Distributed Systems II

Software Architecture

Brae Webb & Richard Thomas

March 25, 2024

$Distributed\ Systems\ Series$

Distributed I Reliability and scalability of stateless systems.

Distributed II Complexities of stateful systems.

Distributed III *Hard problems* in distributed systems.

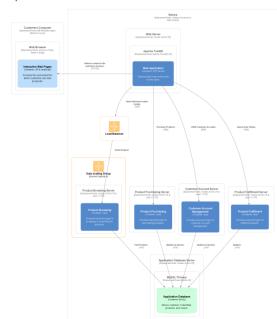
$Distributed\ Systems\ Series$

Distributed I Reliability and scalability of stateless systems.

Distributed II Complexities of stateful systems.

Distributed III Hard problems in distributed systems.

Previously in CSSE6400...

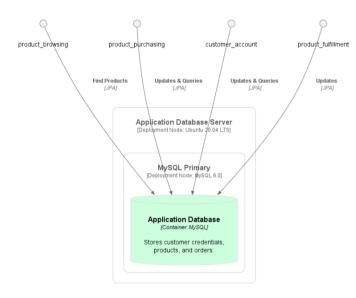


- We scaled a stateless service.
- Stateless: Services don't require persistent data between requests.
- Persistent state is saved in the database.
- This is normally easy to do.

What is the *problem*?

The database

Database



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

Disclaimer

This is not a database course.



HOME STARTING AT UQ PROGRAMS AND COURSES FACULTIES AND SCHOOLS

Advanced Database Systems





Print A Feedback

Advanced Database Systems (INFS3200)

Course level	Current cours	Current course offerings					
Undergraduate	Course offerings	Location	Mode	Course Profile			
Faculty	Semester 1, 2022	St Lucia	Internal	COURSE PROFILE			
Engineering, Architecture & Information	Semester 1, 2022	External	External	COURSE PROFILE			
Technology	Semester 2, 2022	External	External	PROFILE UNAVAILABLE			
School	Semester 2, 2022	St Lucia	Internal	PROFILE UNAVAILABLE			
Info Tech & Elec Engineering	Please Note: Course	Please Note: Course profiles marked as not available may still be in development.					
Units							
	Course description						
2	Course descr	iption					
	Distributed databas	e design, query					
2 Duration One Semester	Distributed databas warehousing, data of	e design, query cleansing, mana					
Duration	Distributed databas	e design, query cleansing, mana					
Duration One Semester Class contact 2 Lecture hours, 1 Tutorial hour, 1 Practical or	Distributed databas warehousing, data of	e design, query cleansing, mana					
Duration One Semester Class contact 2 Lecture hours, 1 Tutorial hour, 1 Practical or Laboratory hour	Distributed databas warehousing, data distributed devices.	e design, query cleansing, mana		tion processing, data integration, d patial data, and data from large sci Course Profile			
Duration One Semester Class contact 2 Lecture hours, 1 Tutorial hour, 1 Practical or Laboratory hour Incompatible	Distributed databas warehousing, data of distributed devices. Archived offer	e design, query cleansing, mana rings	gement of s	patial data, and data from large sco			
Duration One Semester Class contact 2 Lecture hours, 1 Tutorial hour, 1 Practical or Laboratory hour Incompatible INFS7907	Distributed databas warehousing, data of distributed devices. Archived offer Course offerings	e design, query cleansing, mana rings Location	gement of s	patial data, and data from large sco			
Duration One Semester Class contact 2 Lecture hours, 1 Tutorial hour, 1 Practical or Laboratory hour Incompatible	Distributed databas warehousing, data of distributed devices. Archived offer Course offerings Semester 1, 2021	e design, query cleansing, mana rings Location St Lucia	gement of s Mode Flexible Deliv	patial data, and data from large sca Course Profile			
Duration One Semester Class contact 2 Lecture hours, 1 Tutorial hour, 1 Practical or Laboratory hour Incompatible INFS7907	Distributed databas warehousing, data distributed devices. Archived offer Course offerings Semester 1, 2021 Semester 1, 2021	e design, query cleansing, mana rings Location St Lucia External	Mode Flexible Deliv	Course Profile COURSE PROFILE COURSE PROFILE			

This is a database course.

How do we fix database scaling issues?

How do we fix database scaling issues?

Answer

Replication

How do we fix database scaling issues?

- Replication
- Partitioning

How do we fix database scaling issues?

- Replication
- Partitioning
- Independent databases

How do we fix database scaling issues?

- Replication
- Partitioning
- Independent databases

What is *replication*?

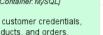
Definition 1. Replication Data copied across multiple different machines.

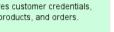
Application Database [Container: MySQL]

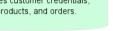
Application Database [Container: MySQL]

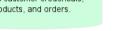
Stores customer credentials, products, and orders.

ores custo	
,	











name

name

product_id

product_id

1234

4321

1234

4321



Lifelike Elephant Inflatable

Nicholas Cage Reversable Pillow

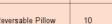
Lifelike Elephant Inflatable

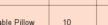












stock

stock

10





price



\$50.00

price

\$10.00

\$50.00







Definition 2. Replica

Database node which stores a copy of the data.



What are the advantages of *replication*?

What are the advantages of *replication*?

Answer

• Scale our database to cope with higher loads.

What are the advantages of *replication*?

- Scale our database to cope with higher loads.
- Provide *fault tolerance* from a single instance failure.

What are the advantages of *replication*?

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

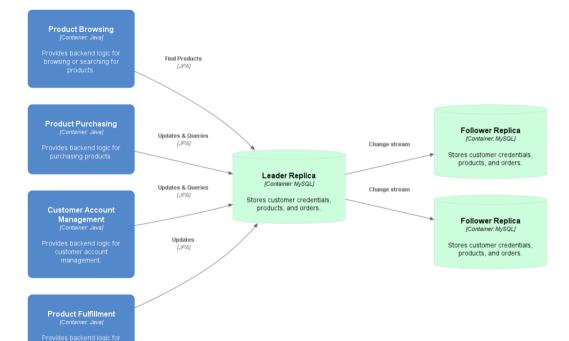
- Scalability
- Reliability
- Performance

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.

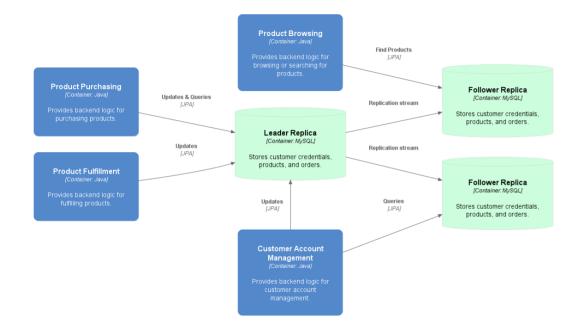
Leader-based Replication

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

Leader-based Replication

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

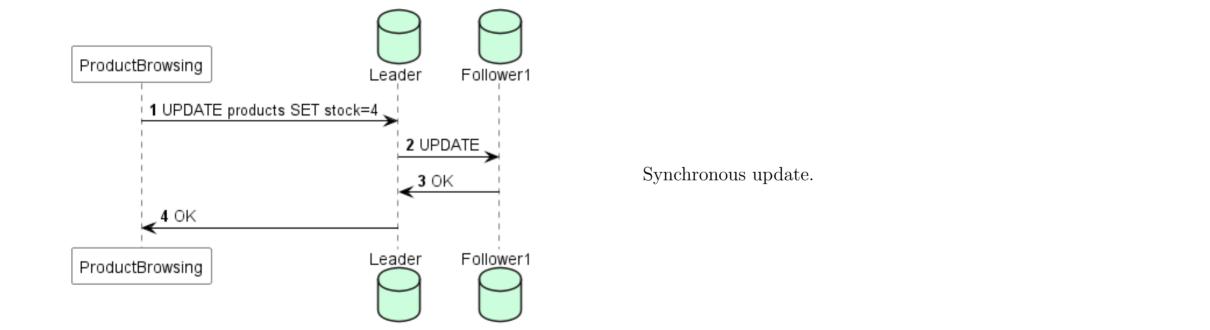
On read Any replica can be queried.

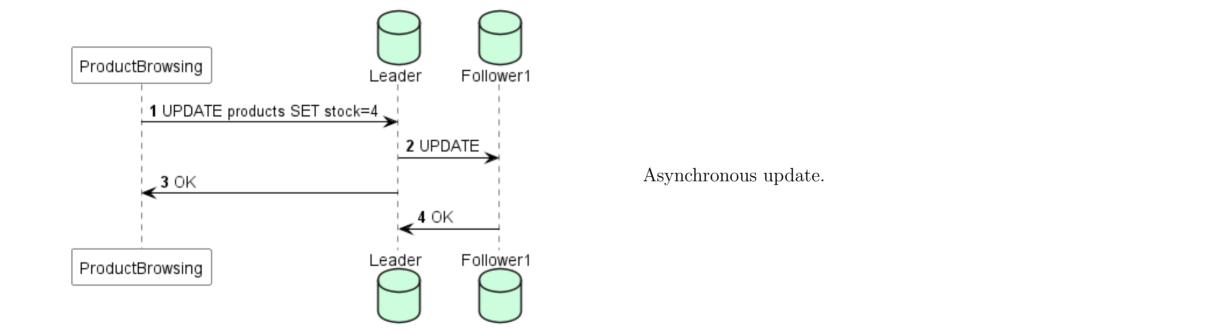


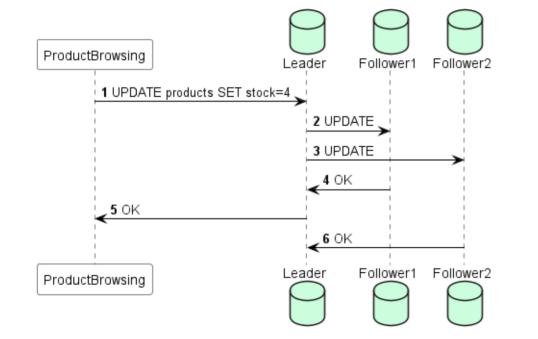
- Built-in to PostgreSQL, MySQL, MongoDB, RethinkDB, and Espresso.
- Can be added to Oracle and SQL Server.

Propagating changes

Synchronous vs. Asynchronous







- 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.

Synchronous Propagation

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

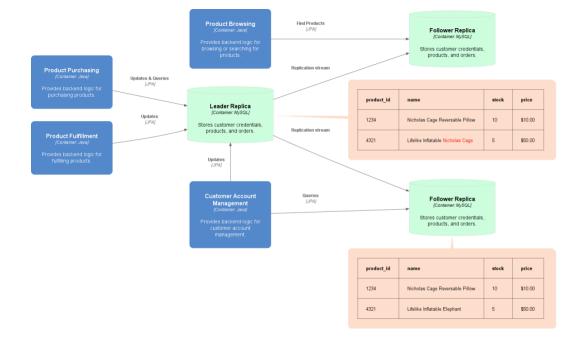
Asynchronous Propagation

- 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.

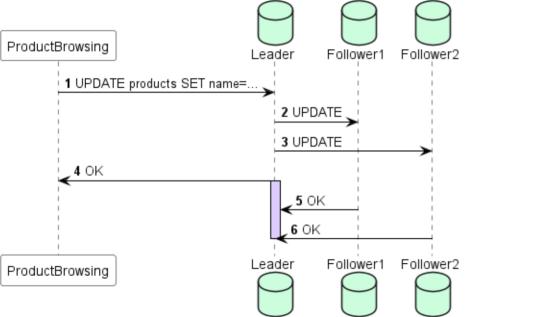
Definition 3. Replication Lag

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.

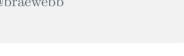
- If writes stop for long enough.
- Eventually is intentionally *ambiguous*.

Eventual Consistency
Problems?



1. Read user details.

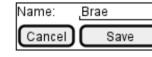




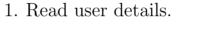
Name: Brae

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





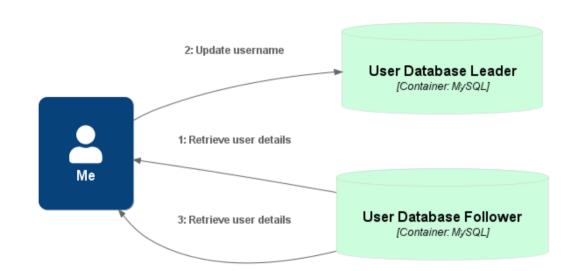




2. Decide I don't like my name.

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.

Users always see the updates that *they have made*.



Definition 4. Read-your-writes Consistency

Doesn't care what other users see.



1. Misspell a tweet.



@braewebb

Brae Webb

My first post

1. Misspell a tweet.

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





My fist post



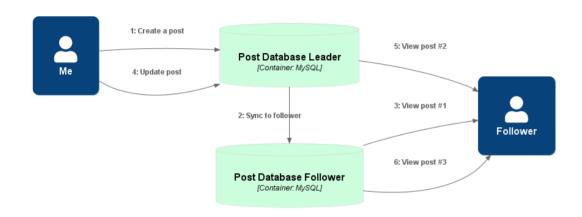


@braewebb

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*.

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

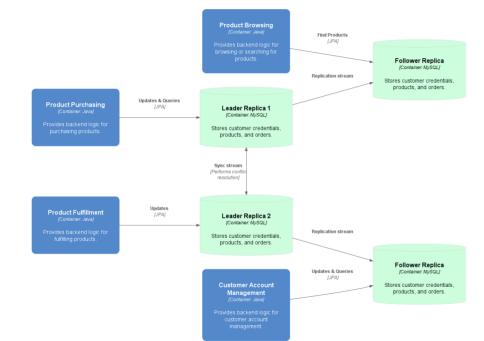
Definition 5. Monotonic Reads User doesn't travel back in time.

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?

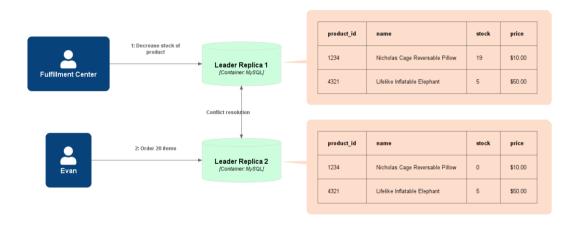
Question

What might go wrong?

Answer

Write conflicts

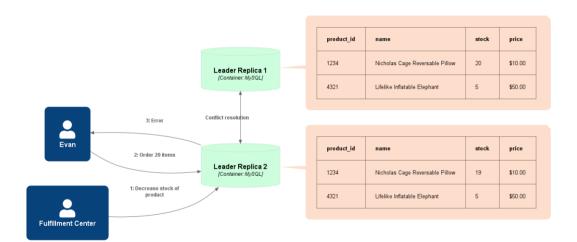
Conflict needs to be *resolved*.



- 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

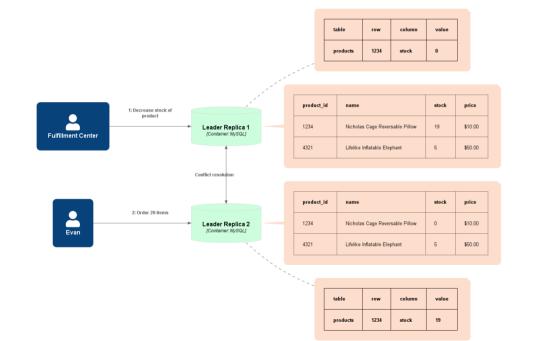
- Assign each *write* a unique ID.
- Assign each *leader replica* a unique ID.

Convergence Strategies

- , . .
- Assign each *write* a unique ID.

Yes, this can be *challenging*.

- Assign each *leader replica* a unique ID.
- Custom resolution logic.



$Resolving\ Conflicts$

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

On Read When a conflict is next read, ask for a resolution.

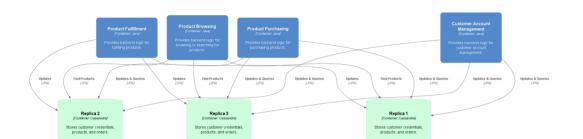
- \bullet Bucardo allows a perl script for on write resolution.
- CouchDB prompts reads to resolve the conflict.

Third Approach

Leaderless Replication

• Early distributed databases were leaderless. • Resurgance after Amazon created Dynamo.

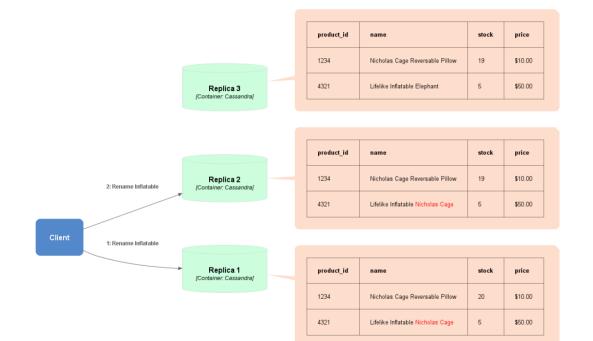
• Dynamo is an internal service and *not* DynamoDB. • Riak, Cassandra, and Voldemort are leaderless databases.



Reads and writes can be written to any node.

How do they work?

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



Leaderless Write



Lifelike Inflatable Nicholas Cage

\$50.00

4321

2: Query Products

1: Query Products

Client

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.

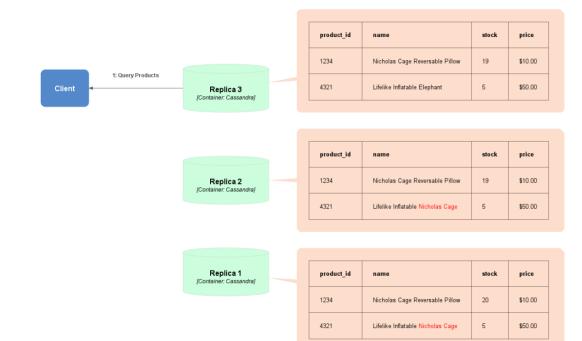
How are changes propagated?

- Read Repair
- Anti-Entropy Process

- updated data to that replica.
 - 2. Anti-Entropy Process: Background process looks for stale or missing data and updates replicas.

1. Read Repair: Client detects stale data on read and writes

How do we know it's consistent?



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

How do we know it's consistent?

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

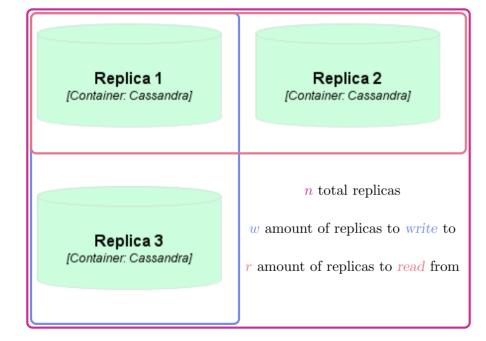
$$1 + 3 > 3$$

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.



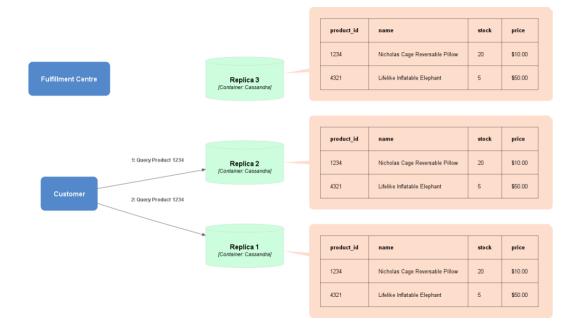
- Graphical representation of previous equation.
- Orange inner group (reads), overlaps with Blue inner group (writes).
 - Showing how reads overlap writes via a quorum.

What about write conflicts?

What about write conflicts?

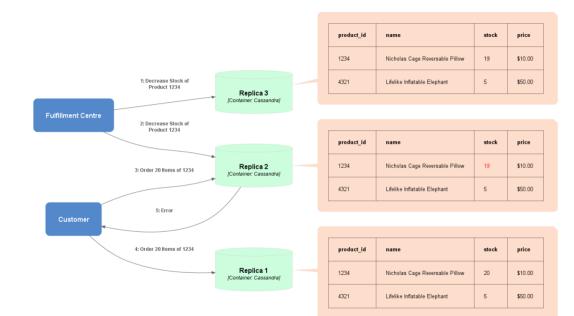
Answer

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?

• Replication copies data to multiple replicas.

- Replication copies data to multiple replicas.
- Leader-based replication is most common and simpliest.

- Replication copies data to multiple replicas.
- Leader-based replication is most common and simpliest.
- Replication introduces *eventual consistency*.

- Replication copies data to multiple replicas.
- Leader-based replication is most common and simpliest.
- Replication introduces *eventual consistency*.
- *Multi-leader* replication scales writes as well as reads but introduces *write conflicts*.

- Replication copies data to multiple replicas.
- Leader-based replication is most common and simpliest.
- 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.

How do we fix database scaling issues?

How do we fix database scaling issues?

Answer

- Replication
- Partitioning
- Independent databases

How do we fix database scaling issues?

Answer

- Replication
- Partitioning
- Independent databases

Definition 6. Partitioning

Split the data of a system onto multiple nodes.

Also called shards, regions, tablets, etc.

These nodes are *partitions*.

Application Database [Container: MySQL]

Stores customer credentials, products, and orders.

product_id	name	stock	price
4321	Lifelike Elephant Inflatable	5	\$50.00

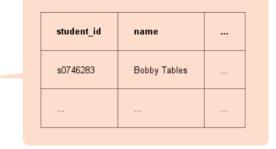
Application Database [Container: MySQL]

Stores customer credentials, products, and orders.

product_id	name	stock	price
1234	Nicholas Cage Reversable Pillow	10	\$10.00

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

How should we decide which data is stored where?



Student Database [Partition 1]	student_id	name	
Student ID ranges [s1000001-s2000000]	s1637285	Brae Webb	

Student Database

(Partition 0)
Student ID ranges
[s0000000-s1000000]

An example partitioning based on primary key, student ID.

What is the problem with this?

What is the problem with this?

Answer

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

How should we decide which data is stored where?

How should we decide which data is stored where?

Answer

Maximize spread of requests, avoiding *skewing*.

Have we seen this before?

Have we seen this before?

Answer

Hashing?

Hash tables hash entries to maximize the spread between buckets.

What is the problem with this?

What is the problem with this?

Answer

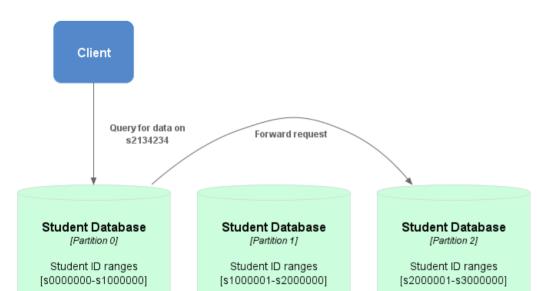
Range queries are inefficient, i.e. get all students between s4444444 and s4565656.

How do we route queries?

Unlike stateless, only one node can process queries.

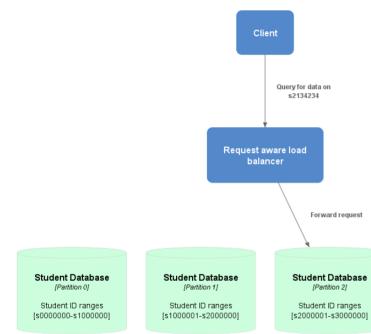
Query-insensitive Load Balancer

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



Query-sensitive Load Balancer

A load balancer which understands which queries should be forwarded to which node.



Client-aware Queries

Place the responsibility on clients to choose the correct node.



Query for data on s2134234

Student Database [Partition 0]

Student ID ranges [s0000000-s1000000] Student Database [Partition 1]

Student ID ranges [s1000001-s2000000] Student Database [Partition 2]

Student ID ranges

[s2000001-s3000000]

• Partitioning splits data across multiple nodes.

- Partitioning splits data across multiple nodes.
- A *consistent method* to chose which node is required.

- Partitioning splits data across multiple nodes.
- A *consistent method* to chose which node is required.
- Partitioning by *primary key* can create *skewing*.

- Partitioning splits data across multiple nodes.
- A *consistent method* to chose which node is required
- required.

 Partitioning by *primary key* can create
- skewing.
 Partitioning by hash makes range queries less efficient.

Partitioning splits data across multiple

- nodes.
 A consistent method to chose which node is
- required.

 Partitioning by mim any key can create
- 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.

Question

How do we fix database scaling issues?

Question

How do we fix database scaling issues?

Answer

- Replication
- Partitioning
- Independent databases

Question

How do we fix database scaling issues?

Answer

- Replication
- Partitioning
- Independent databases

Replications

- Replications
 - Leader-based, multi-leader, and leaderless

- Replications
 - Leader-based, multi-leader, and leaderless
 - Eventual consistency

- Replications
 - Leader-based, multi-leader, and leaderless
 - Eventual consistency
 - Write conflicts

- Replications
 - Leader-based, multi-leader, and leaderless
 - Eventual consistency
 - Write conflicts
- Partitioning

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

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