

Distributed Systems II

Software Architecture

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March 30, 2026

Distributed Systems Series

Distributed I *Reliability* and *scalability* of
stateless systems

Distributed II *Complexities* of *stateful*
systems

Distributed III *Hard problems* in distributed
systems

Distributed Systems Series

Distributed I Reliability and scalability of stateless systems

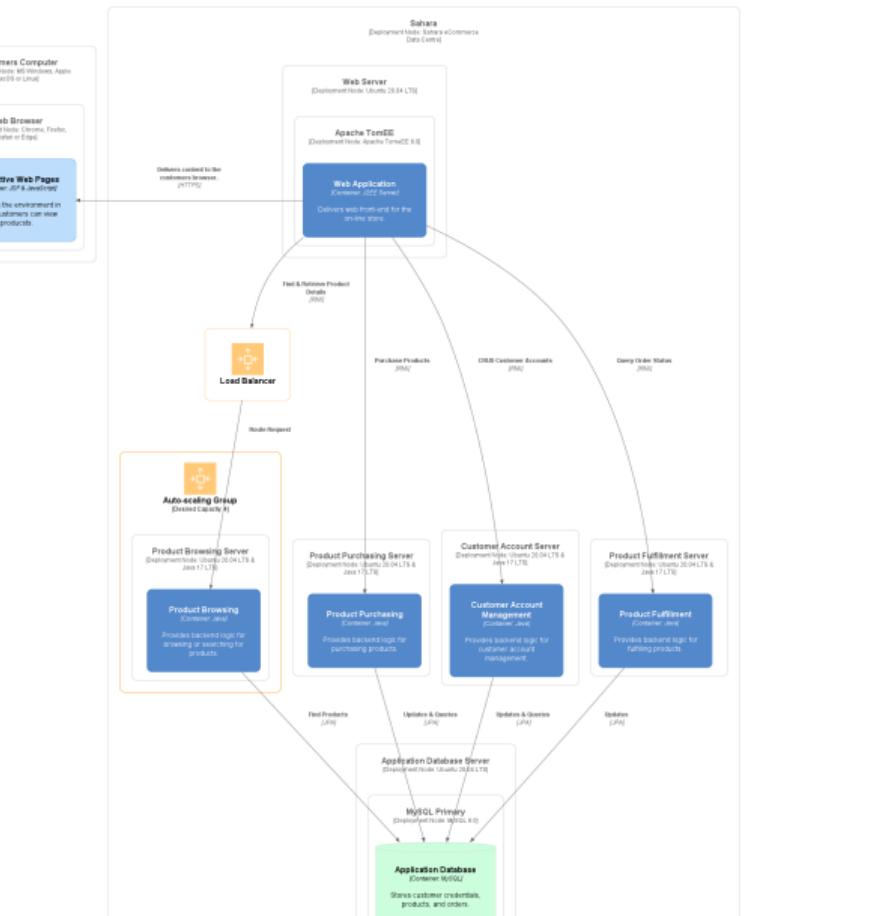
Distributed II *Complexities* of *stateful* systems

Distributed III Hard problems in distributed systems

Previously in Distributed I: Benefits...

- Improved *reliability*
- Improved *scalability*
- Improved *latency*

Previously in Distributed I...



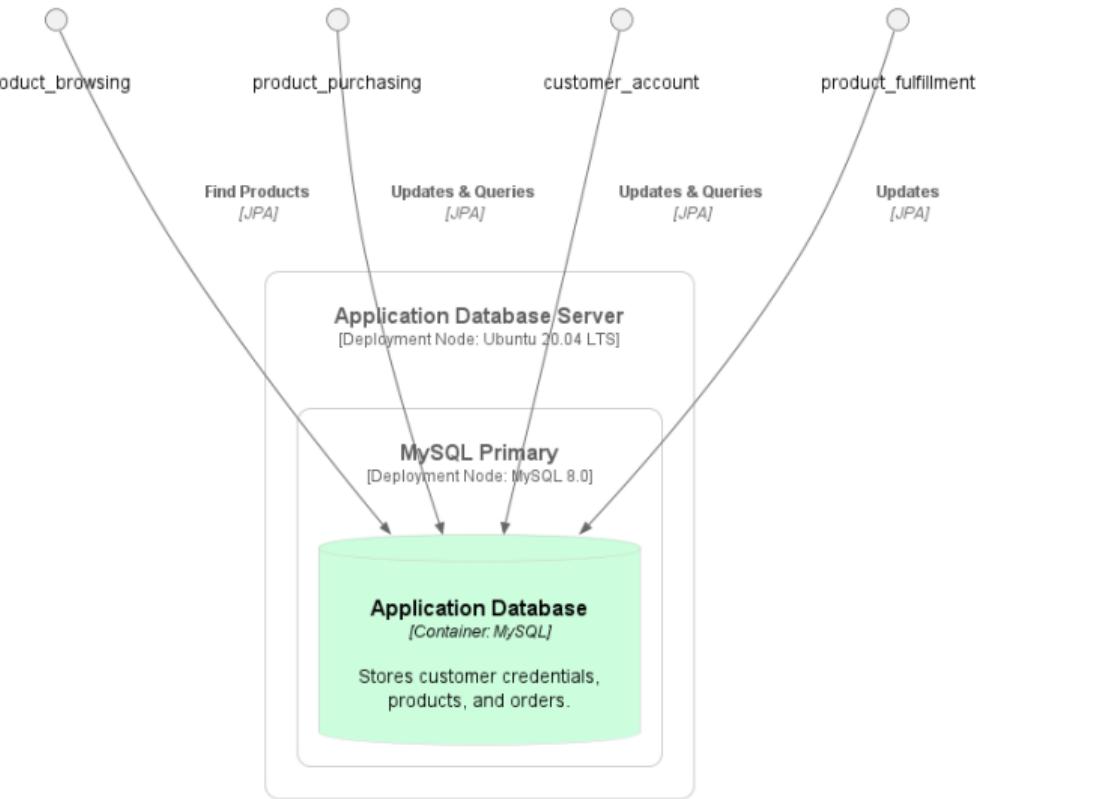
- We scaled a stateless service.
- Stateless: Services don't require persistent data *between* requests.
- Persistent state is saved in the database.
- This is normally easy to do.

Question

What is the *problem*?

The database

Database



- Database has state, persistent data.
- This is much harder to scale.

Stateless vs. Stateful Systems

Stateless Does *not* utilise *persistent data*. Or:
each request is independent.

Stateful Does utilise *persistent data*. Or: the
server or service remembers and uses
data from previous interactions.

Disclaimer

This is *not* a database course

Advanced Database Systems (INFS3200)

Course level

Undergraduate

Faculty

Engineering, Architecture & Information Technology

School

Info Tech & Elec Engineering

Units

2

Duration

One Semester

Class contact

2 Lecture hours, 1 Tutorial hour, 1 Practical or Laboratory hour

Incompatible

INF57907

Prerequisite

INFS2200

Assessment methods**Current course offerings****Course offerings** **Location** **Mode** **Course Profile**Semester 1, 2022 St Lucia Internal [COURSE PROFILE](#)Semester 1, 2022 External External [COURSE PROFILE](#)

Semester 2, 2022 External External PROFILE UNAVAILABLE

Semester 2, 2022 St Lucia Internal PROFILE UNAVAILABLE

Please Note: Course profiles marked as not available may still be in development.

Course description

Distributed database design, query and transaction processing, data integration, data warehousing, data cleansing, management of spatial data, and data from large scale distributed devices.

Archived offerings**Course offerings** **Location** **Mode** **Course Profile**Semester 1, 2021 St Lucia Flexible Delivery [COURSE PROFILE](#)Semester 1, 2021 External External [COURSE PROFILE](#)Semester 2, 2021 External External [COURSE PROFILE](#)Semester 2, 2021 St Lucia Internal [COURSE PROFILE](#)Semester 1, 2020 St Lucia Internal [COURSE PROFILE](#)

This is a database course.

Question

How do we fix database scaling issues?

Question

How do we fix database scaling issues?

Answer

- Replication

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- Replication
- Partitioning

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- Replication
- Partitioning
- Independent databases

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Answer

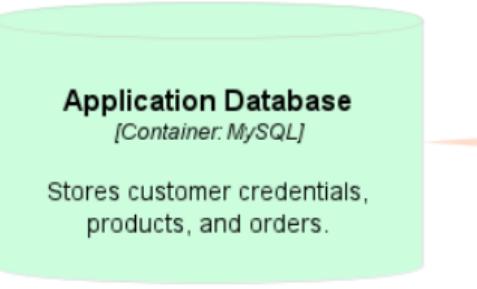
- *Replication*
- Partitioning
- Independent databases

Question

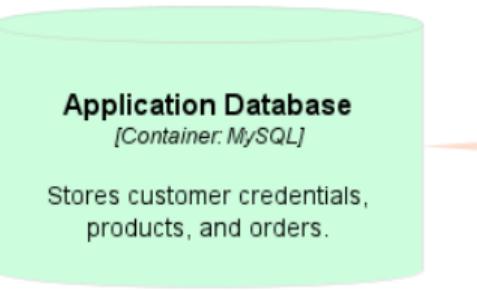
What is *replication*?

Definition 0. Replication

Data copied across multiple different machines.



product_id	name	stock	price
1234	Nicholas Cage Reversible Pillow	10	\$10.00
4321	Lifelike Elephant Inflatable	5	\$50.00



product_id	name	stock	price
1234	Nicholas Cage Reversible Pillow	10	\$10.00
4321	Lifelike Elephant Inflatable	5	\$50.00

Definition 0. Replica

Database node which stores a copy of the data.

Question

What are the advantages of *replication*?

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Answer

- *Scale* our database to cope with higher loads.

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- Provide *fault tolerance* from a single instance failure.

Question

What are the advantages of *replication*?

Answer

- *Scale* our database to cope with higher loads.
- Provide *fault tolerance* from a single instance failure.
- Locate instances *closer to end-users*.
- Scalability
- Reliability
- Performance

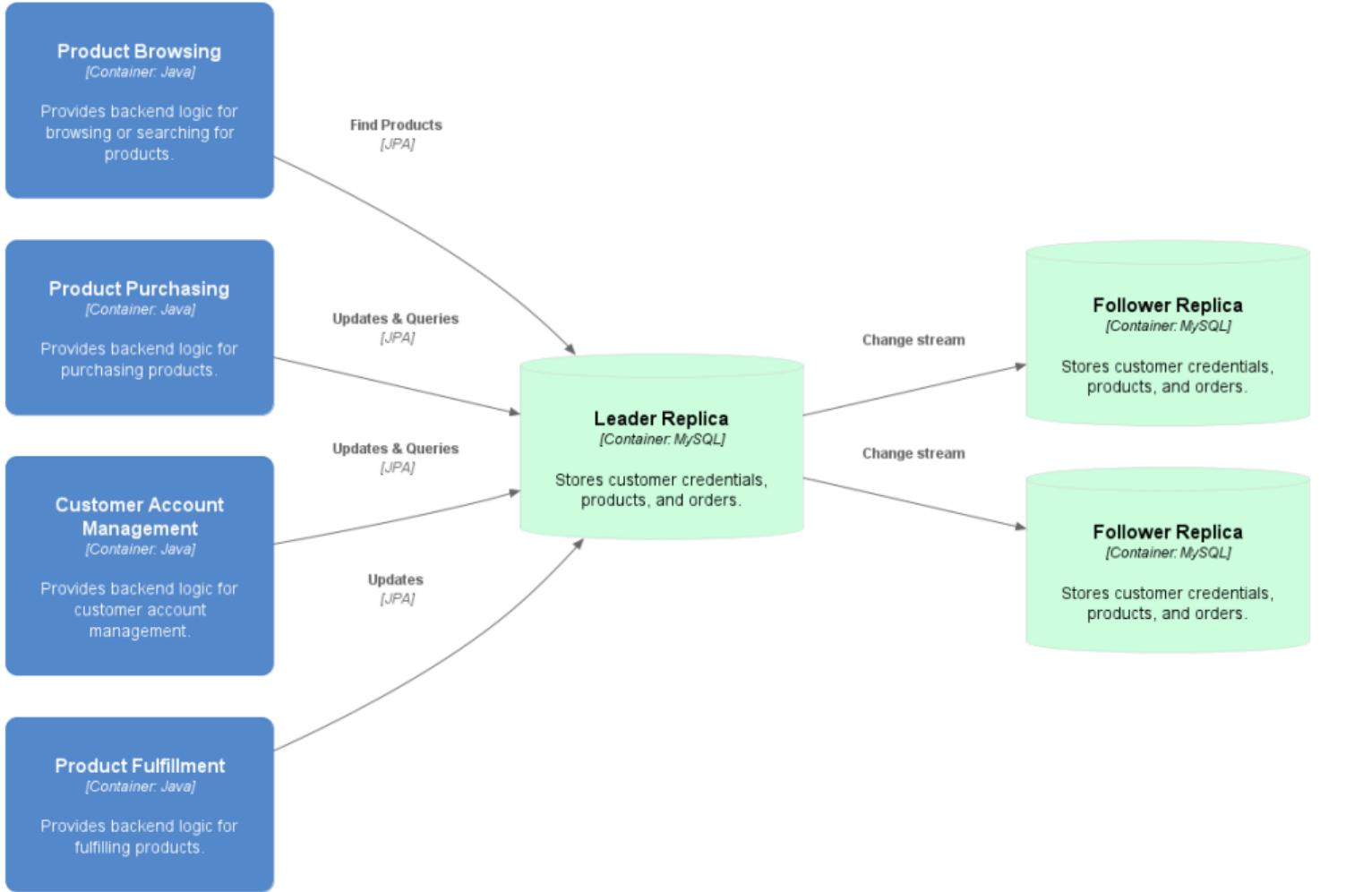
Question

How do we replicate our data?

- Easy without updates, just copy it.
- Updates, or writes, must *propagate* changes.

First Approach

Leader-Follower Replication



- Leader-Follower is the most common implementation.
- Multiple followers, only *one* leader.

Definition 0. Leader-based Replication

one node (the leader) handles all write operations, and multiple other nodes (followers) replicate the data and handle read operations.

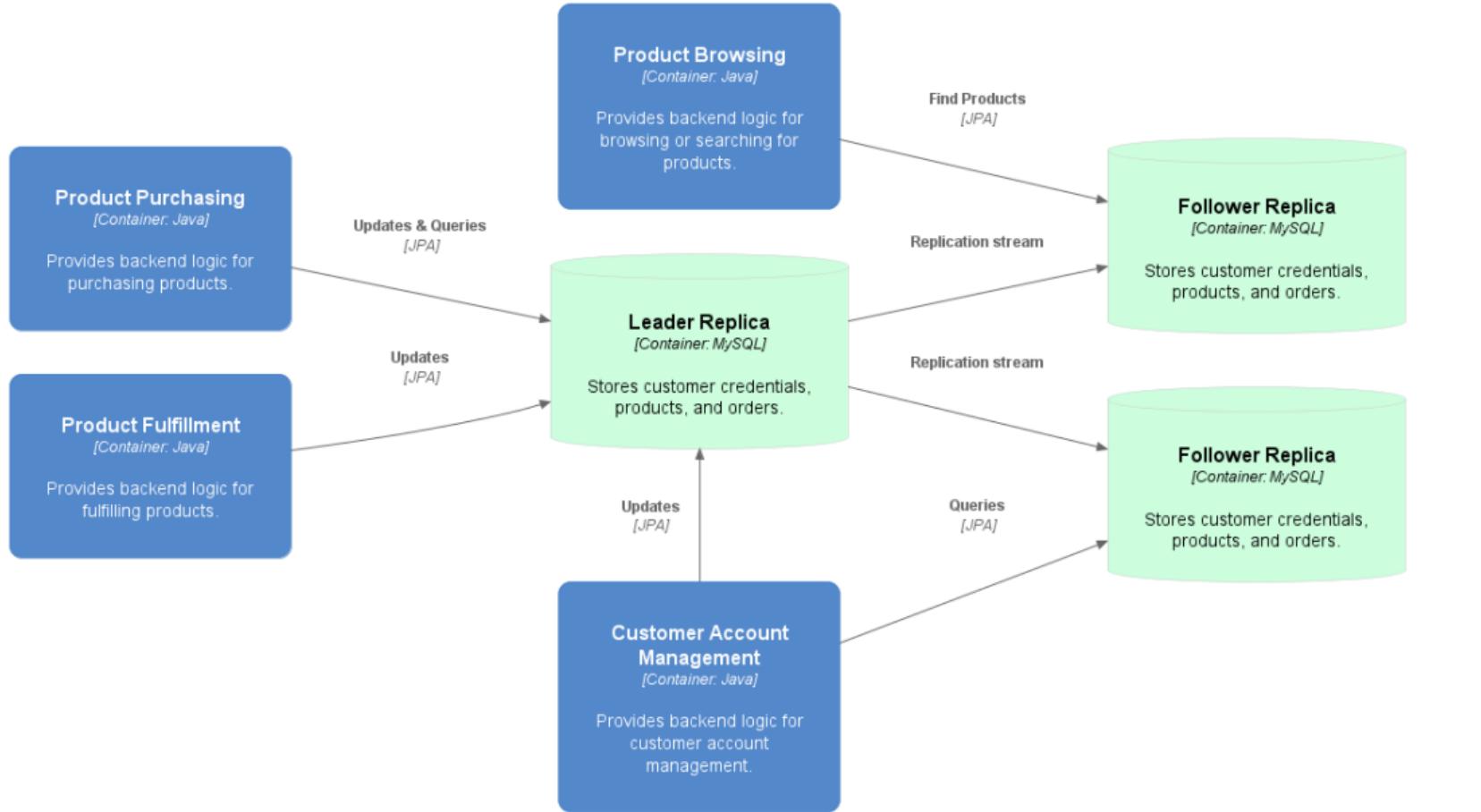
Leader-based Replication

On write Writes sent to *leader*, change is propagated via change stream.

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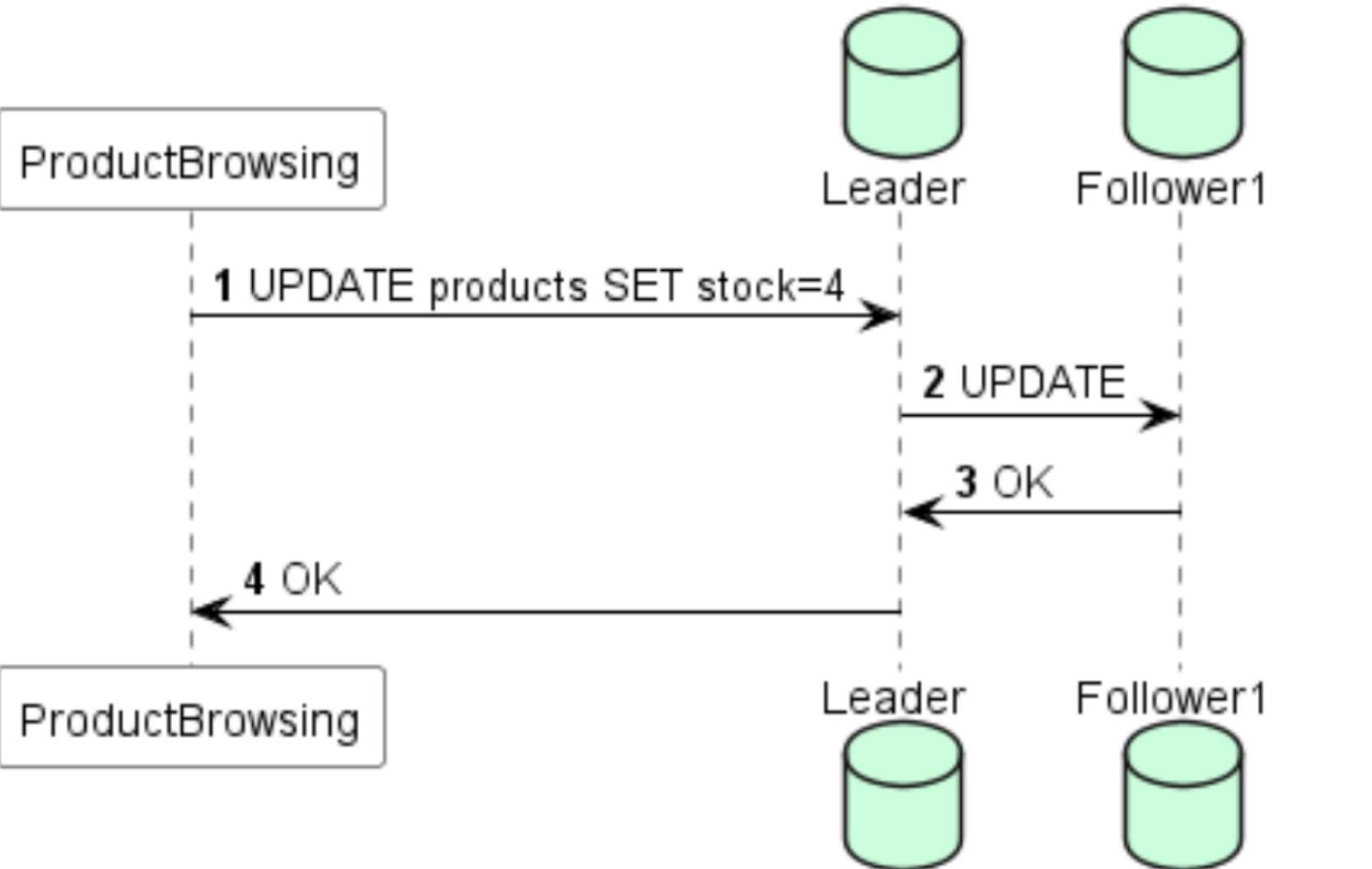
On read Any *replica* can be queried.



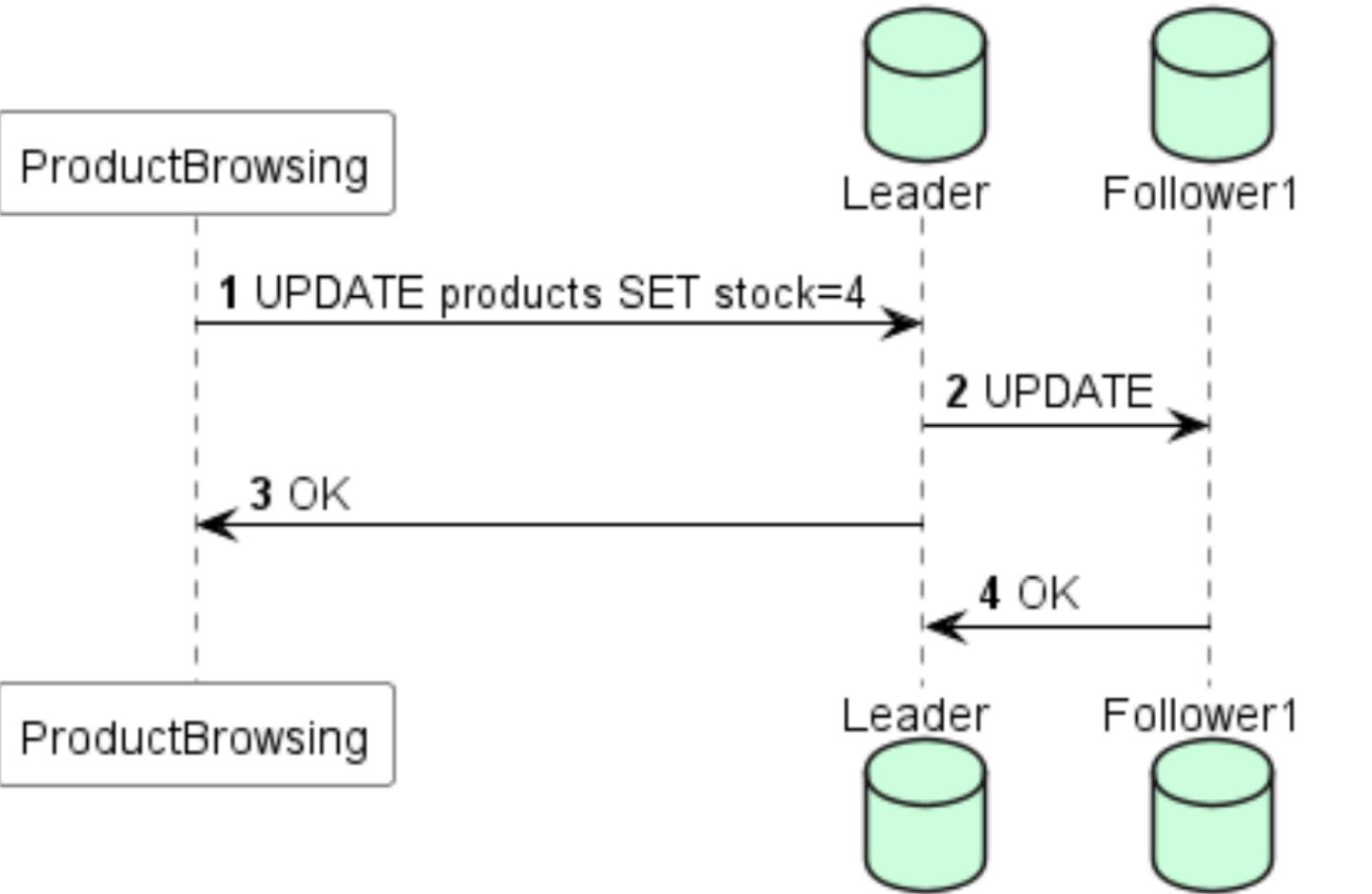
- Built-in to PostgreSQL, MySQL, MongoDB, RethinkDB, and Espresso.
- Can be added to Oracle and SQL Server.

Propagating Changes

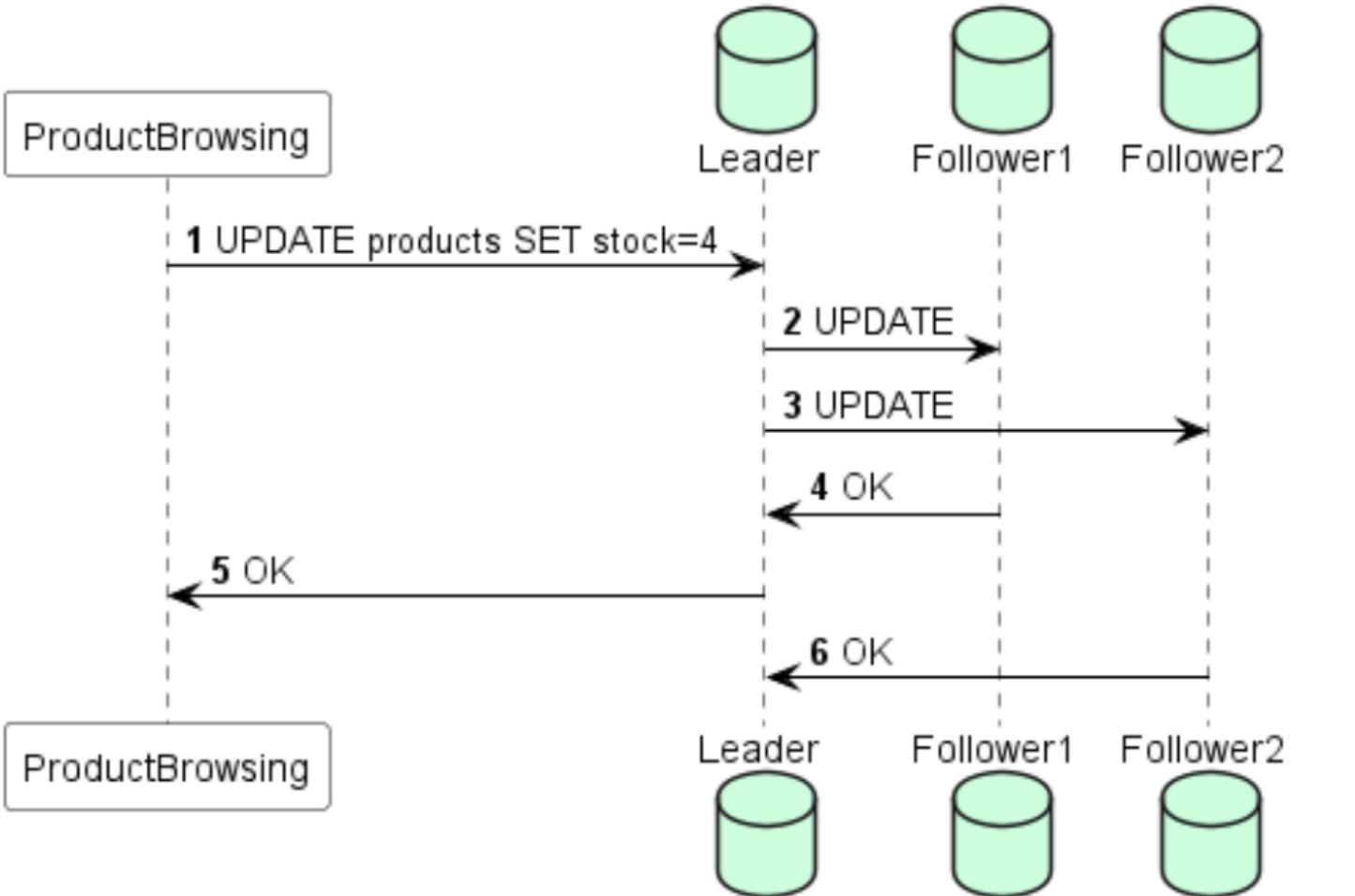
Synchronous vs. Asynchronous



Synchronous update.



Asynchronous update.



- What could go wrong here?
- *Follower1* can get out of sync with *Follower2*.
- Following material deals with *leader* or a *replica* going down.

Synchronous Propagation

- Writes must propagate to *all followers* before being successful.

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- *Any* replica goes down, *all* replicas are un-writeable.

Synchronous Propagation

- Writes must propagate to *all followers* before being successful.
- *Any* replica goes down, *all* replicas are un-writeable.
- Writes must *wait* for propagation to *all* replicas.

Asynchronous Propagation

- Writes *don't* have to *wait* for propagation.

Asynchronous Propagation

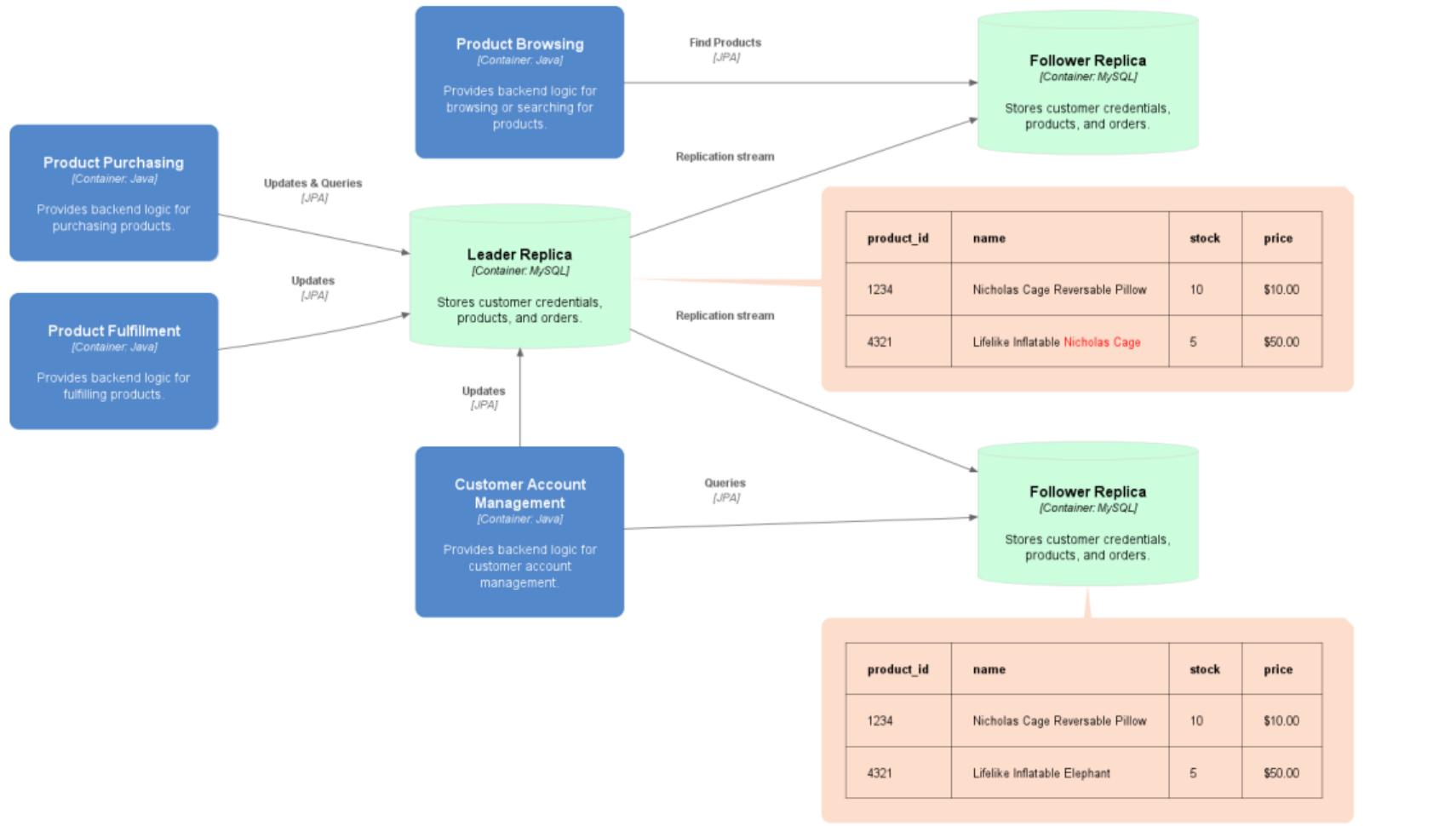
- Writes *don't* have to *wait* for propagation.
- If the leader goes down before propagating, the *write is lost*.

Asynchronous Propagation

- Writes *don't* have to *wait* for propagation.
- If the leader goes down before propagating, the *write is lost*.
- Replicas can have out-dated or *stale* data.

Definition 0. Stale data

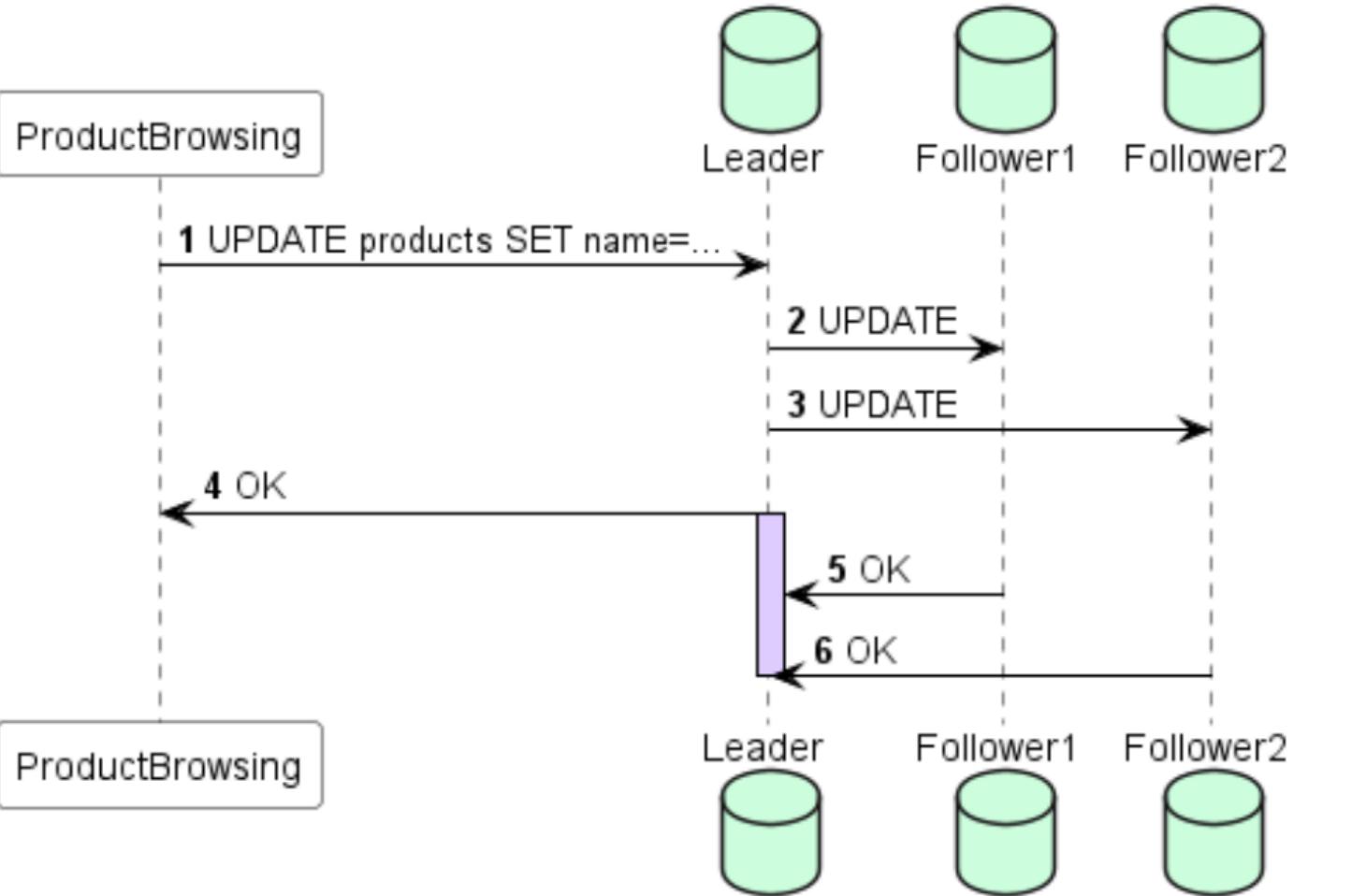
Outdated or inconsistent data that does not reflect the latest updates, mainly due to replication delays, caching, or network issues in distributed systems.



Definition 0. Replication Lag

The time taken for replicas to update *stale* data.

Replication Lag: Time it takes for the product name change to update across *all* followers.



The purple lifeline bar is *replication lag*.

Eventually, all replicas must become consistent

The system is *eventually consistent*.

- If writes stop for long enough.
- Eventually is intentionally *ambiguous*.

Eventual Consistency

Sufficient? Problems?



Brae Webb
@braewebb

1. Read user details.



Brae Webb
@braewebb

Name:	<input type="text" value="Brae"/>
<input type="button" value="Cancel"/>	<input type="button" value="Save"/>

1. Read user details.
2. Decide I don't like my name.
3. Update name.



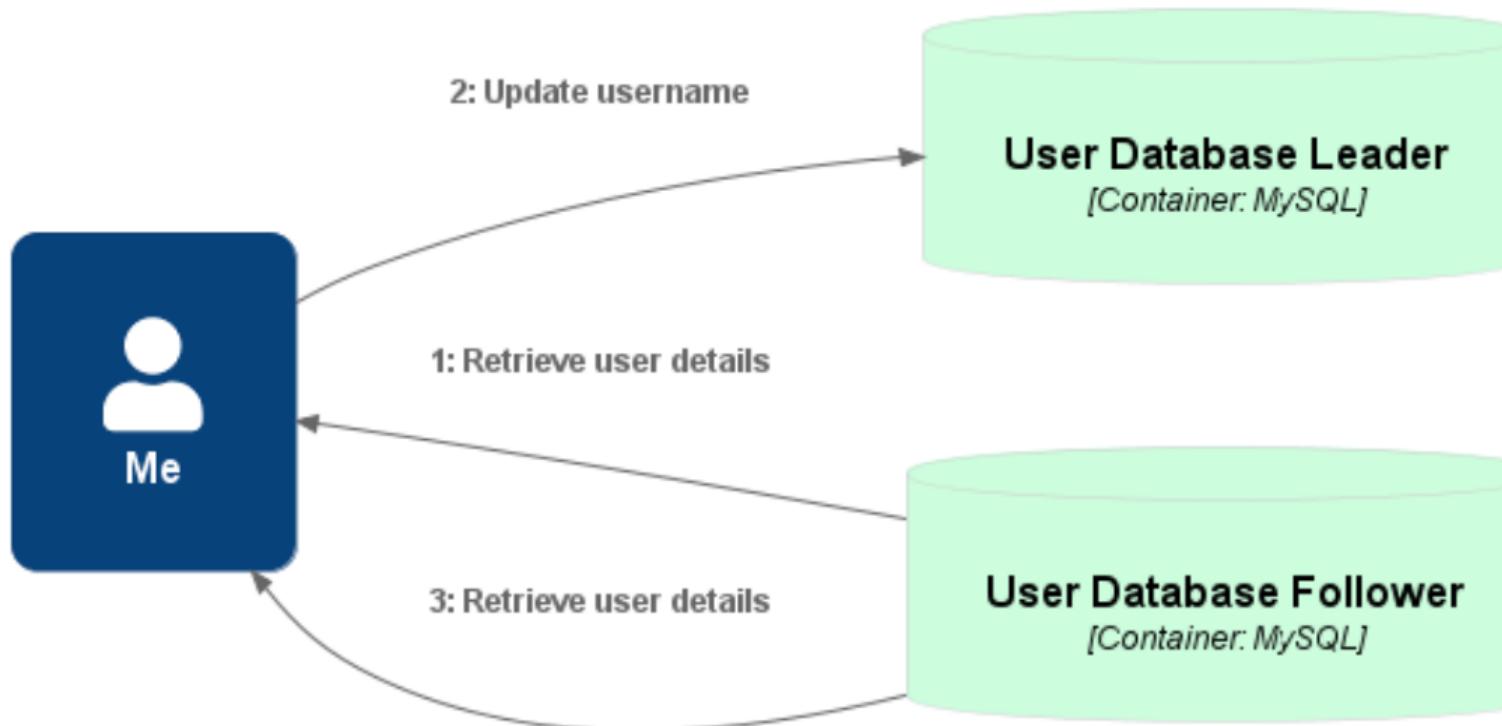
Brae Webb
@braewebb

Name:

A screenshot of a user profile edit dialog. It shows the current name 'Brae' in a text input field. Below the input field are two buttons: 'Cancel' on the left and 'Save' on the right.

Brae Webb
@braewebb

1. Read user details.
2. Decide I don't like my name.
3. Update name.
4. Read user details.



- Typical interaction in simple Leader-Follower replication.
- Write is to *leader*.
- Read is from *follower*.
- *Replication lag* means reading a field immediately after updating it *may* lead to reading *stale* data.

Definition 0. Read-your-writes Consistency

Users always see the updates that *they have made* (even though others see stale data).

Doesn't care what other users see.



Brae Webb
@braewebb

My fist post

1. Misspell a tweet.



Brae Webb
@braewebb

My fist post



Brae Webb
@braewebb

My first post

1. Misspell a tweet.
2. Correct spelling, and I see my *updated* tweet.



Brae Webb
@braewebb

My fist post



Brae Webb
@braewebb

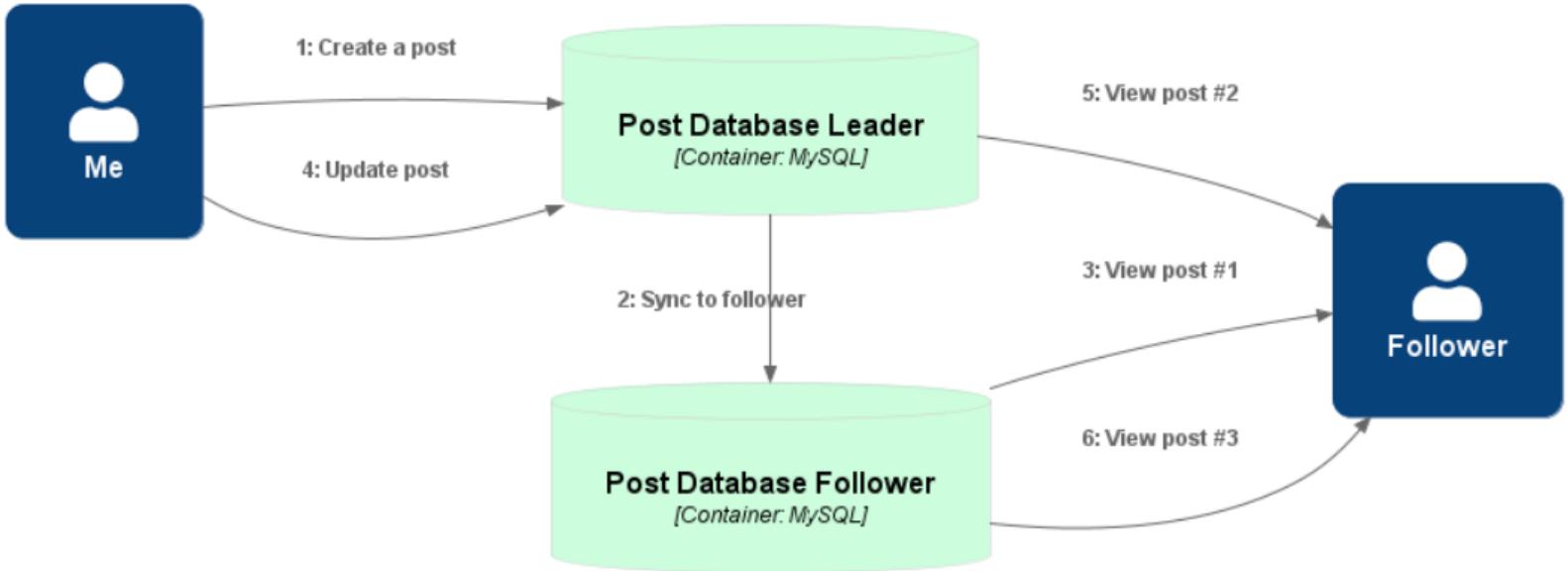
My first post



Brae Webb
@braewebb

My fist post

1. Misspell a tweet.
2. Correct spelling, and I see my *updated* tweet.
3. Other users may still see *stale*, misspelt post.



- Go through each step in sequence.
- Step 6: 3rd view post, gets *old value*.

Definition 0. Monotonic Reads

Once a user reads an updated value, they don't later see the old value.

User doesn't travel back in time.

Definition 0. Causal Consistency

Causally related updates appear in order (e.g., comments appear under the right post).

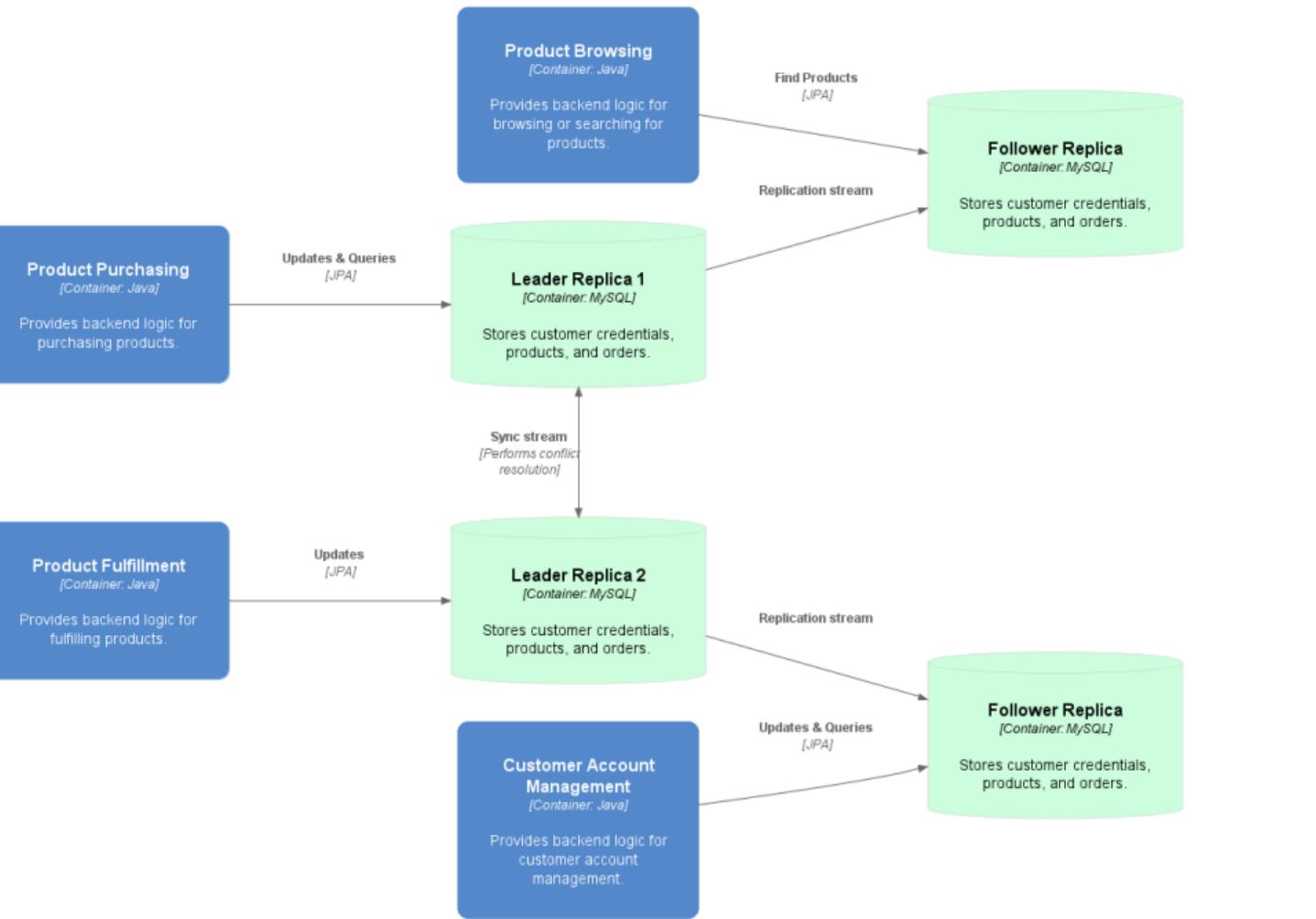
To generalize time to any dependent relations.

Summary

- Leader-follower databases allow *reads to scale* more effectively.
- Asynchronous propagation weakens consistency to *eventually consistent*.
- Leader-follower databases still have a *leader write bottle-neck*.

Second approach

Multi-leader Replication



- Application can be partitioned to perform certain types of writes to a specific leader.
- Reads are from replicas, as with Leader-Follower replication.

Why multi-leader?

- If you have multiple leaders, you can write to any, allowing *writes to scale*.

Why multi-leader?

- If you have multiple leaders, you can write to any, allowing *writes to scale*.
- A leader going down doesn't prevent writes, giving *better fault-tolerance*.
 - Available via extensions in most databases, often not supported natively.
 - Best to avoid where possible.
 - Example: Globally distributed data centres.

Question

What might go wrong?

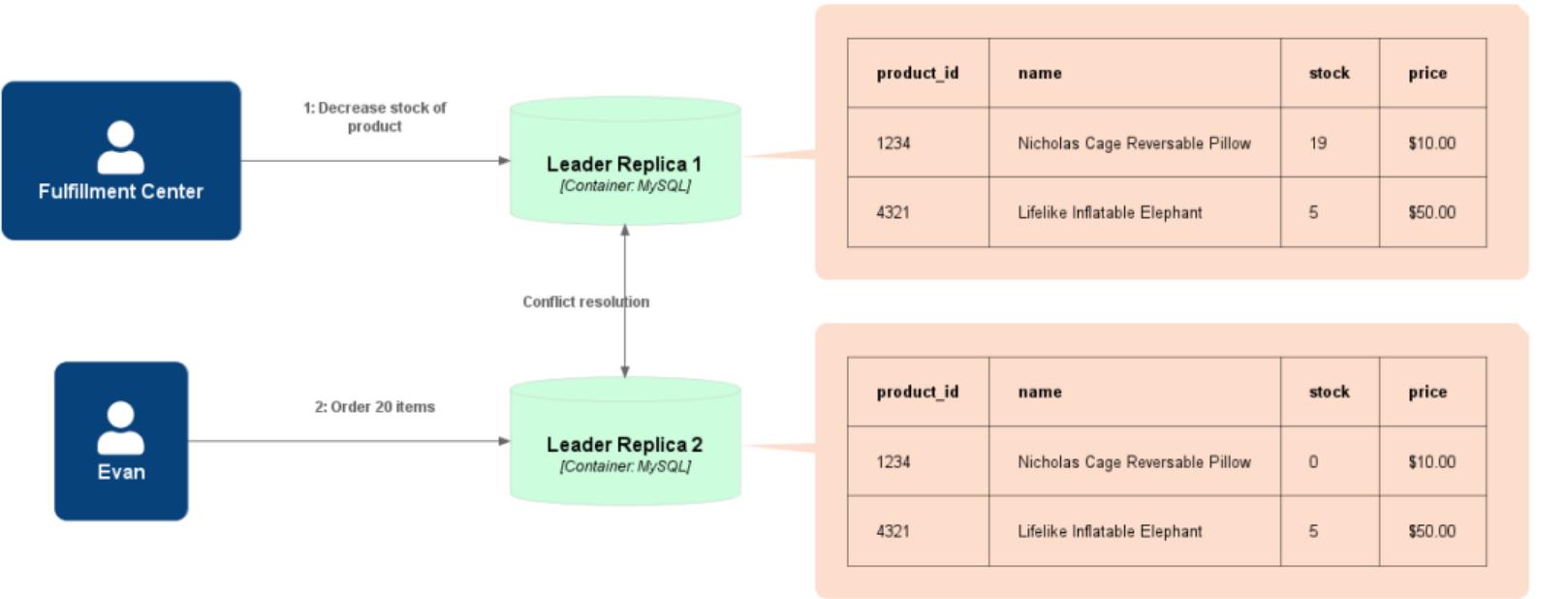
Question

What might go wrong?

Answer

Write conflicts

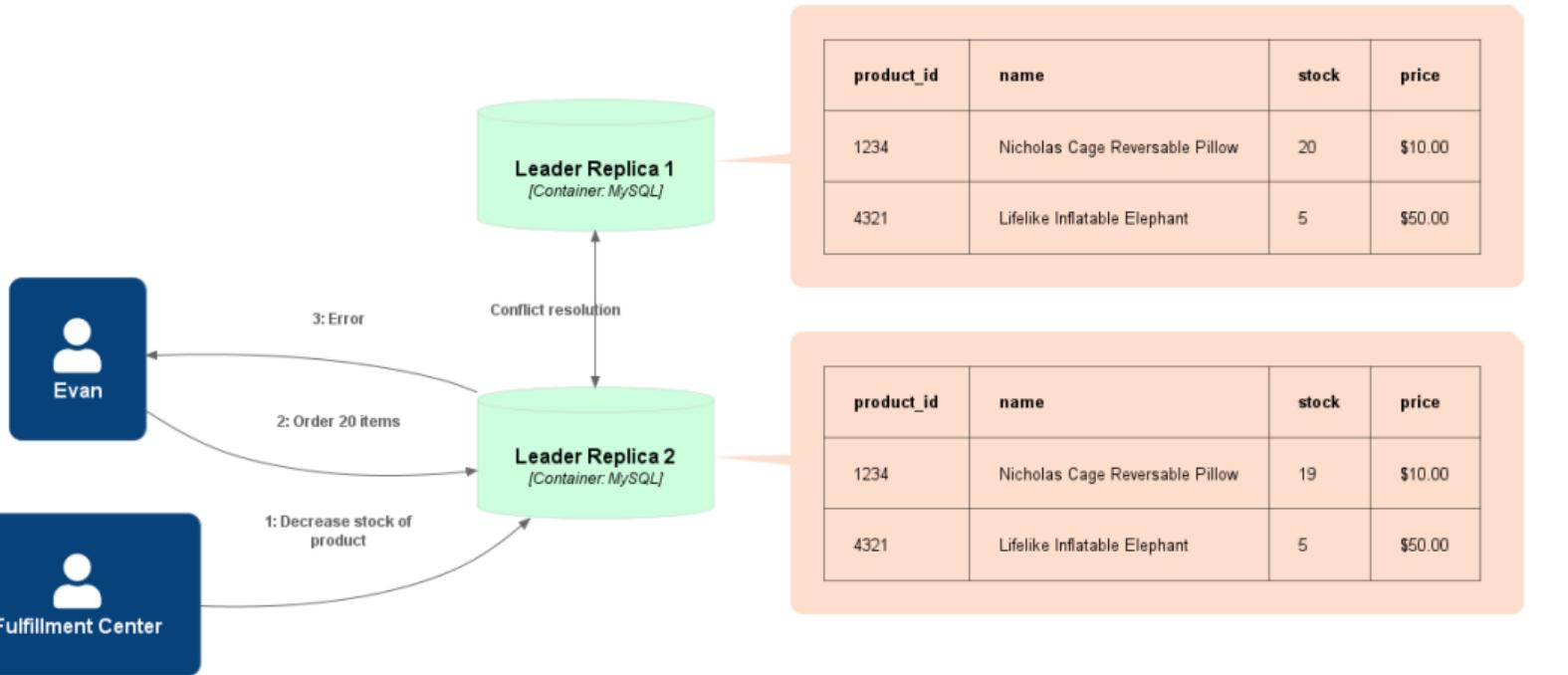
Conflict needs to be *resolved*.



1. Fulfilment centre finds faulty pillow and decreases inventory.
2. Customer orders 20 pillows, what they saw as the number available.
3. -1 Pillows?
4. How do we resolve this?

Where possible

Avoid write conflicts



Requires application to ensure all writes to a field/table/shard are via the *same* leader.

Where impossible
Convergence

Convergence Strategies

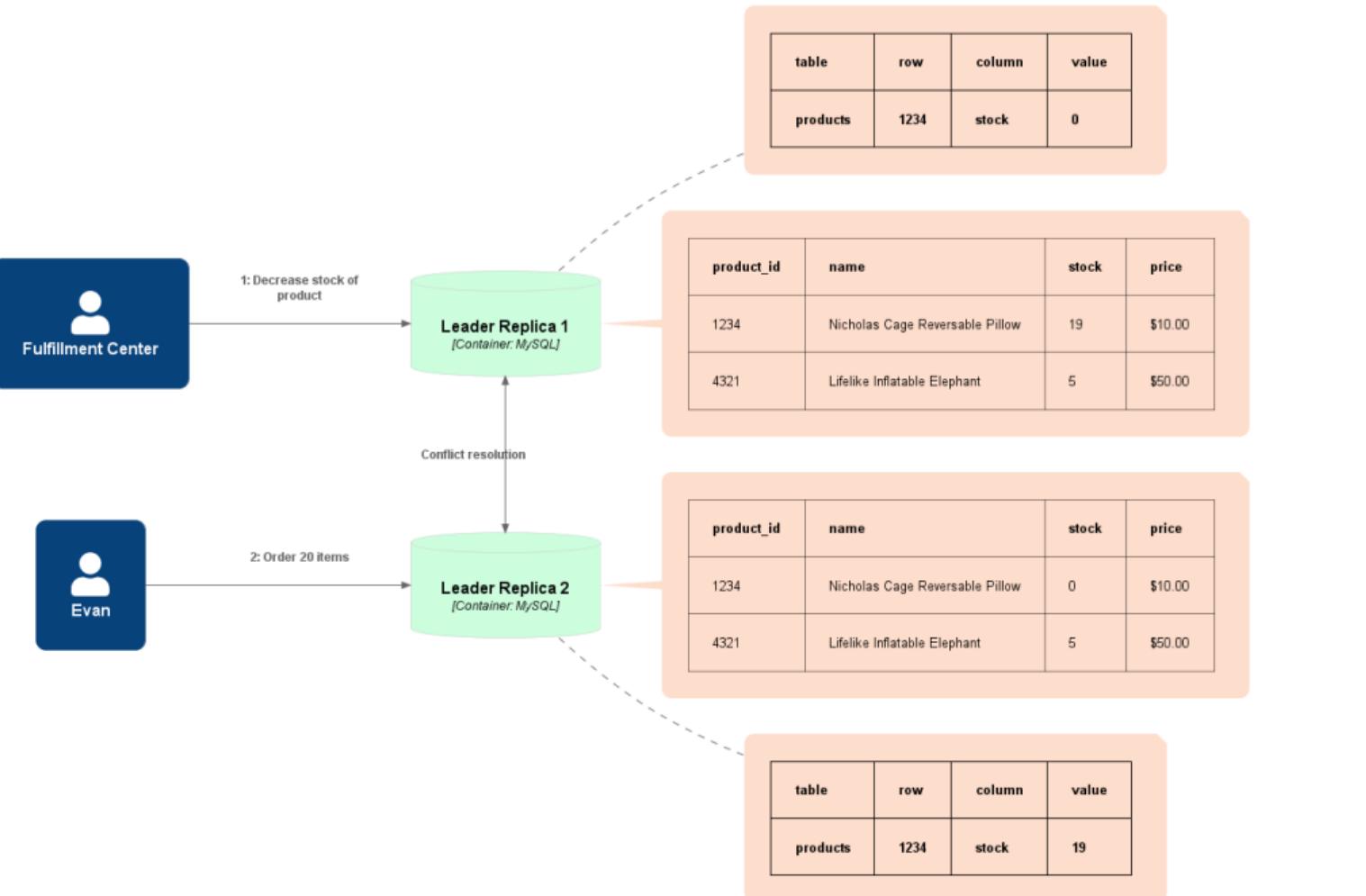
- Assign each *write* a unique ID

Convergence Strategies

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- Assign each *leader replica* a unique ID

Convergence Strategies

- Assign each *write* a unique ID
- Assign each *leader replica* a unique ID
 - Yes, this can be *challenging*.
- Custom resolution logic



Resolving Conflicts

On Write When a conflict is first noticed, take proactive resolution action.

Example: Last Write Wins (LWW) in DynamoDB.

On Read Stores all conflicting versions on write. When a conflict is next read, ask for a resolution.

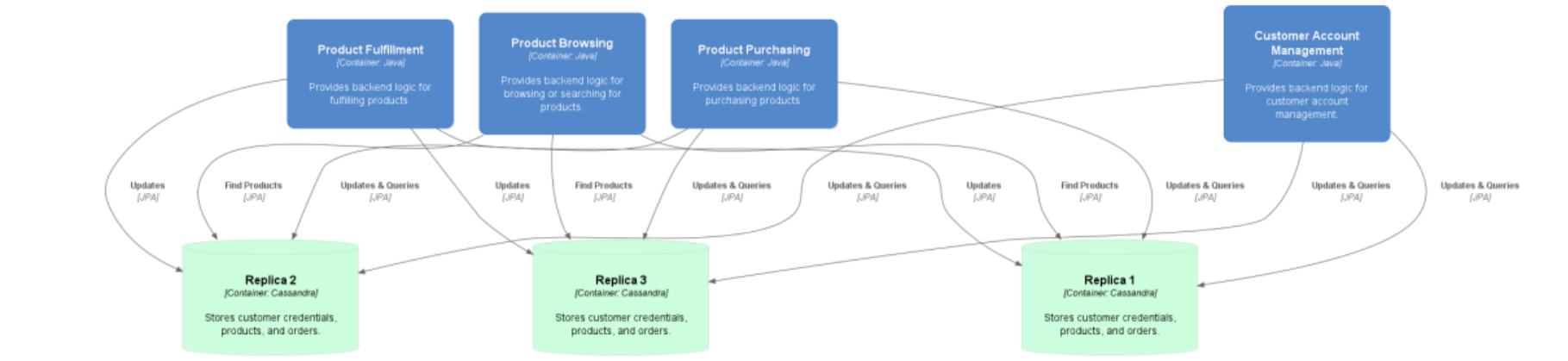
Example: *git pull*

- Bucardo allows a perl script for on write resolution.
- CouchDB prompts reads to resolve the conflict.

Third Approach

Leaderless Replication

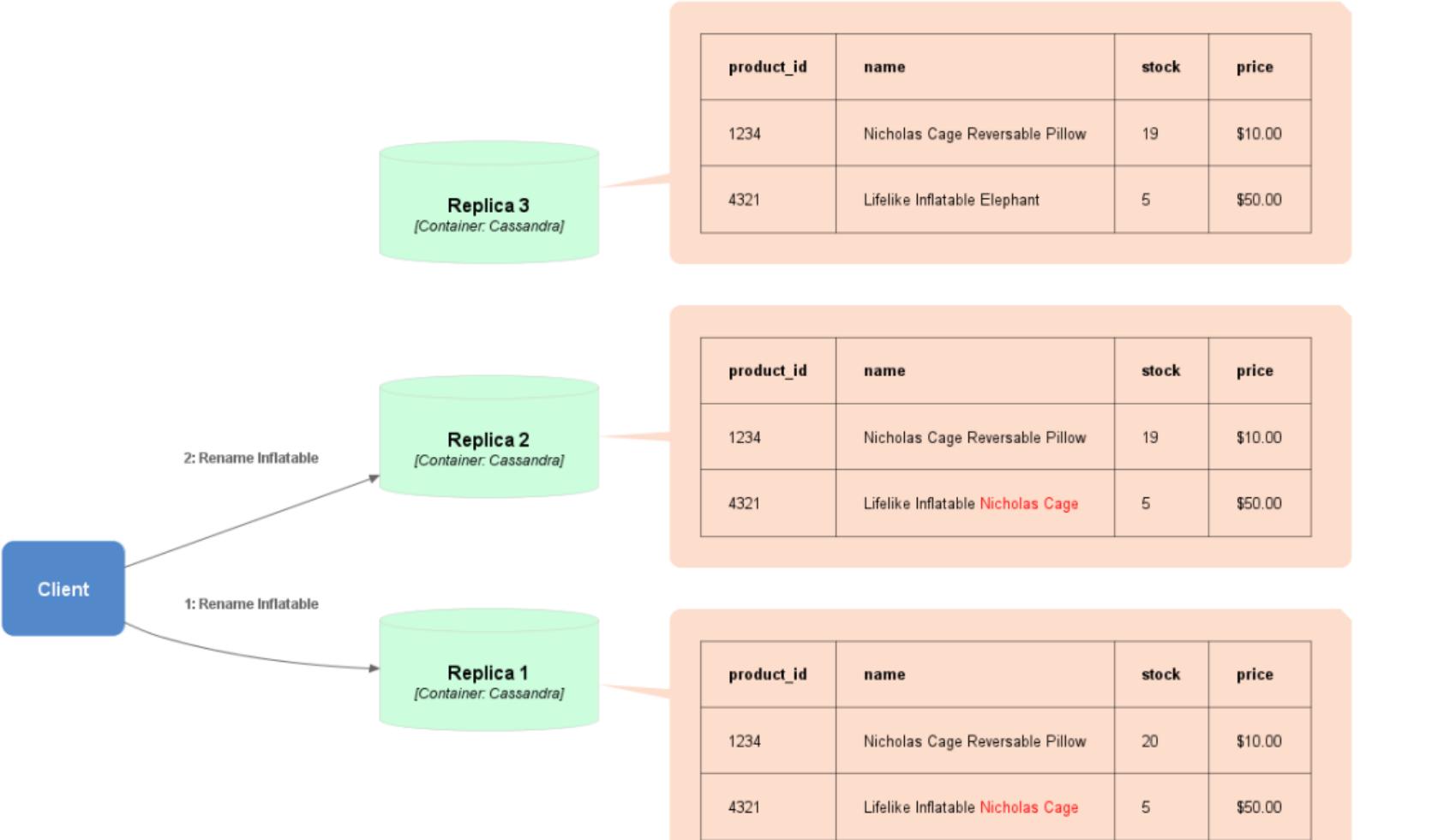
- Early distributed databases were leaderless.
- Resurgence after Amazon created Dynamo.
- Dynamo is an internal service and *not* DynamoDB.
- Riak, Cassandra, and Voldemort are leaderless databases.



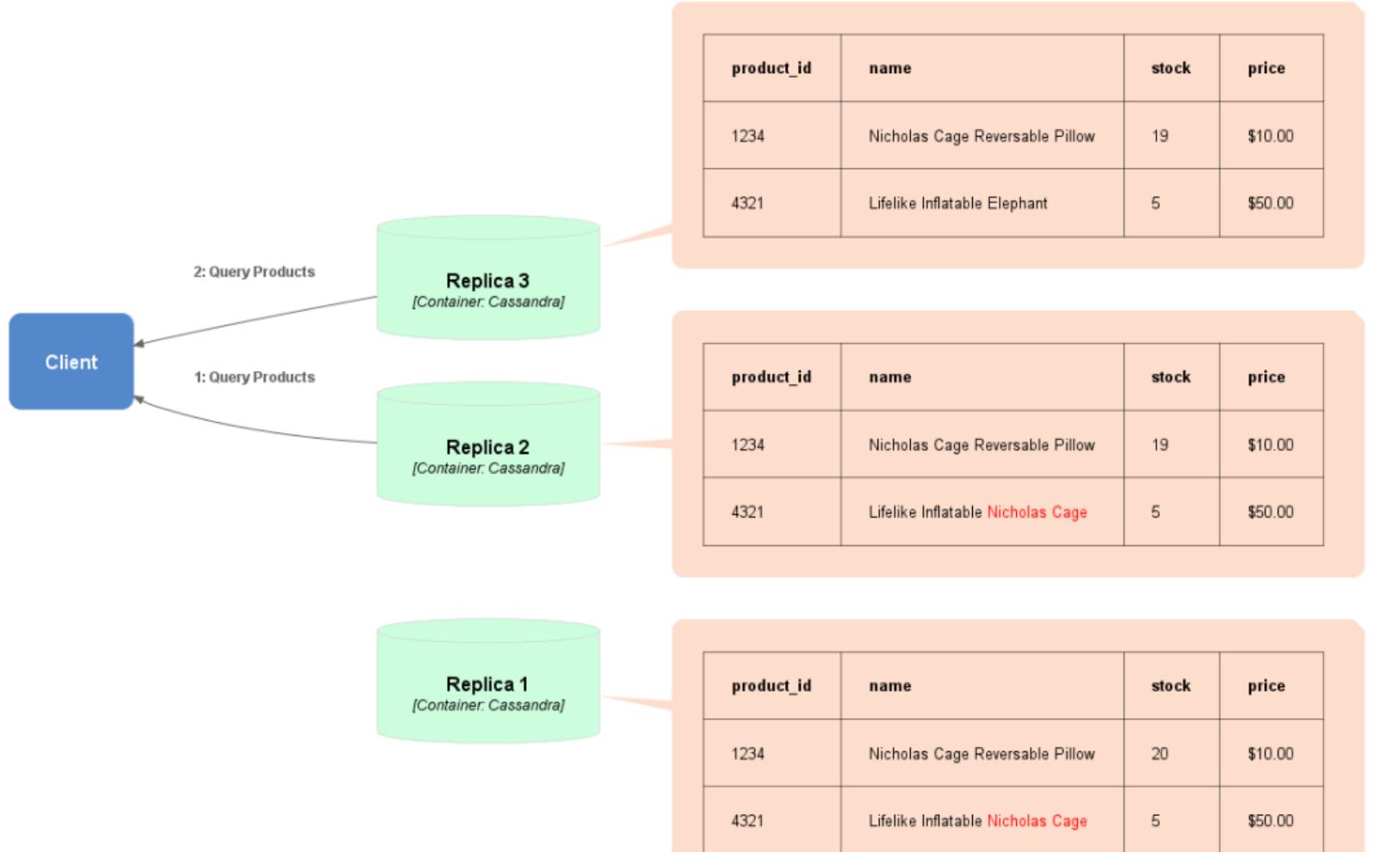
Reads and writes can be written to any node.

How do they work?

Each read/write is sent to *multiple* replicas.



Leaderless Write



Leaderless Read: At least one of the reads has the updated value.

How are changes propagated?

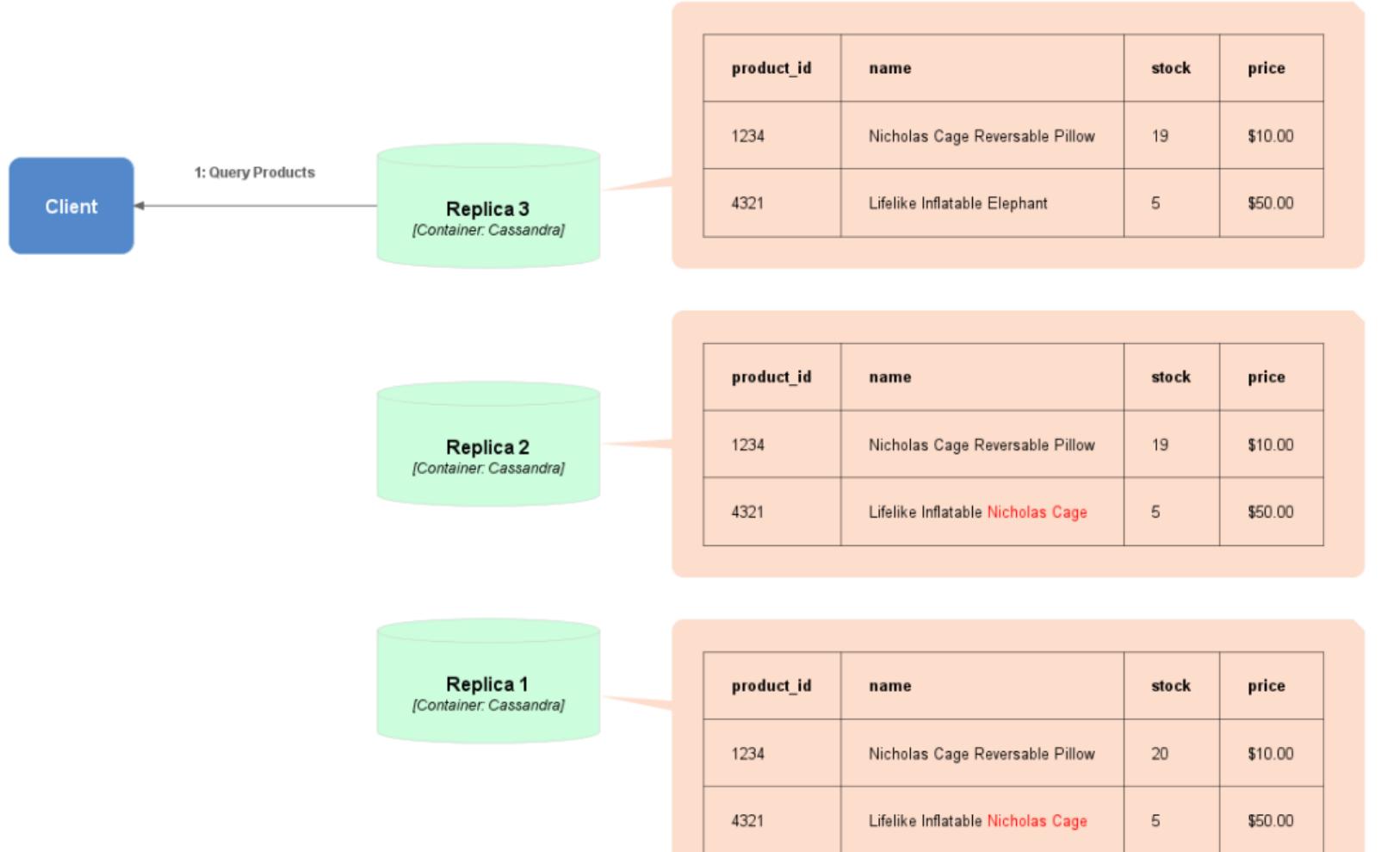
- Read Repair
 1. Read Repair: Client detects stale data on read and writes updated data to that replica.

How are changes propagated?

- Read Repair
 - Anti-Entropy Process
1. Read Repair: Client detects stale data on read and writes updated data to that replica.
 2. Anti-Entropy Process: Background process looks for stale or missing data and updates replicas.

Question

How do we know it's consistent?



Reading from a single replica means we can't know if data is stale or inconsistent.

Question

How do we know it's consistent?

Question

How do we know it's consistent?

Answer

Quorum Reads and Writes

Quorum Consistency

$$w + r > n$$

n total replicas

w amount of replicas to *write* to

r amount of replicas to *read* from

Quorum Consistency

$$2 + 2 > 3$$

n total replicas

w amount of replicas to *write* to

r amount of replicas to *read* from

Quorum Consistency

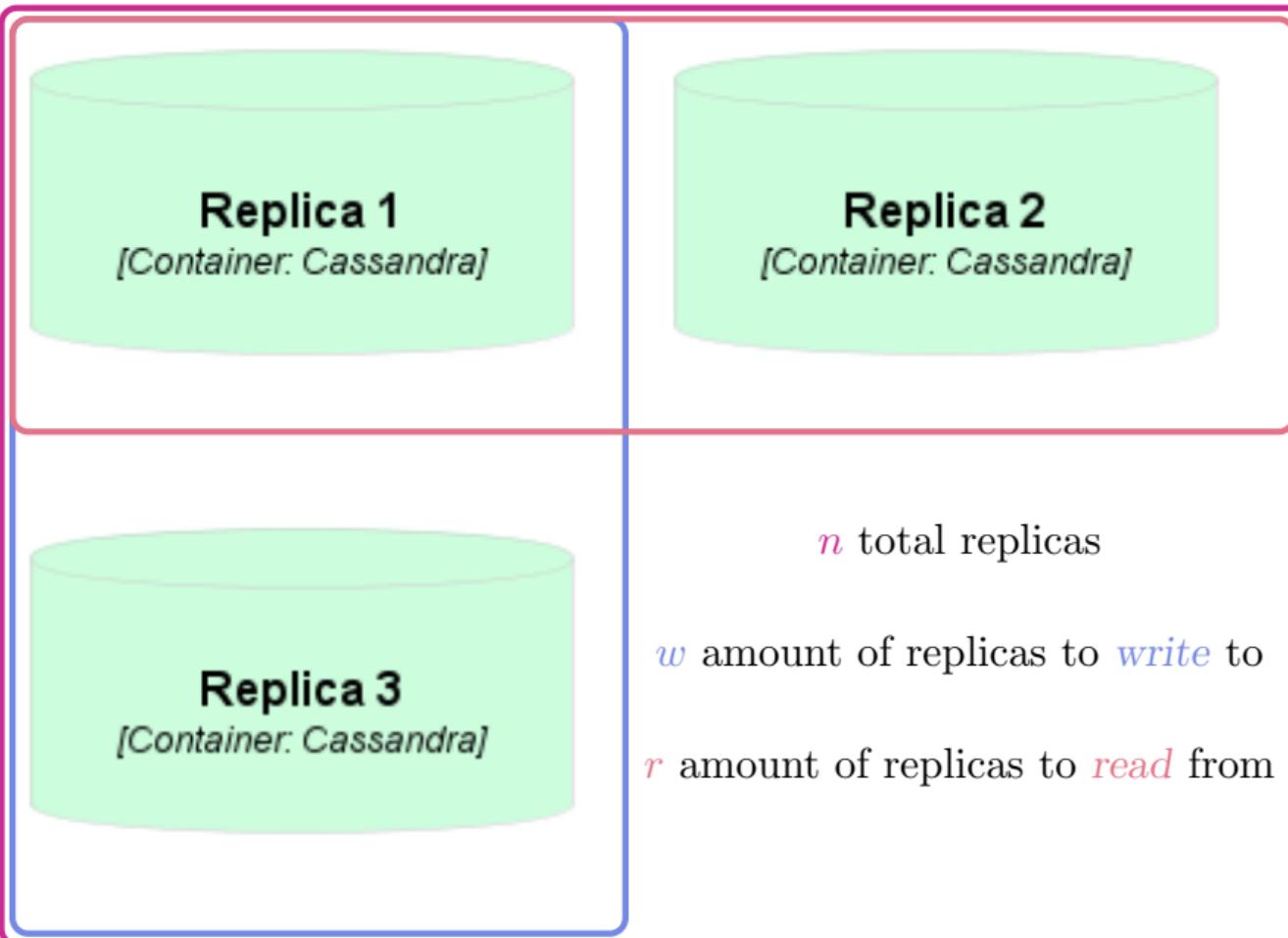
$$1 + 3 > 3$$

n total replicas

w amount of replicas to *write* to

r amount of replicas to *read* from

Nodes read from must overlap with the nodes written to.



- Graphical representation of previous equation.
- **Orange** inner group (*reads*), overlaps with **Blue** inner group (*writes*).
- Showing how reads overlap writes via a quorum.

Question

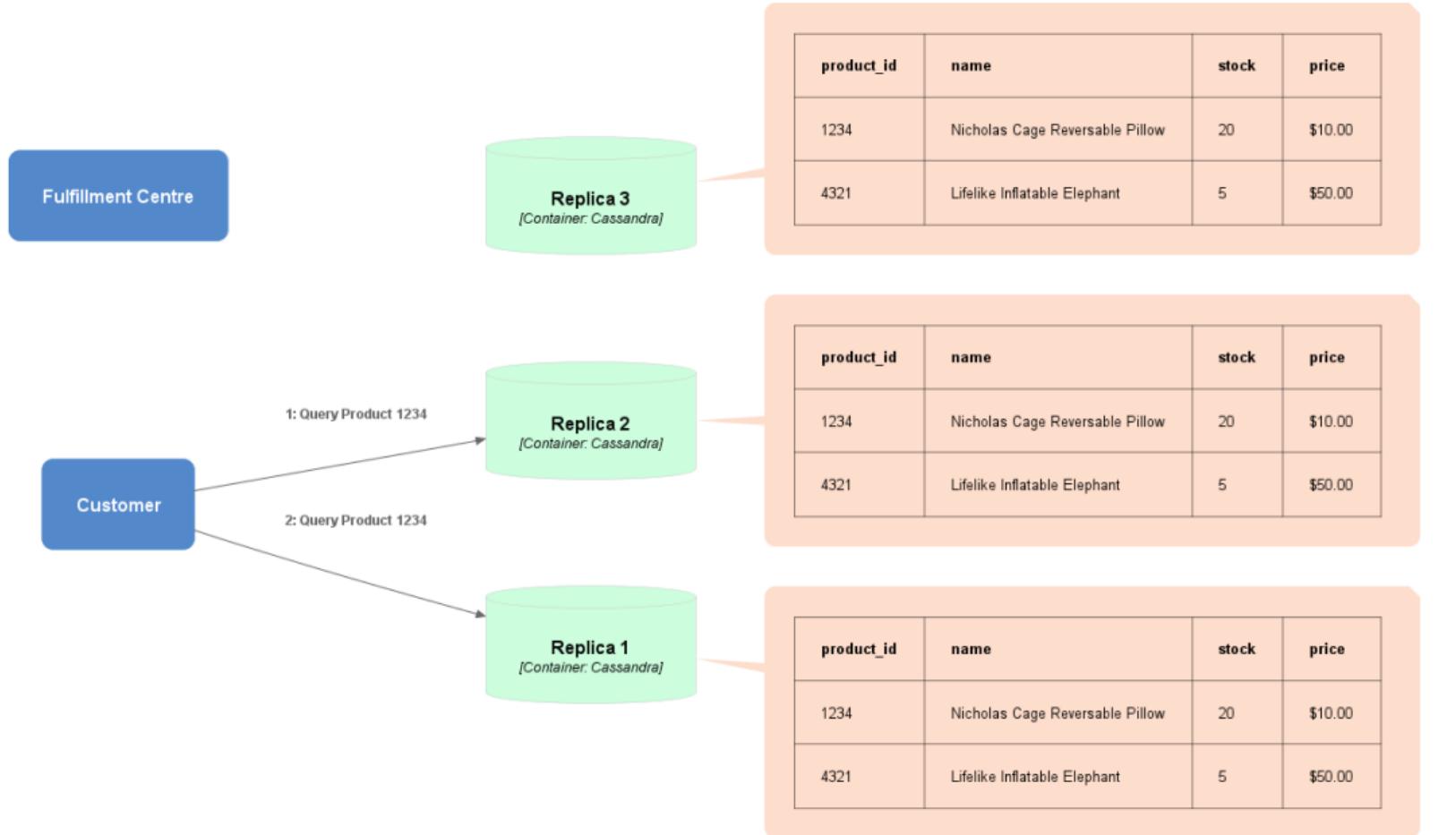
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Question

What about write conflicts?

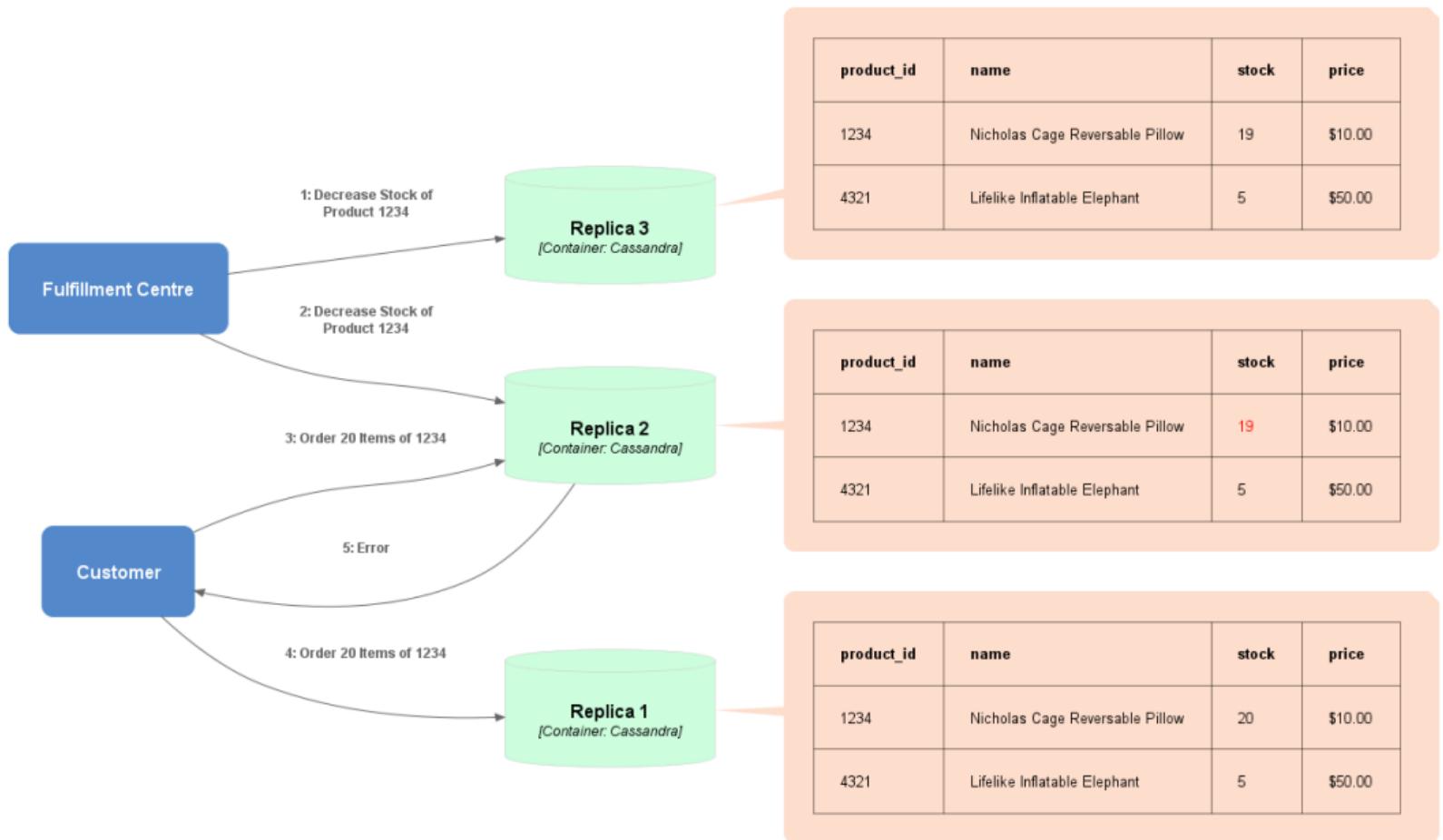
Answer

Same problem as with Multi-leader replication.



Same scenario as before:

1. Customer queries how many pillows are available.
2. Retrieves 20.



Same scenario as before:

1. Fulfilment centre finds faulty pillow and decreases inventory.
2. Customer orders 20 pillows, what they saw as the number available.
3. -1 Pillows?
4. How do we resolve this?

Summary

- *Replication* copies data to multiple replicas.

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- *Replication* copies data to multiple replicas.
- *Leader-based* replication is most common and simplest.
- Replication introduces *eventual consistency*.
- *Multi-leader* replication scales writes as well as reads but introduces *write conflicts*.
- *Leaderless* replication is another approach which keeps the problems of multi-leader.

Question

How do we fix database scaling issues?

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Answer

- *Replication*
- Partitioning
- Independent databases

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Definition 0. Partitioning

Split the data of a system onto multiple nodes.

These nodes are *partitions*.

Also called shards, regions, tablets, etc.



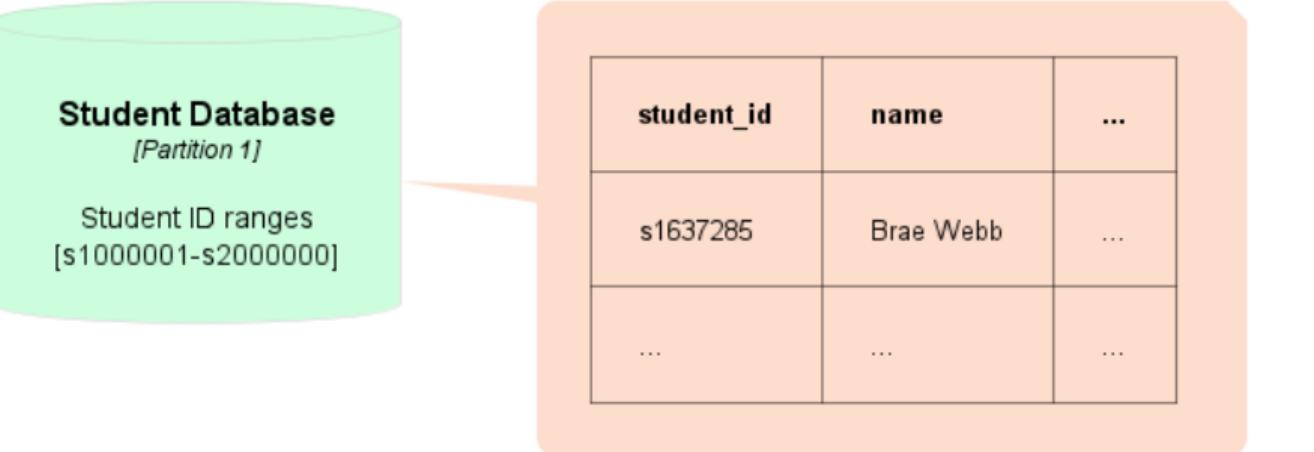
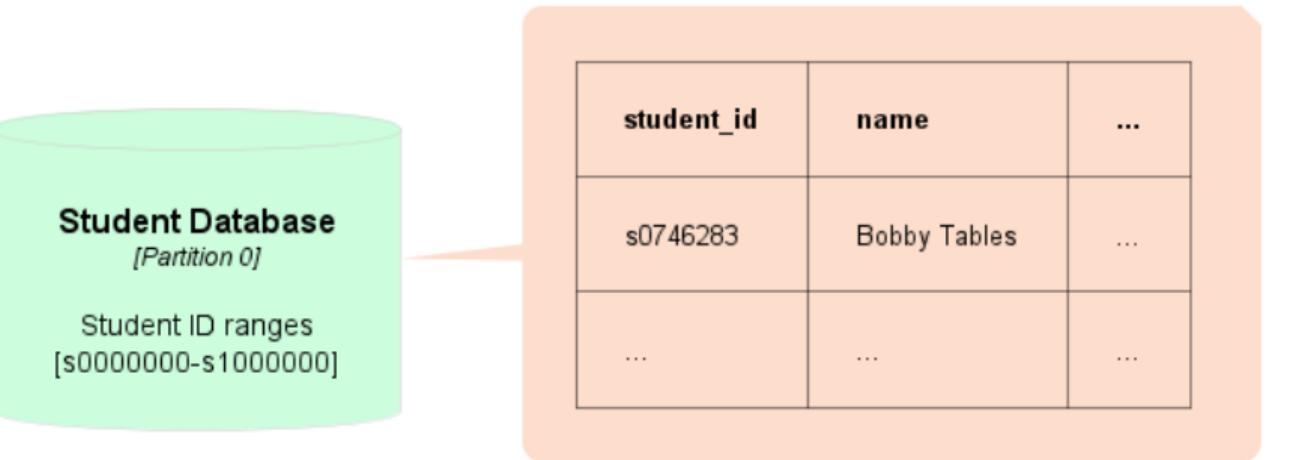
product_id	name	stock	price
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- Pioneered in the 1980s.
- Allow scalability of large data, not just large load.
- Partitioning is normally combined with replication.

Question

How should we decide which data is stored where?



An example partitioning based on primary key, student ID.

Question

What is the problem with this?

Question

What is the problem with this?

Answer

Over time some partitions become inactive,
while others receive almost all load.

Question

How should we decide where data is stored?

Question

How should we decide where data is stored?

Answer

Maximize spread of requests, avoiding *skewing*.

Question

Have we seen this before?

Question

Have we seen this before?

Answer

Hashing?

Hash tables hash entries to maximize the spread between buckets.

Question

What is the problem with this?

Question

What is the problem with this?

Answer

Range queries are inefficient, i.e. get all
students between s4444444 and s4565656.

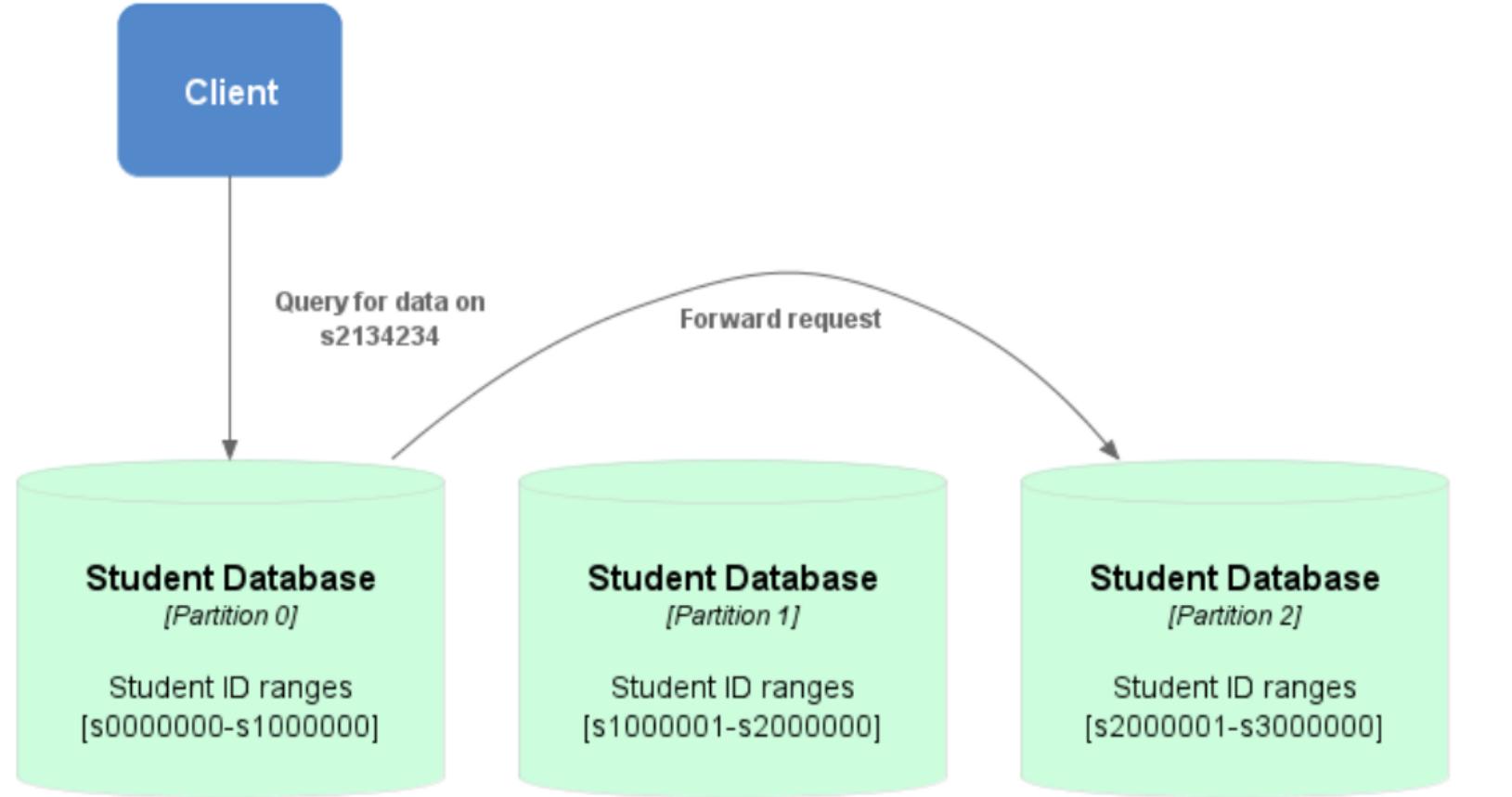
Question

How do we route queries?

Unlike stateless, only one node can process queries.

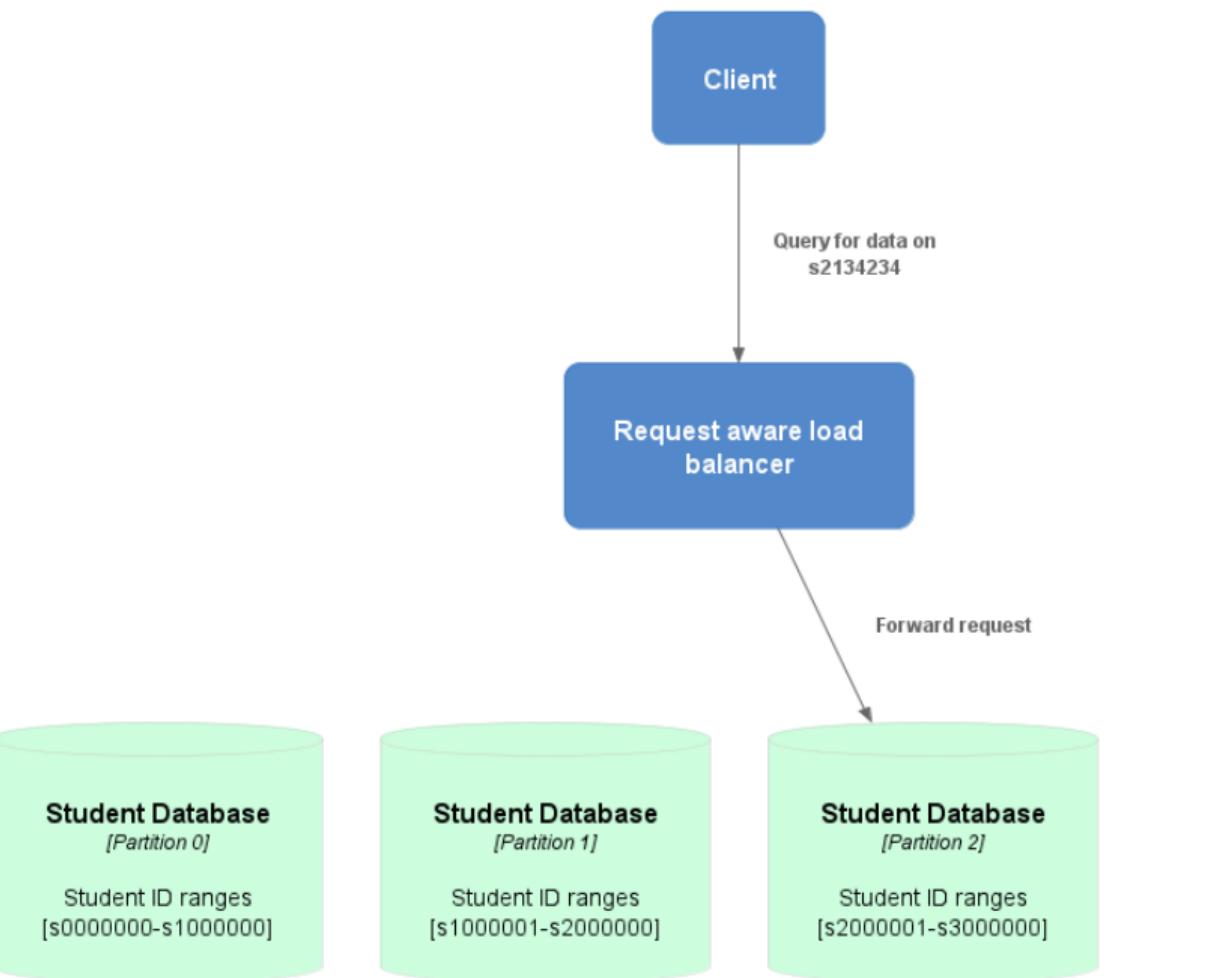
Query-Insensitive Load Balancer

Randomly route to any node, responsibility of the node to re-route to the correct node.



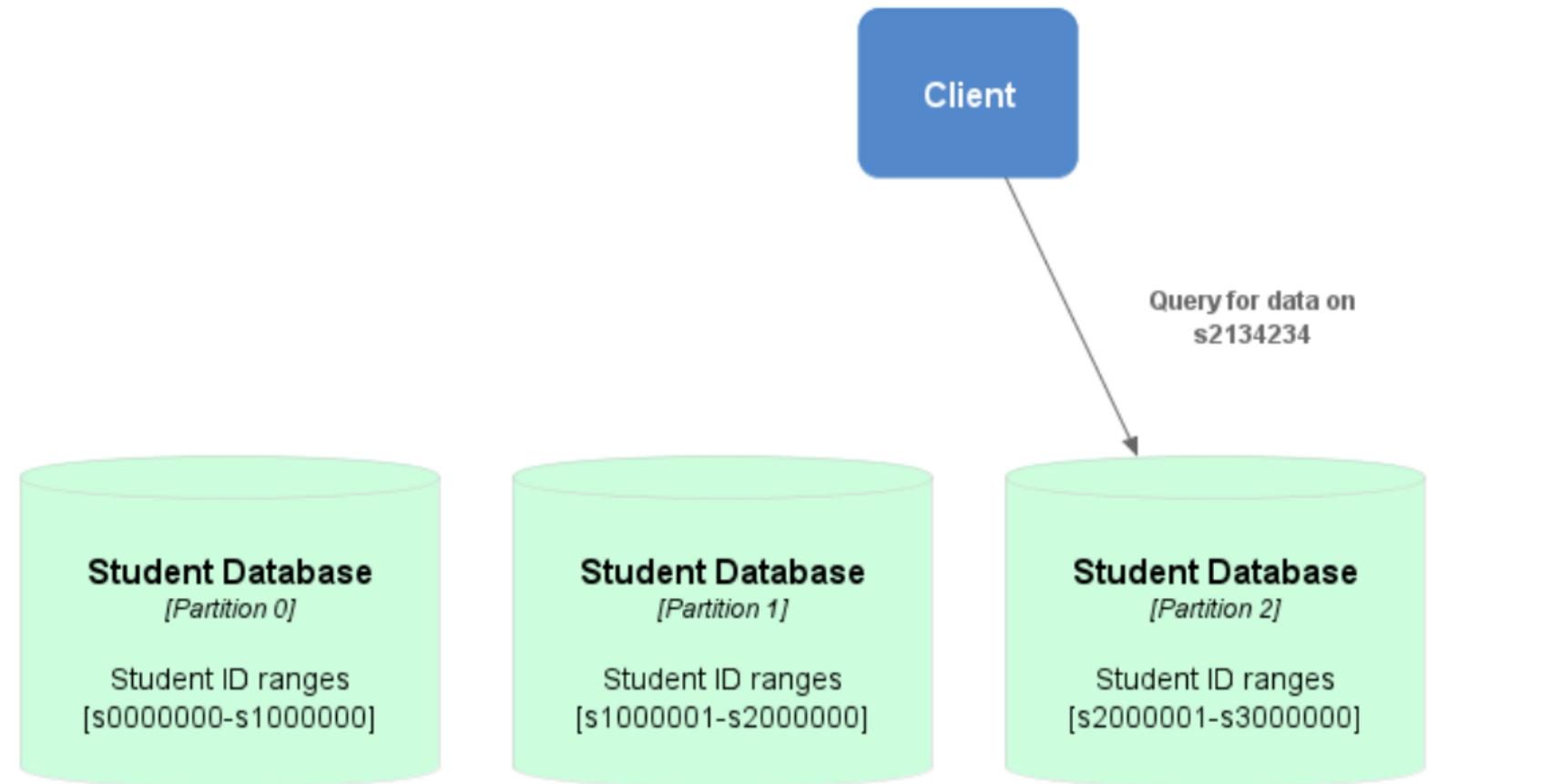
Query-Sensitive Load Balancer

Load balancer understands which queries
should be forwarded to which node.



Client-aware Queries

Place the responsibility on clients to choose the correct node.



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- *Partitioning* splits data across multiple nodes.

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- Requires a *consistent method* to choose appropriate node.

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- *Partitioning* splits data across multiple nodes.
- Requires a *consistent method* to choose appropriate node.
- Partitioning by *primary key* can create *skewing*.
- Partitioning by *hash* makes range queries less efficient.
- Three approaches to *routing requests*.

Disclaimer

We have ignored the *hard* parts of replication.

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 - Leader-based, multi-leader, and leaderless
 - Eventual consistency
 - Write conflicts
- Partitioning
 - Consistent method to pick nodes for data
 - Avoiding skewing