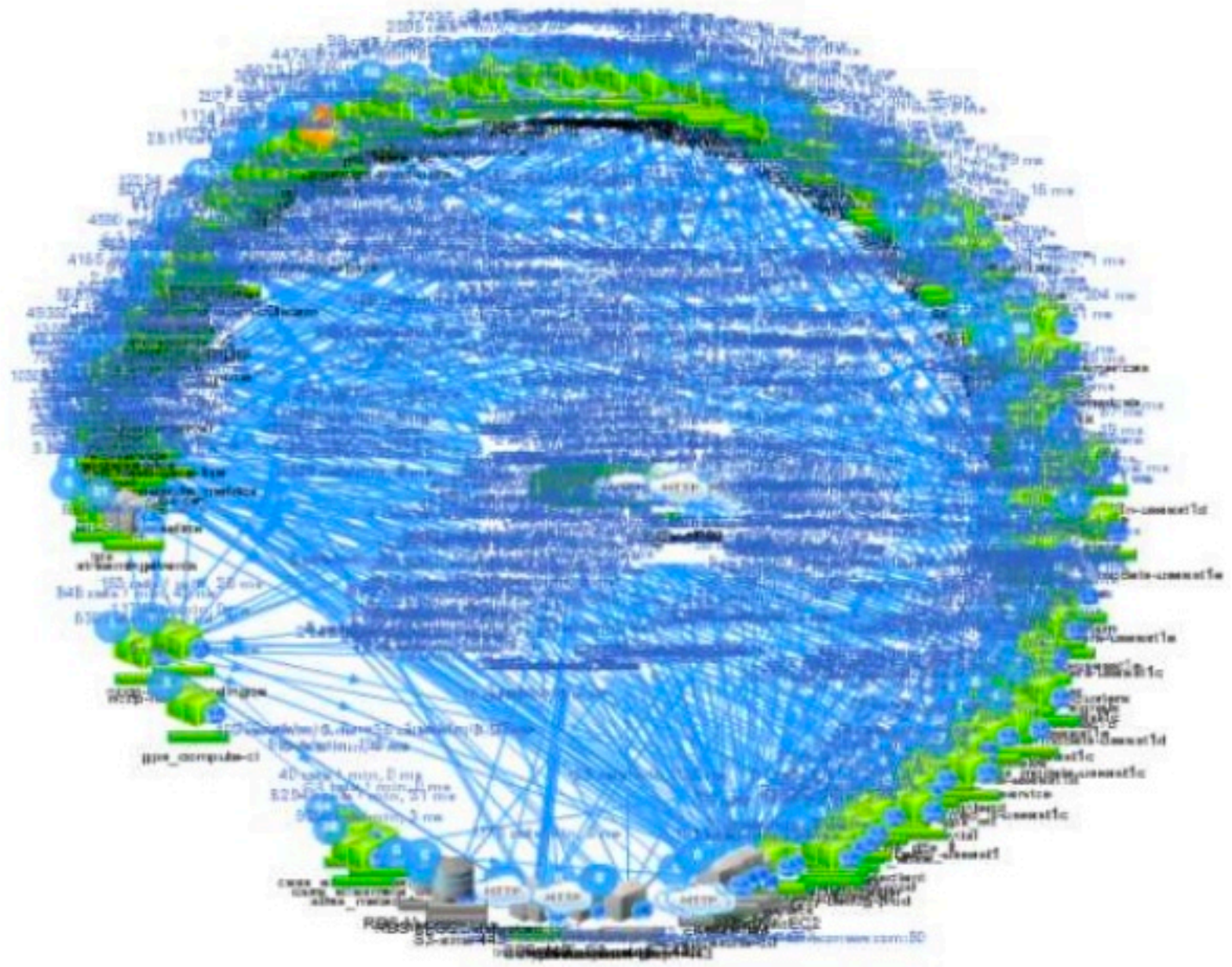


# Netflix “Deathstar”

Microservice architecture results in a extremely distributed application

- Can be very difficult to manage and understand how it is working at scale

How to know if everything is working correctly?



# Netflix Chaos Monkey

Idea: If my system can handle failures, then I don't need to know exactly how all the pieces themselves interact!

## Chaos Monkey:

- Randomly terminate VMs and containers in the production environment
- Ensure that the overall system keeps operating
- Run this 24/7



Make failures the common case, not an unknown!

<http://principlesofchaos.org/>

# Distributed Systems

## CS6421

### **Scaling the Web (Part 2)**

Prof. Tim Wood



# Serverless Computing

Trendy architecture that improves the agility of microservices

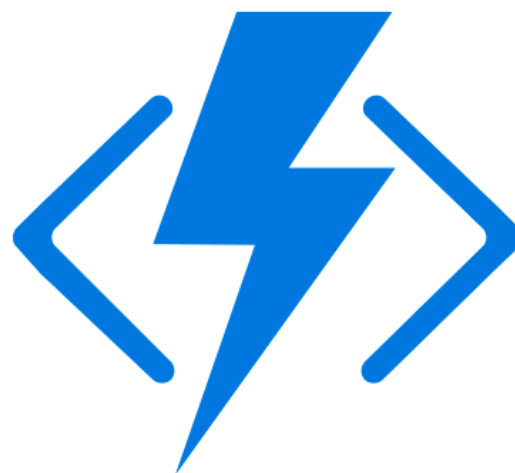
What does “serverless” mean?



APACHE  
OpenWhisk™



AWS Lambda



Azure  
Functions



Google Cloud Functions

# Serverless Computing

Trendy architecture that improves the agility of microservices

What does “serverless” mean?

You still need a server!

BUT, your services will not always be running

Key idea: only instantiate a service when a user makes a request for that functionality

How will this work for stateful vs stateless services?

# Serverless Startup

## AWS Lambda

- Define a stateless “function” to execute for each request
- A container will be instantiated to handle the first request
- The same container will be used until it times out or is killed

**No workload means no resources being used!**

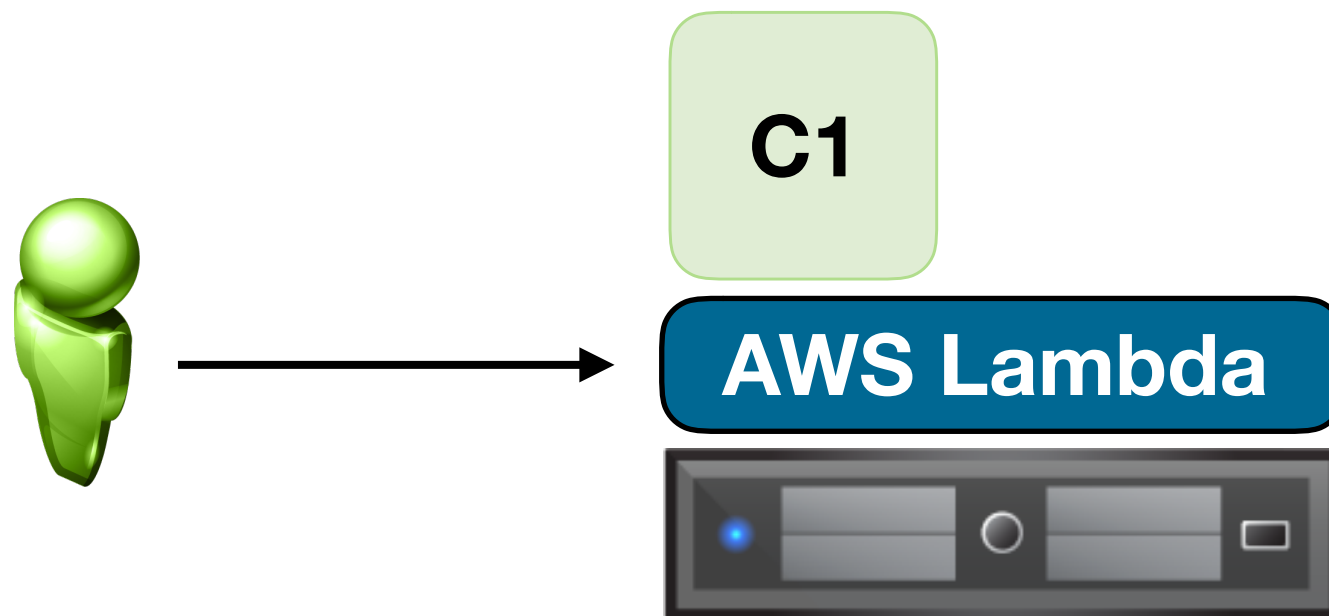


# Serverless Startup

## AWS Lambda

- Define a stateless “function” to execute for each request
- A container will be instantiated to handle the first request
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Request arrives, start green container

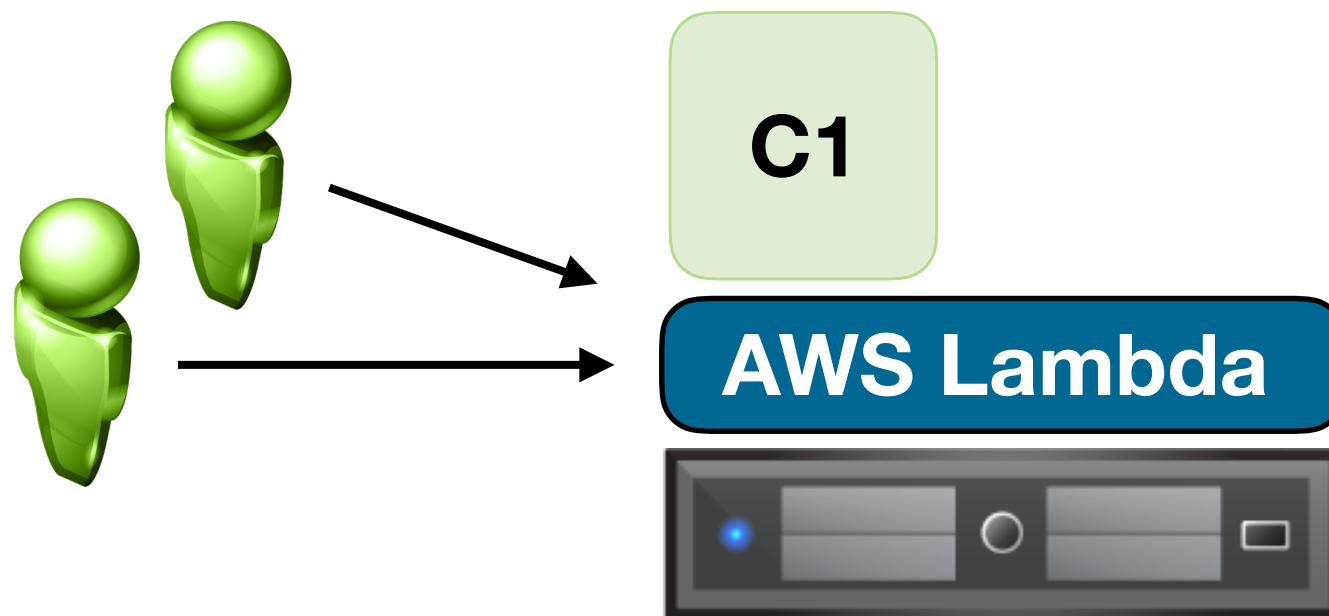


# Serverless Startup

## AWS Lambda

- Define a stateless “function” to execute for each request
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Reuse that container for subsequent requests



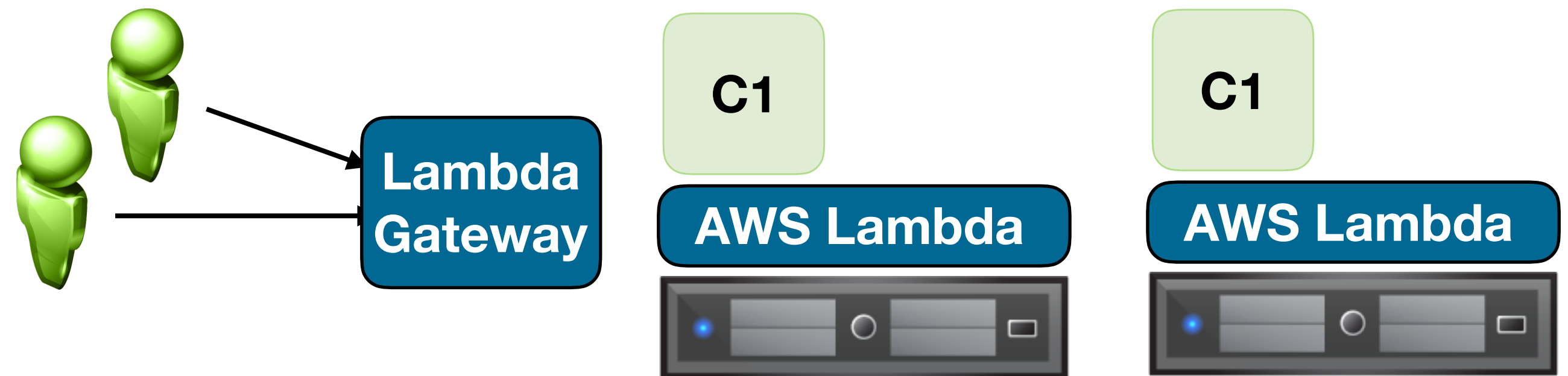


# Serverless Startup

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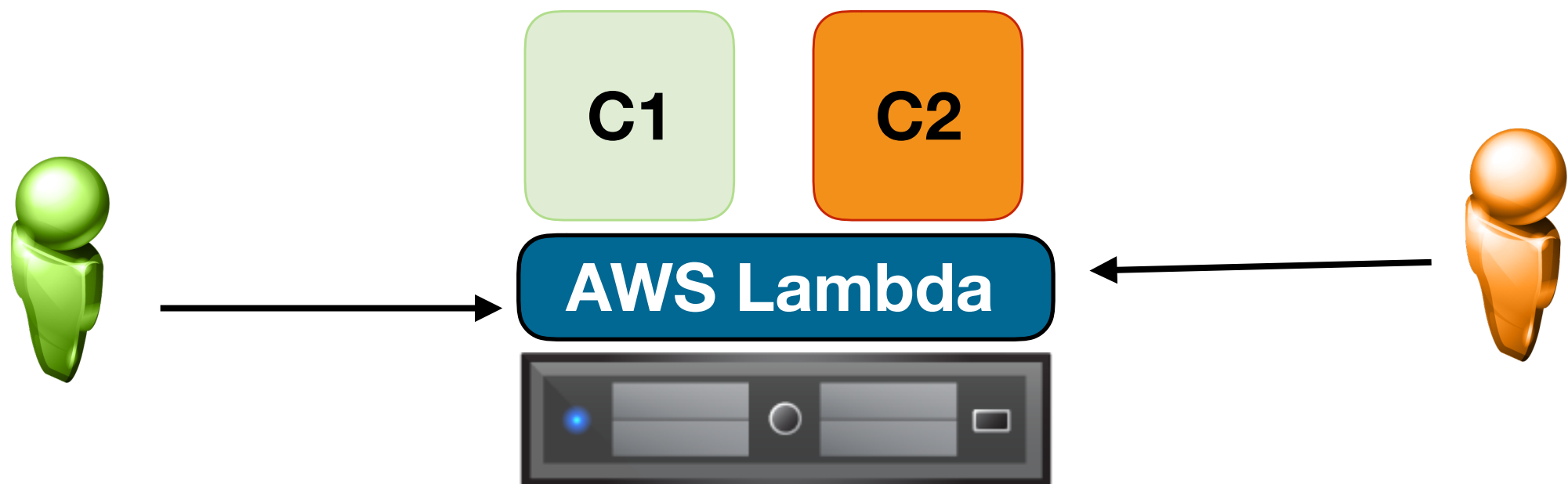


# Serverless Startup

## AWS Lambda

- Define a stateless “function” to execute for each request
- A container will be instantiated to handle the first request
- The same container will be used until it times out or is killed

Start new container if user needs a different function

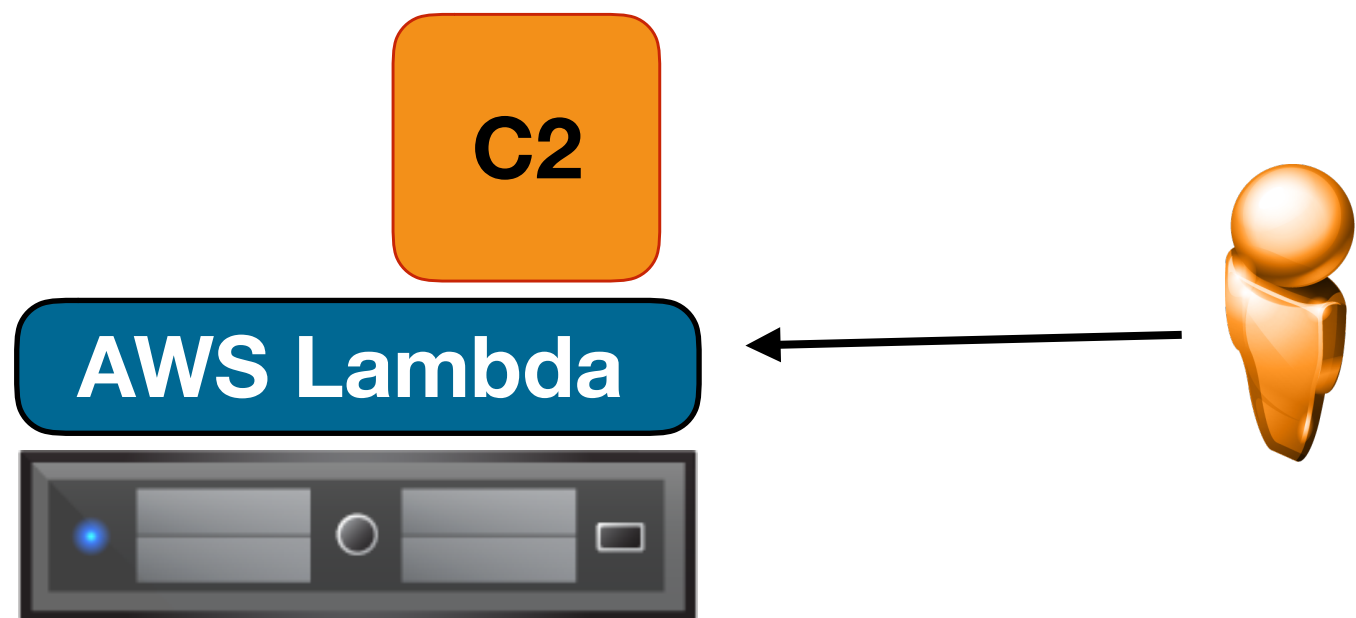


# Serverless Startup

## AWS Lambda

- Define a stateless “function” to execute for each request
- A container will be instantiated to handle the first request
- The same container will be used until it times out or is killed

Clean up old containers once not in use



# Serverless Pros/Cons

## Benefits:

- Simple for developer when auto scaling up
- Pay for exactly what we use (at second granularity)
- Efficient use of resources (auto scale up and down based on requests)
- don't worry about reliability/server management at all

## Drawbacks:

- Limited functionality (stateless, limited programming model)
- High latency for first request to each container
- Some container layer overheads plus the lambda gateway and routing overheads
- Potentially higher and unpredictable costs
- Difficult to debug / monitor behavior
- Security

# Serverless Pros/Cons

Benefits:

Drawbacks:



# Scaling



# Two ways to scale

## Scale UP (vertical)

- Buy a bigger computer



Can only grow so big

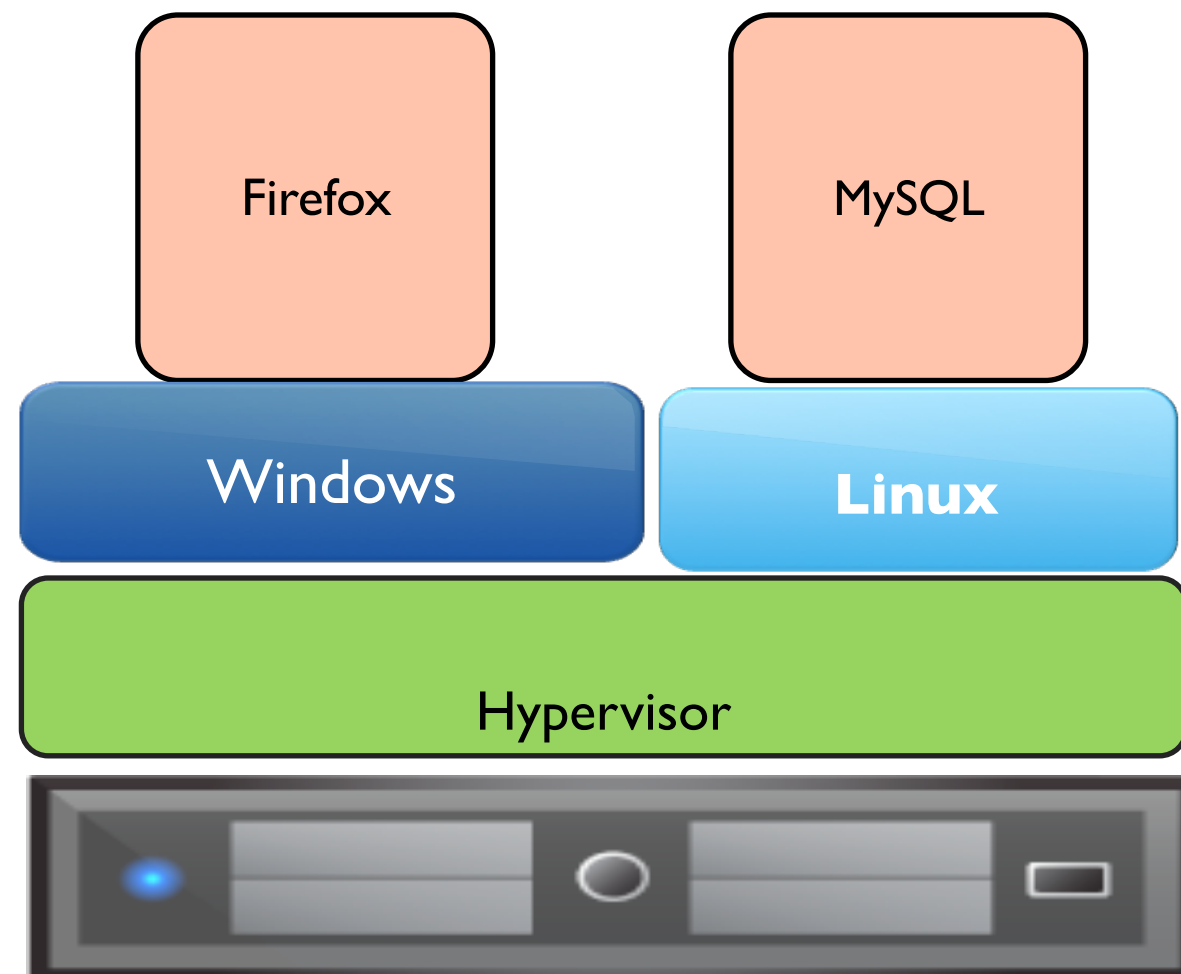
## Scale OUT (horizontal)

- Buy multiple computers



How to spread work? How to keep data consistent?

# Does virtualization help?



# Does virtualization help?

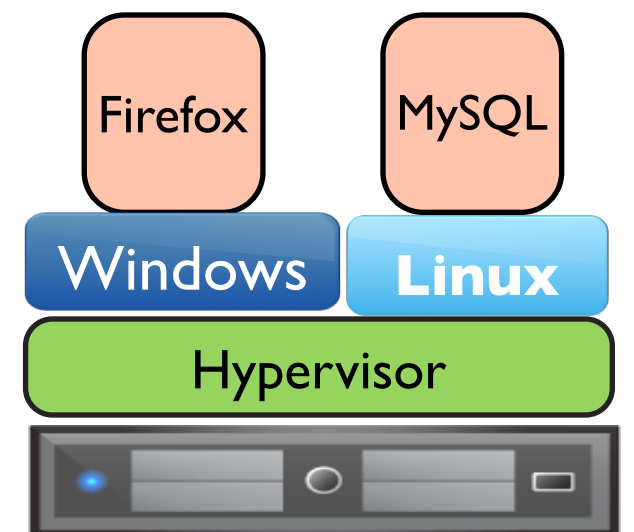
Not exactly...

Virtualization divides something big into smaller pieces

but still has features which can assist with scalability:

- Easy replication of VM images
- Dynamic resource management

Simplifies scale OUT, but has limits on how much you can scale UP



# Replication

Scale Out v1



# Biggest Challenge: Consistency

Replicating data makes it faster to access



**Computer science or computing science**  
(abbreviated **CS** or **compsci**)  
designates the **scientific** and **mathematical** approach in  
information technology and  
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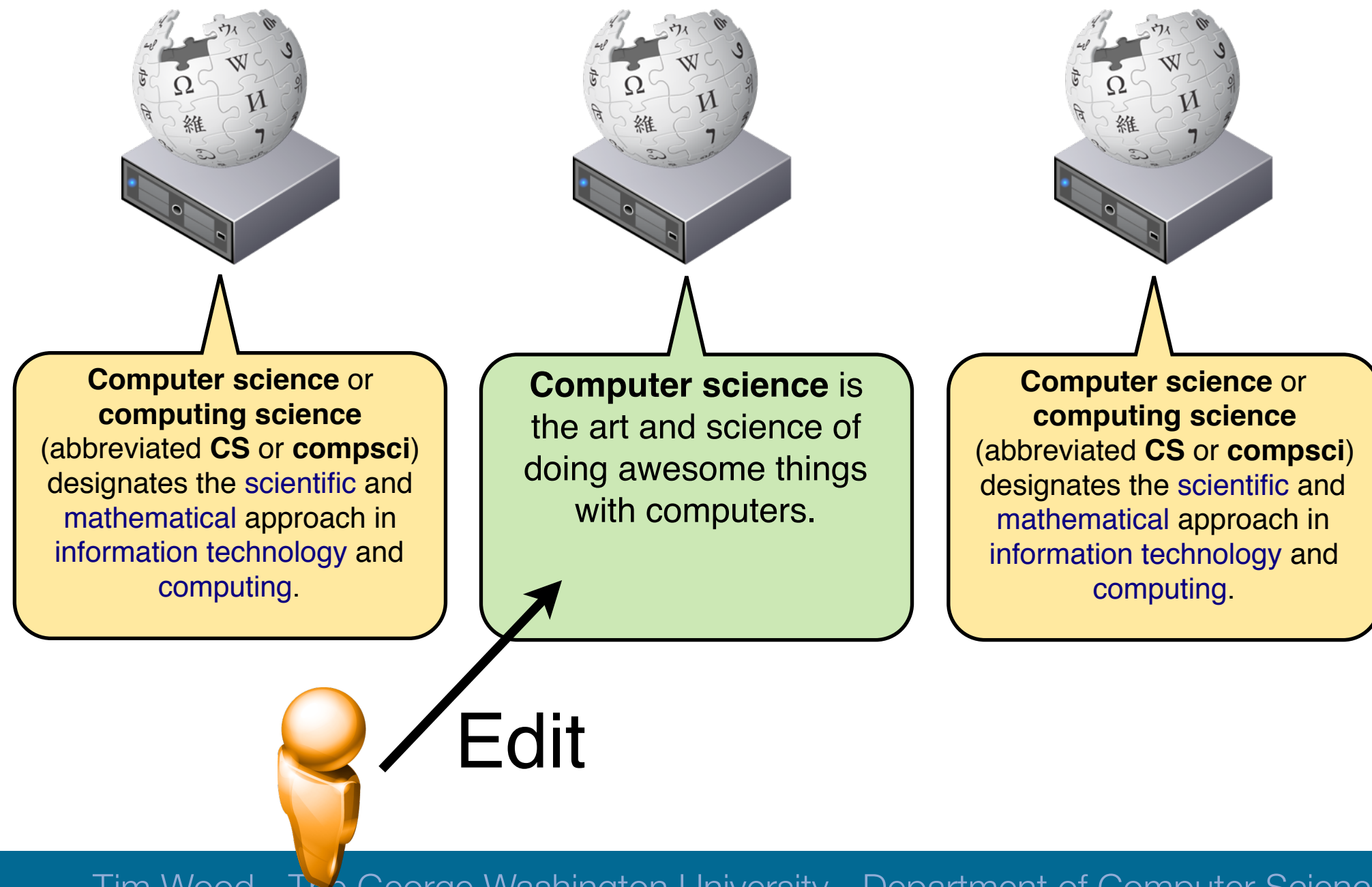


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# Biggest Challenge: Consistency

Replicating data makes it faster to access

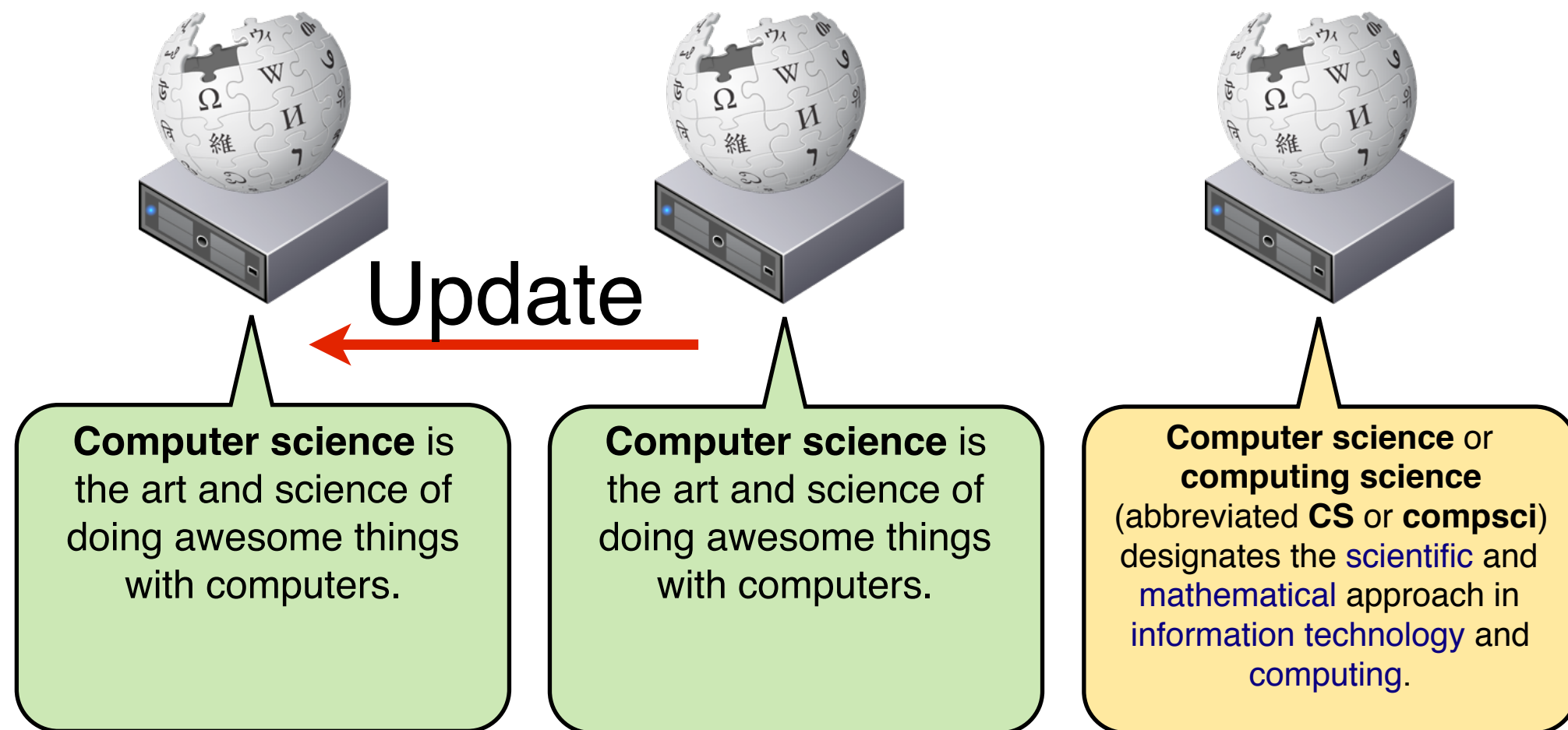
- But how to keep all copies of data consistent?



# Biggest Challenge: Consistency

Replicating data makes it faster to access

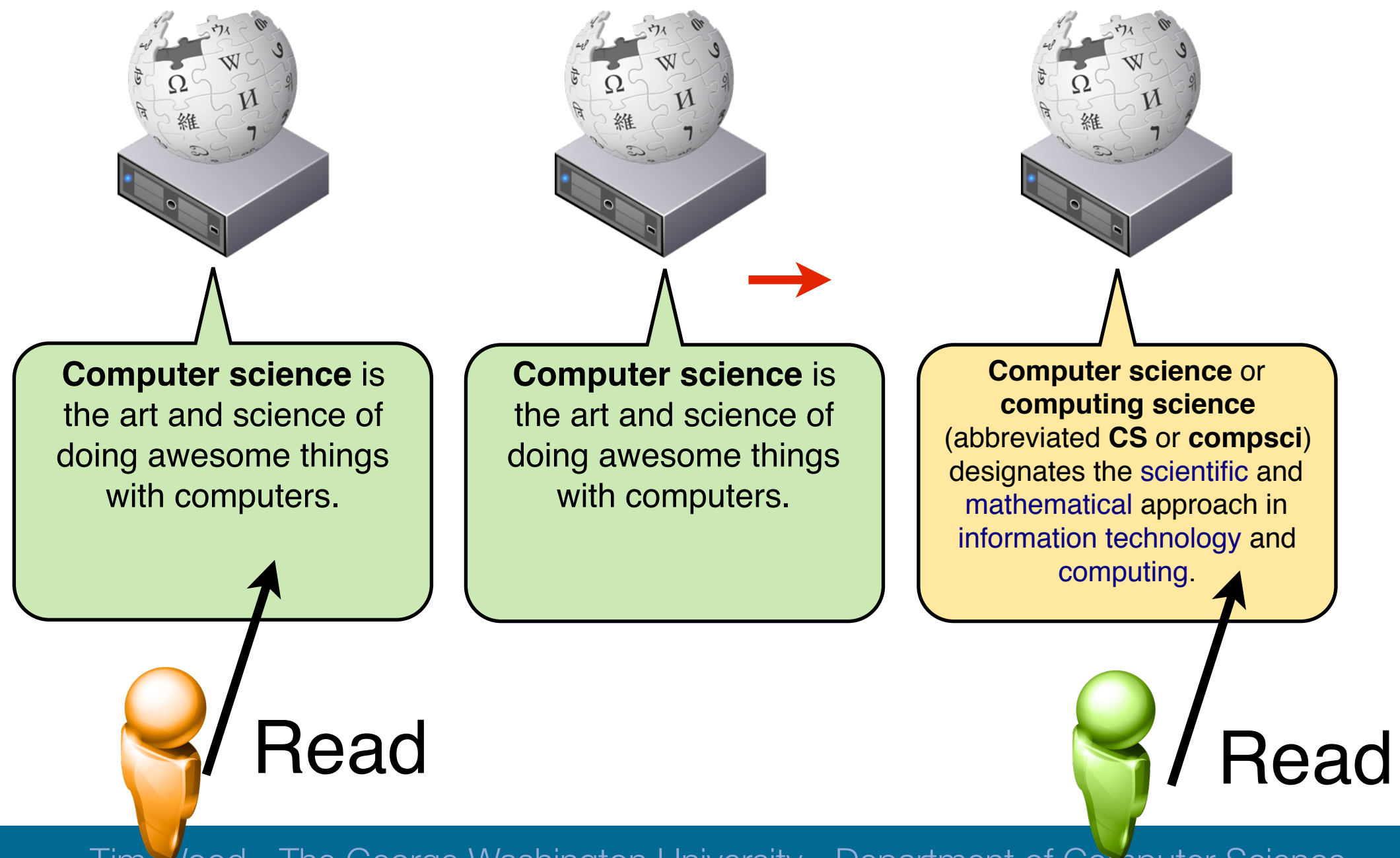
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# Biggest Challenge: Consistency

Replicating data makes it faster to access

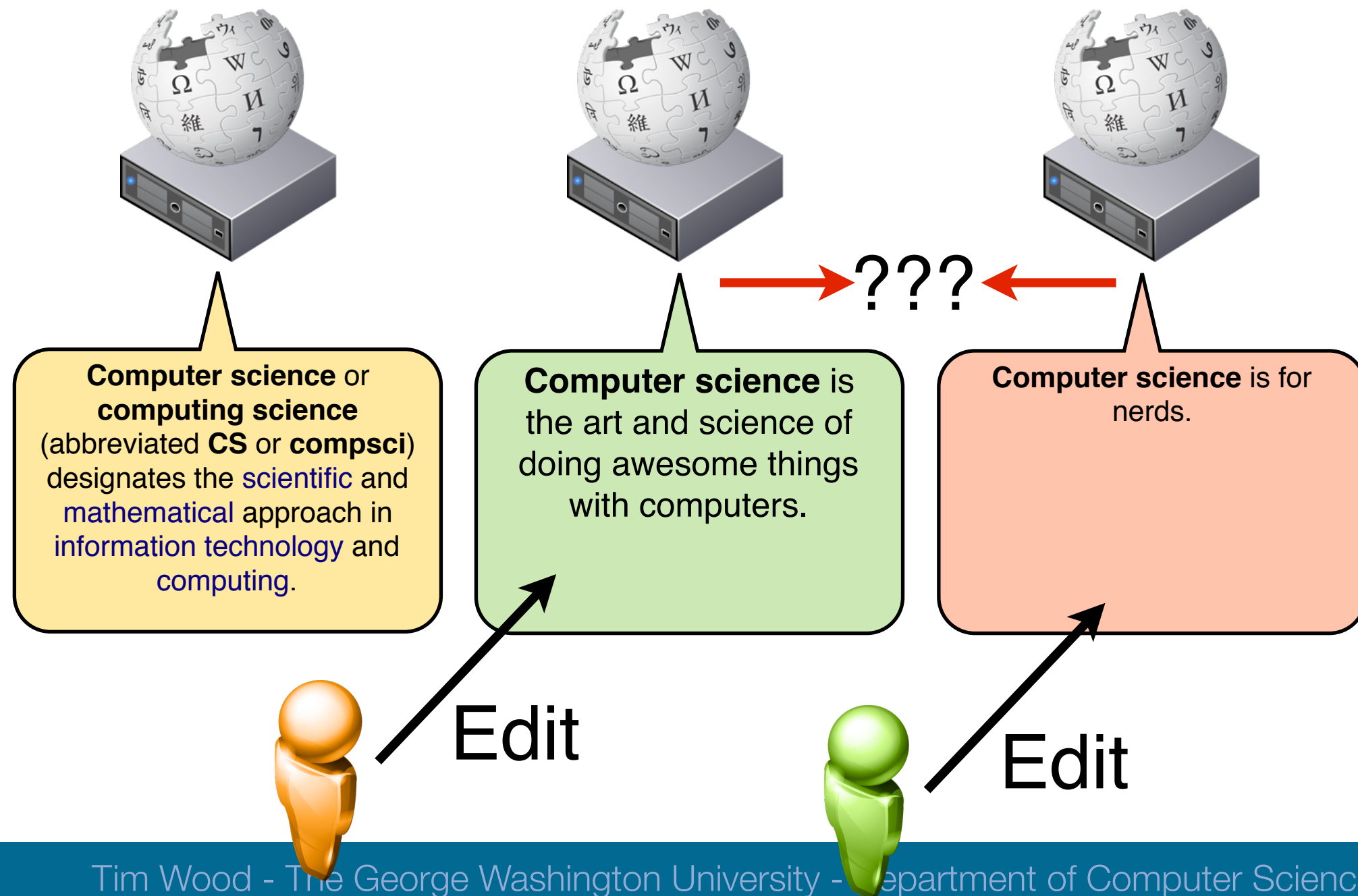
- But how to keep all copies of data consistent?



# Biggest Challenge: Consistency

Writes are even harder

- Would need time stamps or a consistent ordering
- Or, if writes are rare, just have a master coordinate





# Does it Matter?

A slightly out of date wikipedia page?

A post to your facebook profile?

1. Remove boss from friends list
2. Post "My boss is a moron, I want a new job!"

A change to a stock price in the NASDAQ exchange?

# Providing Consistency

We have already seen techniques that will help:

- Version vectors
- Distributed locking based on Lamport Clocks
- Election-based systems with a master/slave setup

There are many different types of **consistency**

- **Strict** - updates immediately available after a write
- **Sequential** - result of parallel updates needs to have the same effect as if they had been done sequentially
- **Causal** - updates that are casually related (e.g., where vector clocks can prove the  $\rightarrow$  relationship) are ordered sequentially, but others may not be  
... (several more) ...
- **Eventual** - updates will converge so at some point reads to any replica will get the same result

# End of Semester

Practice 2 / Projects — Sunday 12/9 (extended)

Exam — Friday 11/30

- Concepts from lecture
- 8.5x11 page (two sided) of hand written notes
- I will post some practice questions this weekend

# Partitioning

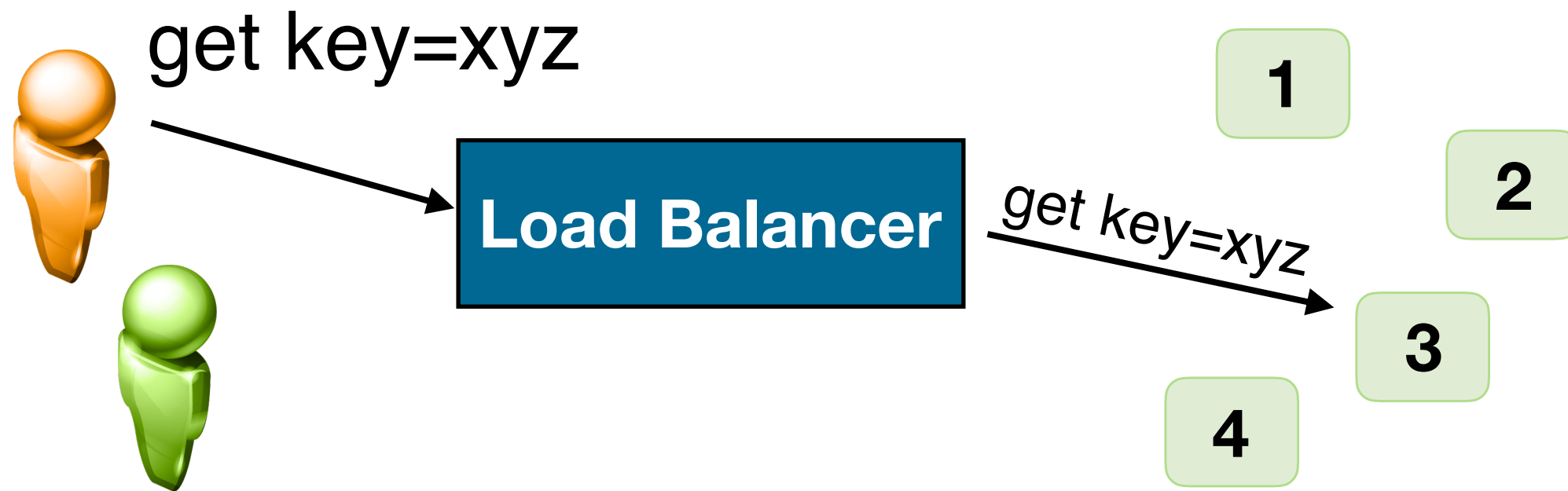
Scale Out v2

# Spread data across servers

Useful if all data does not fit on one server

Let's consider a Key Value store like Memcached

- Lots of data to store
- Consistency is not that important
- Might need to add or remove nodes to the cluster
- **How should we partition the keys across the nodes?**





# DHTs

A **Distributed Hash Table** is a key-value store that can be implemented in a **Peer-2-Peer** fashion.

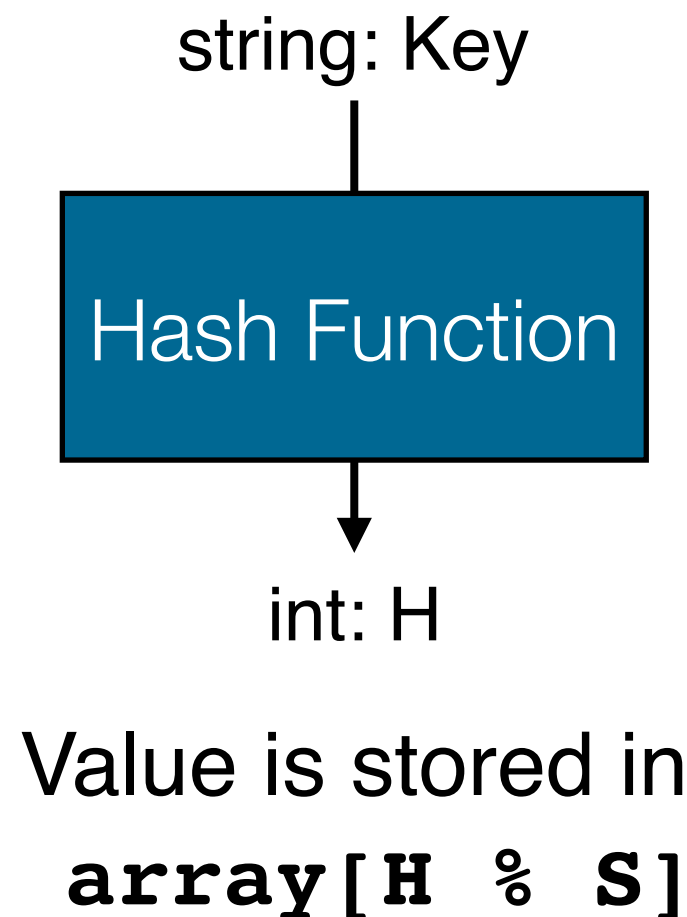
Goals:

- Evenly **partition** data across the nodes
- Efficient lookups
- Gracefully handle nodes leaving and joining

# DHTs

A **Distributed Hash Table** is a key-value store that can be implemented in a P2P fashion.

## Simple Hash Table



S = array size

Array Index	Value
1	v1
2	v2
...	...
S	vs

# DHTs

What if one node can't fit all the data?

Do two hash lookups!

## Simple DHT

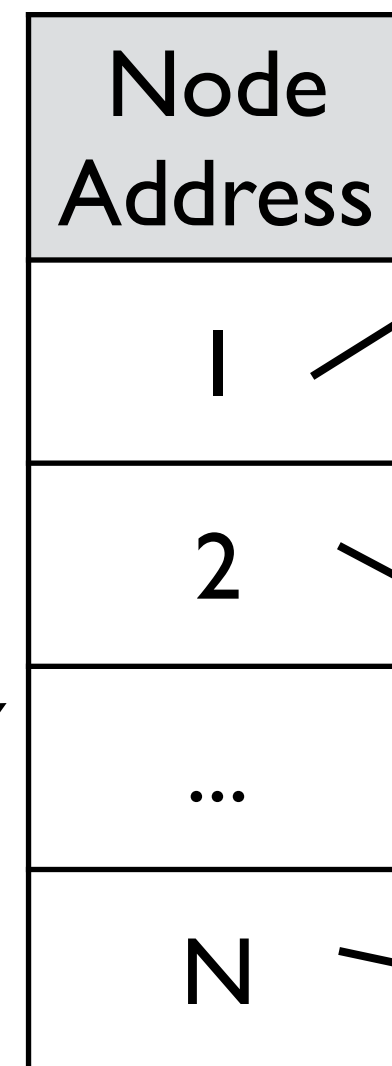
string: Key



int: H

Value is stored on  
**node[H % N]**

N = # of nodes



S = array size

Array Index	Value
1	v1
2	v2
...	...
S	vn

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# DHTs

When will this perform poorly?

## Simple DHT

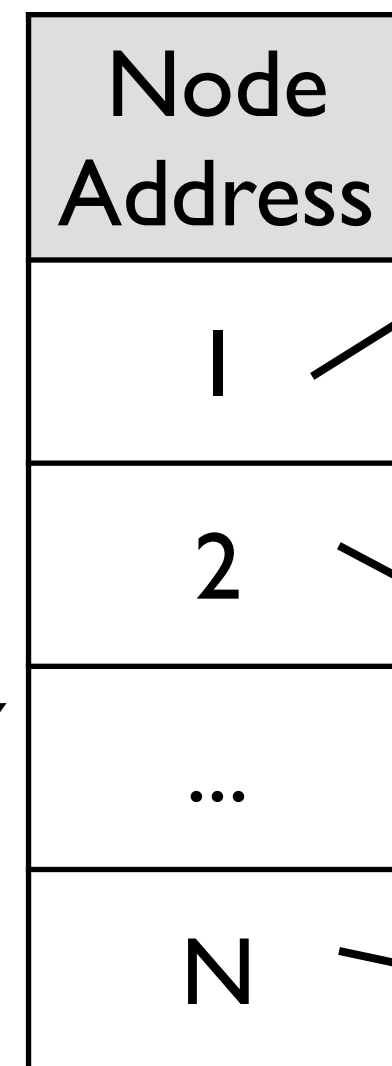
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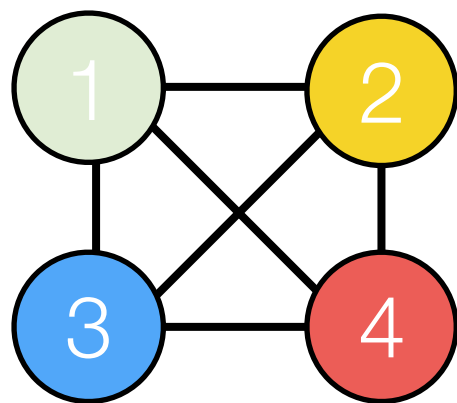
# Churn

Churn is when nodes are frequently joining or leaving

- In a DHT it is OK to lose data when a node leaves, but it shouldn't cause all other nodes to reshuffle their data!

## Simple DHT

Value is stored on  
 **$\text{node}[\text{H} \% 4]$**



Divides hash space  
into 4 equal partitions  
for 4 servers

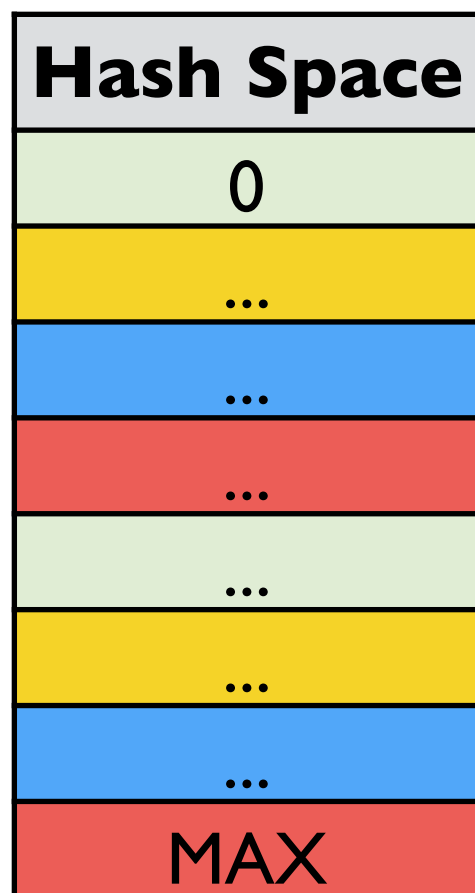


# Churn

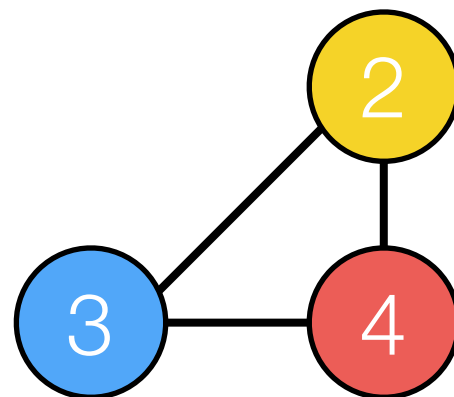
Churn is when nodes are frequently joining or leaving

- In a DHT it is OK to lose data when a node leaves, but it shouldn't cause all other nodes to reshuffle their data!

## Simple DHT



Oops!  
Green node  
failed!



All nodes  
needs to be  
reorganized!

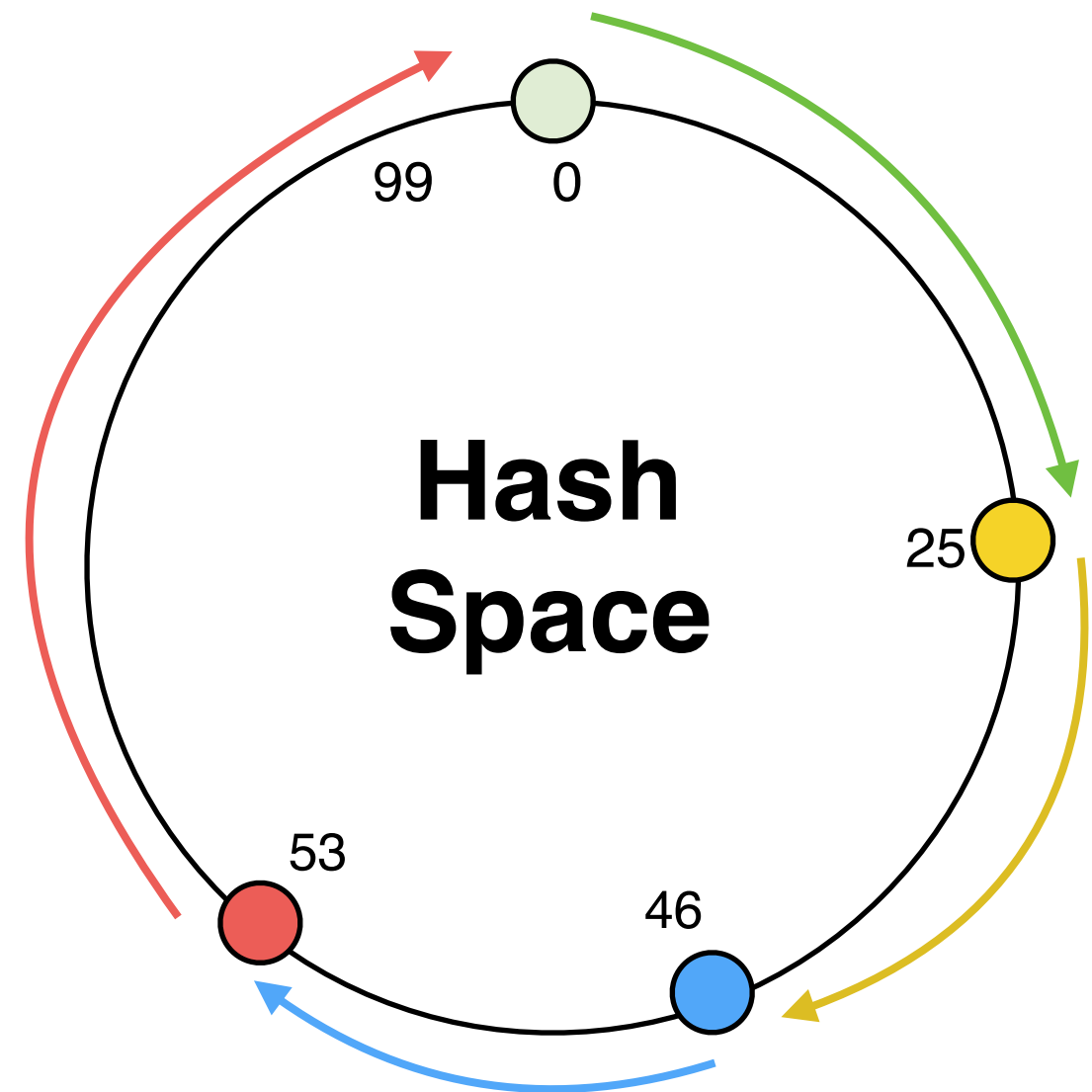
# Chord DHT Architecture

Think of hash space as a ring

Nodes pick a random ID when they join: 0 to MAX-1

Nodes are assigned contiguous portions of the ring starting at their ID until they reach the subsequent node

**Will this evenly divide up the hash space?**

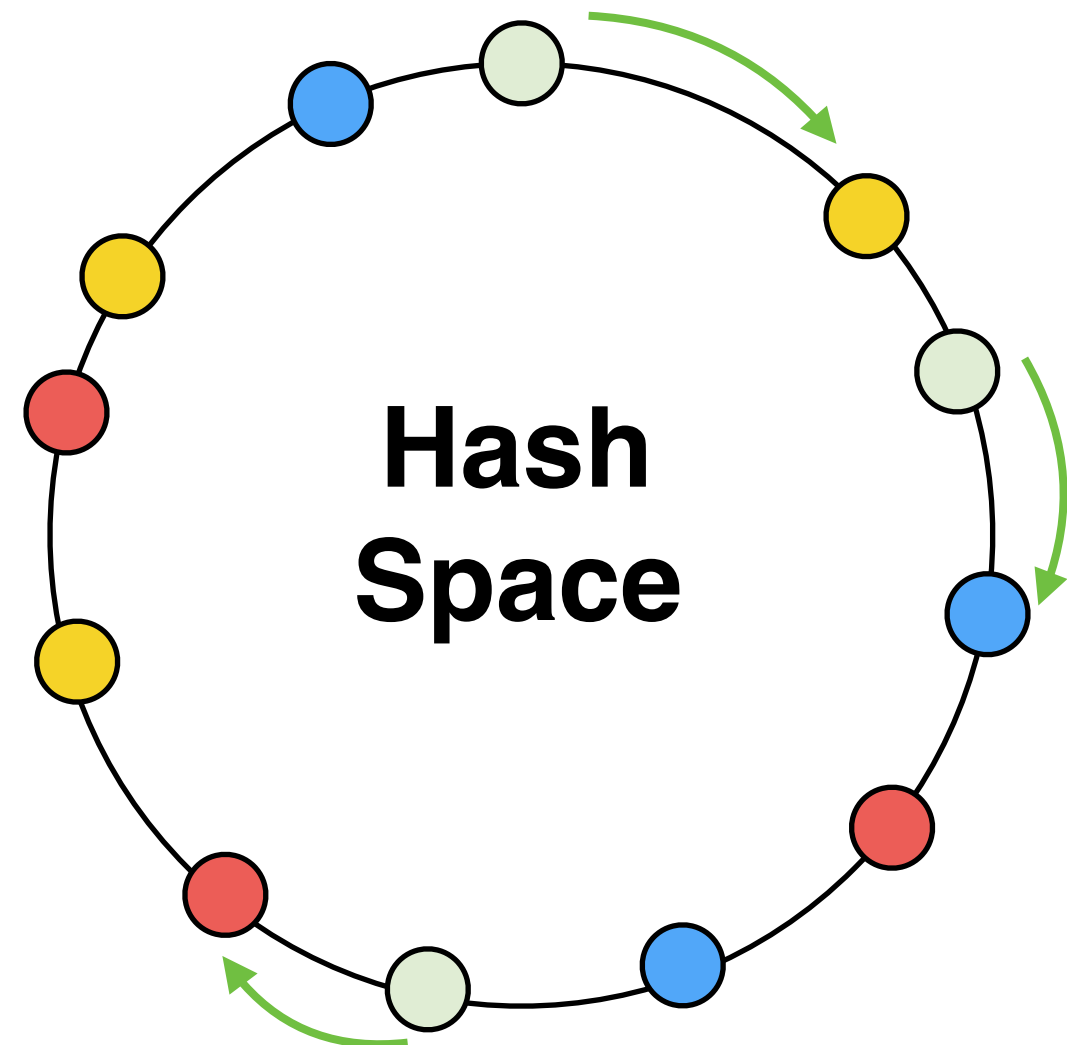


# Chord DHT Architecture

**Will this evenly divide up the hash space?**

If we have a lot of nodes, **probably** yes!

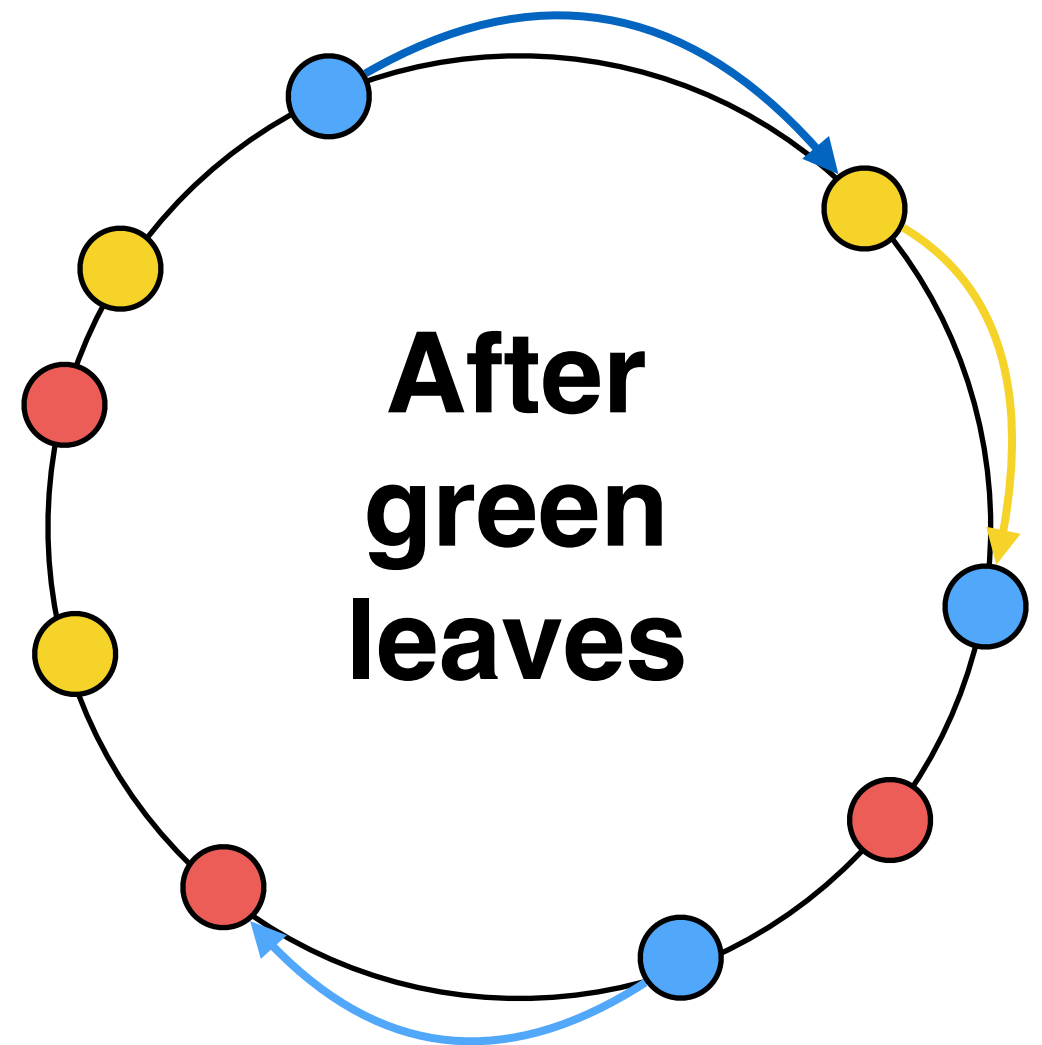
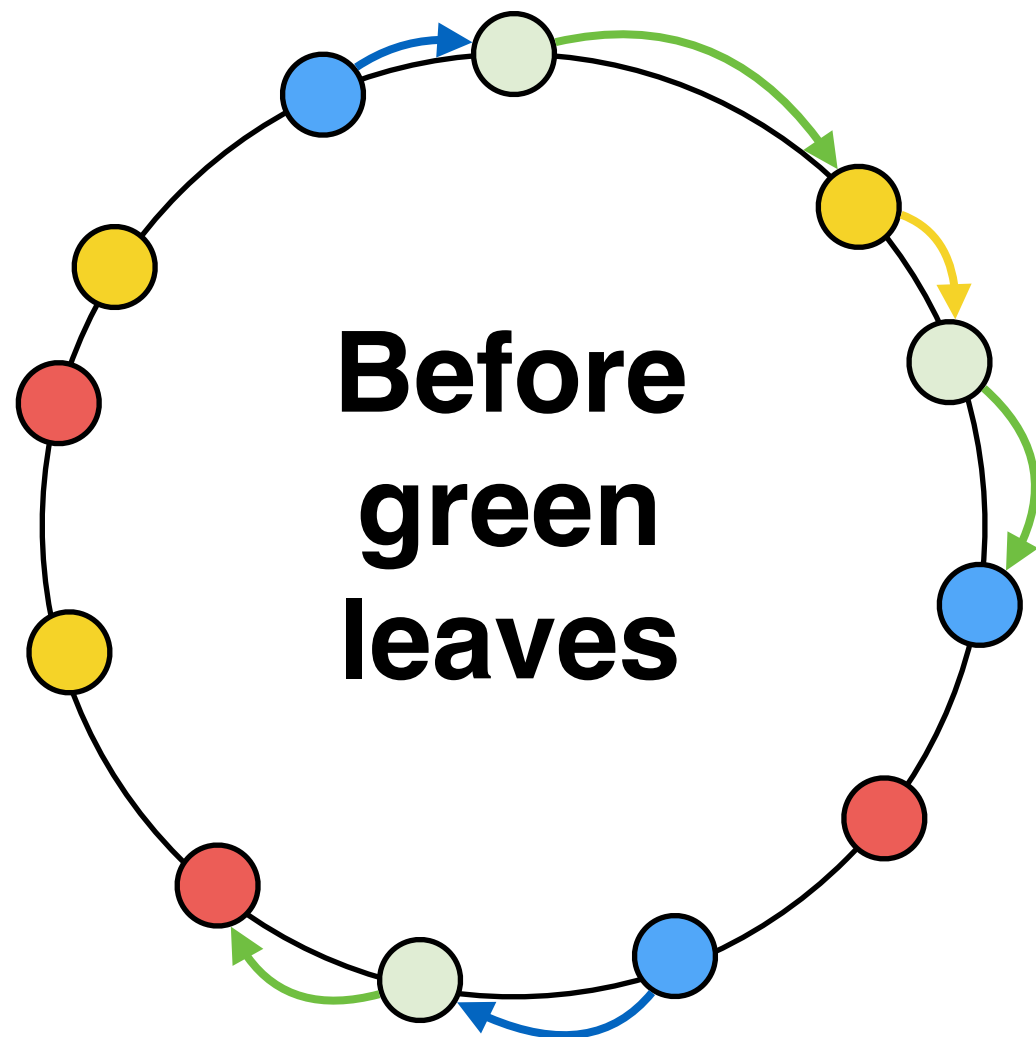
Or, each node can claim multiple IDs (virtual nodes)



# Chord Churn

What happens when a node is removed?

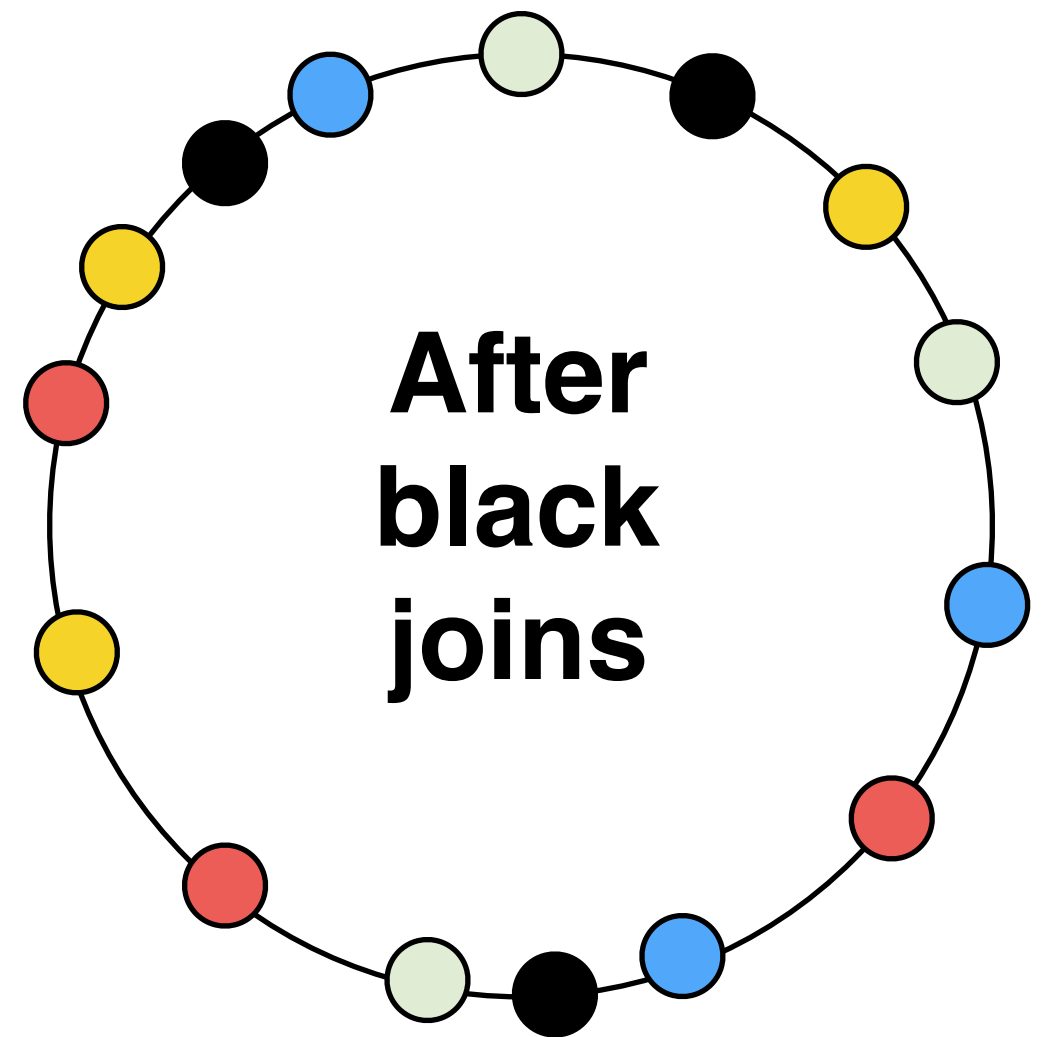
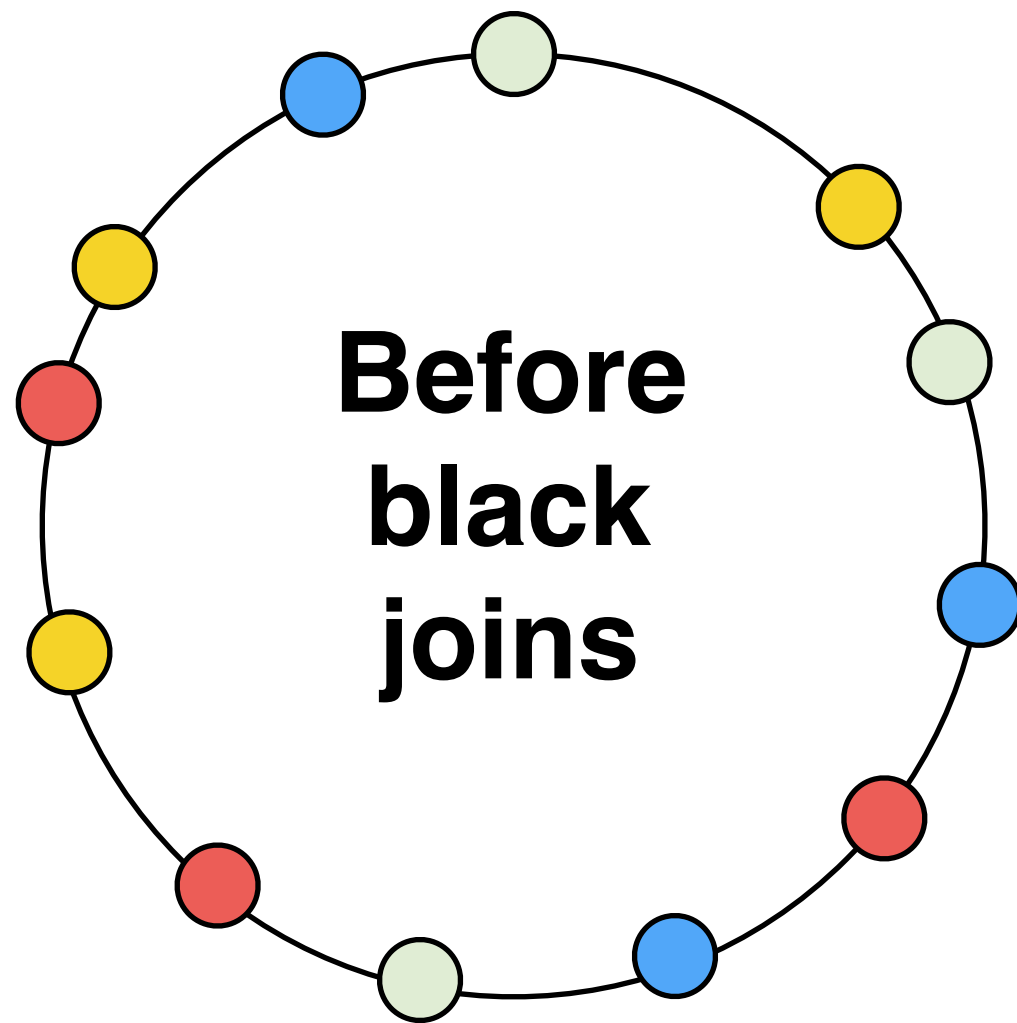
- How many nodes were affected?



# Chord Churn

What happens when a node is added?

- How many nodes were affected?

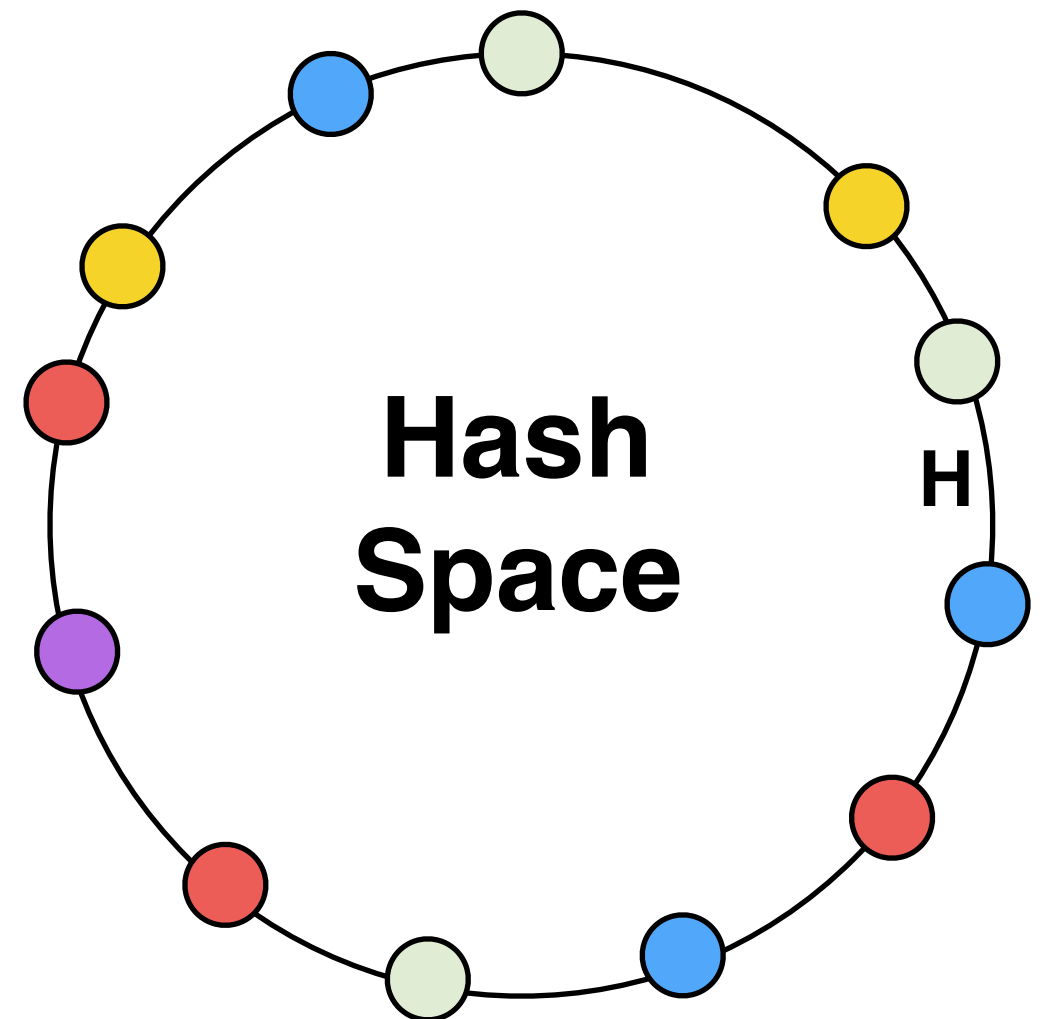




# Chord Lookups

Where can we find the key with hash  $H$ ?

How can the purple node get the data for  $H$ ?



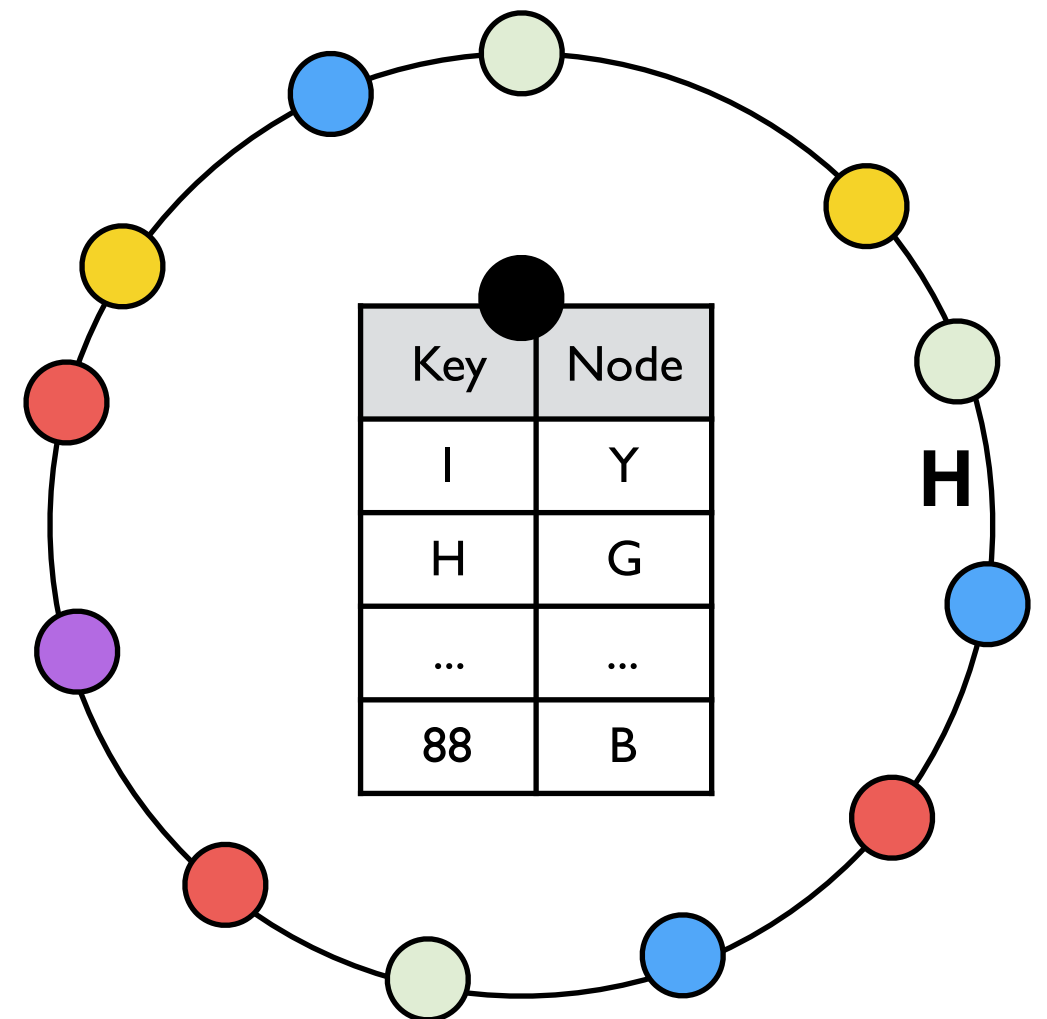
# Chord Lookups

Where can we find the key with hash  $H$ ?

How can the purple node get the data for  $H$ ?

## Options 0: Key Index Table

- Store the node holding each keys in a central server
- Directly access the node!
- If we have millions of keys this table will be really big!
- The node that manages the index table will be a centralized bottleneck!



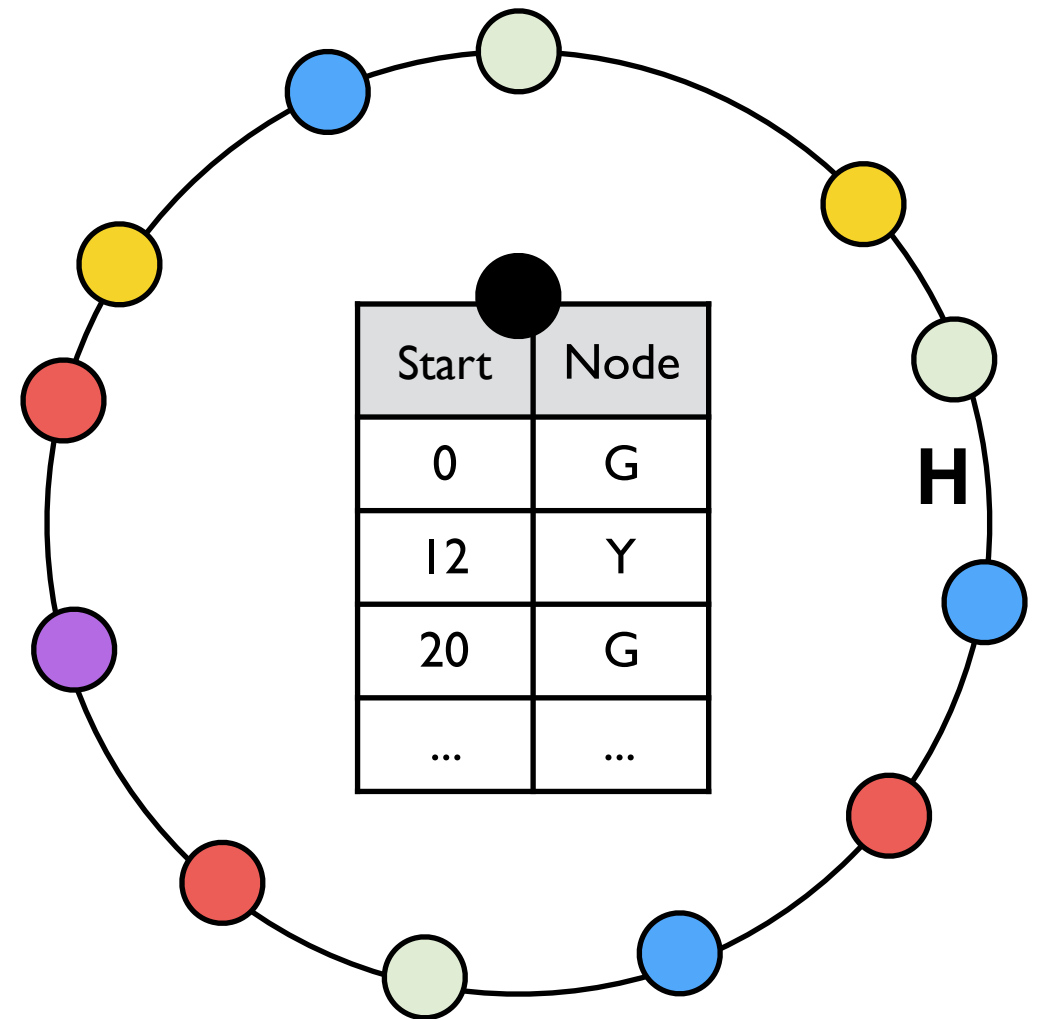
# Chord Lookups

Where can we find the key with hash  $H$ ?

How can the purple node get the data for  $H$ ?

## Options 1: Node Index Table

- Store the indices of all node IDs
- Find which ID is closest to  $H$
- Table is still very large and may be bottleneck!
- Also need to worry about consistently updating the table!



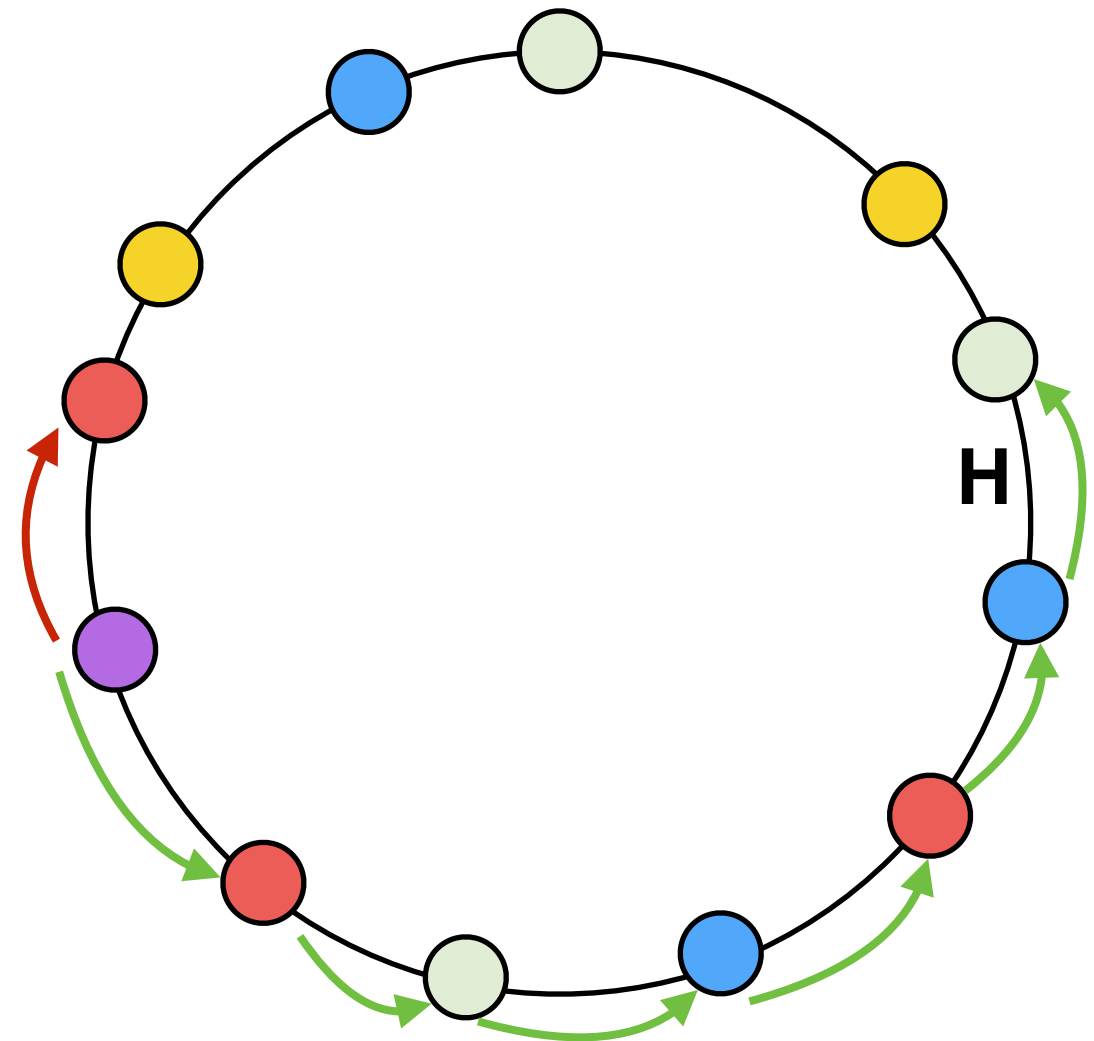
# Chord Lookups

Where can we find the key with hash  $H$ ?

How can the purple node get the data for H?

# Options 2: Neighbors

- Each node tracks its **successor** and **predecessor**
- If  $H > ID$ , ask successor else ask predecessor
- Requires minimal state
- Can take a long time to traverse the ring!  $O(N)$



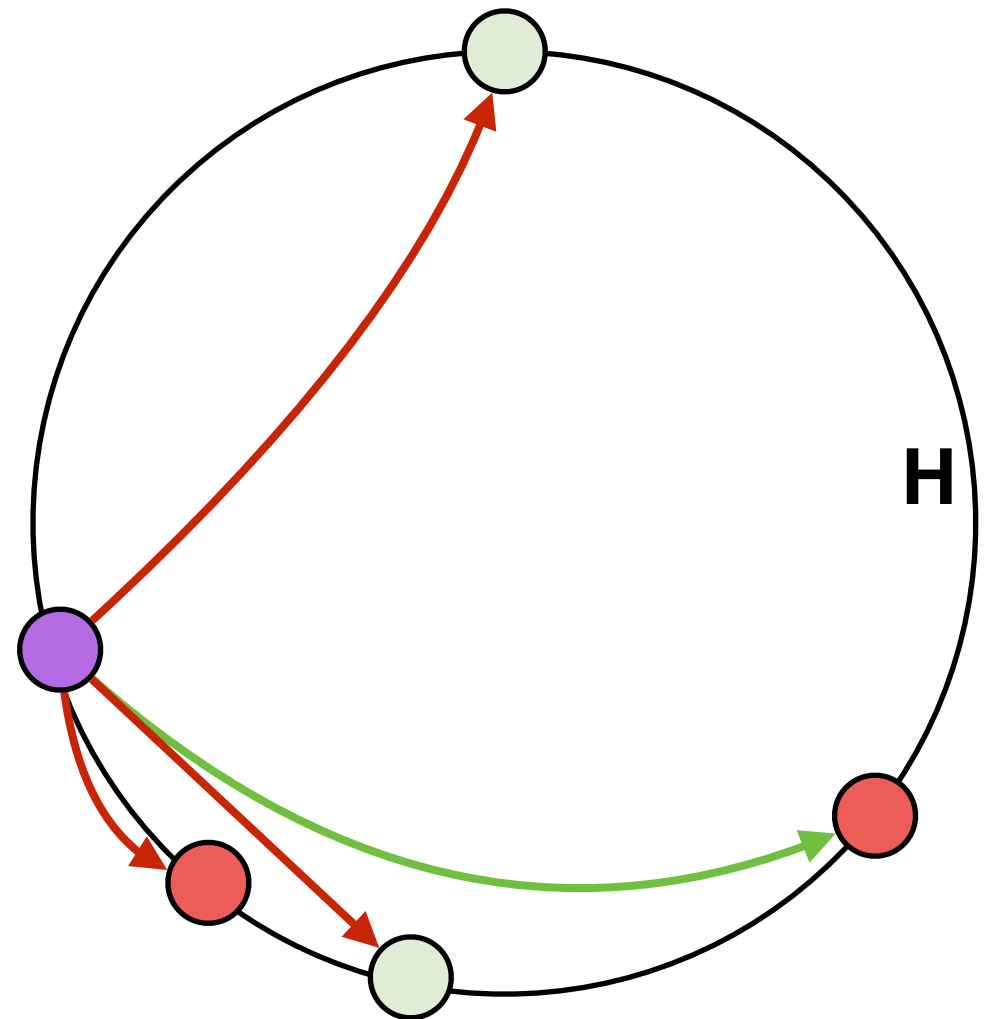
# Chord Lookups

Where can we find the key with hash  $H$ ?

How can the purple node get the data for  $H$ ?

## Options 3: Finger Tables

- Track  $m$  additional neighbors: successor  $2^0, 2^1, 2^2, \dots, 2^m$
- Jump to closest successor to find  $H$ , then jump again
- Requires minimal state
- Can find item in  $\log(N)$  steps



# Chord Lookups

Where can we find the key with hash  $H$ ?

How can the purple node get the data for  $H$ ?

## Options 3: Finger Tables

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