Sharding data across multiple data stores

Querying & processing (Big) data

Genoveva Vargas-Solar
Senior Scientist, French Council of Scientific Research, LIG-LAFMIA
genoveva.vargas@imag.fr

http://vargas-solar.com

DATA PROCESSING

Process the data to produce other data: analysis tool, business intelligence tool, ...

This means

- Handle large volumes of data
- Manage thousands of processors
- Parallelize and distribute treatments
  - Scheduling I/O
  - Managing Fault Tolerance
  - Monitor /Control processes

Map-Reduce provides all this easy!

QUERY EXECUTION

"Optimizing query processing"
→ DATA COMPRESSION, LOCATION, R/W STRATEGIES, CACHE/MEMORY ORGANIZATION

Query

Map

Reduce

2

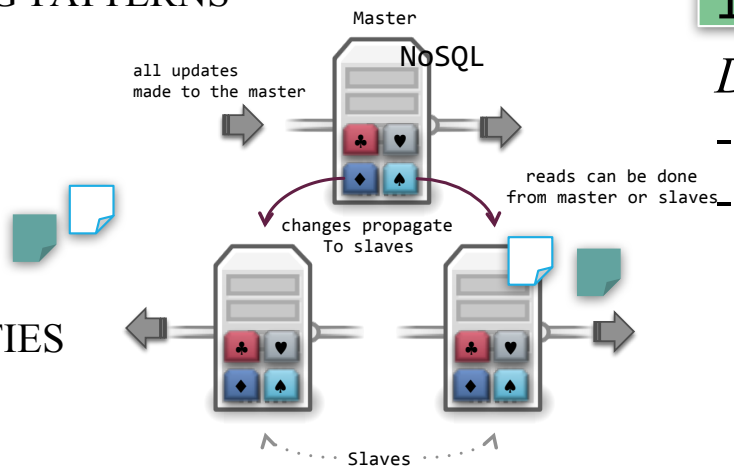
Query programming
→ MAP-REDUCE PROGRAMMING PATTERNS

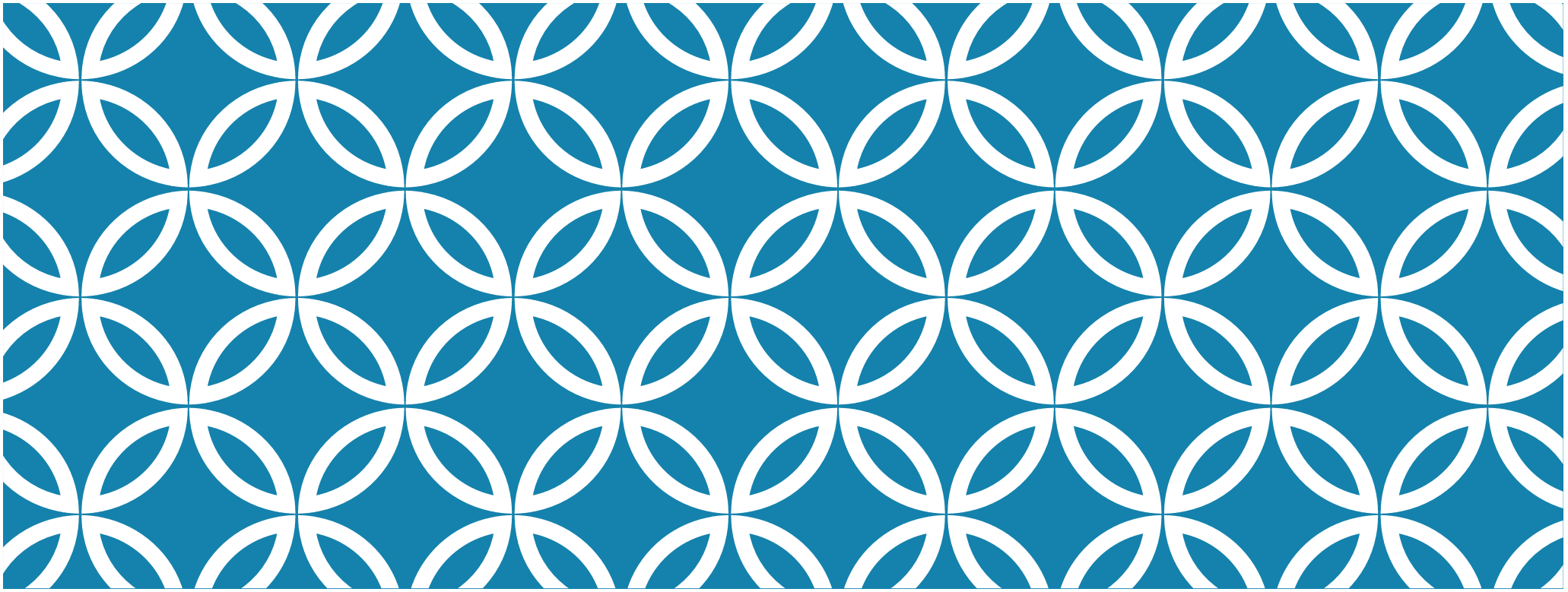
1

Data organization
- DISTRIBUTED FILE SYSTEM
- INDEXING

3

Query execution
- EXECUTION MODEL
- DATA PROCESSING PROPERTIES





DATA ORGANIZATION

DATA MANAGEMENT PROPERTIES

Indexing

- Distributed hashing like *Memcached* open source cache
  - In-memory indexes are scalable when distributing and replicating objects over multiple nodes
- Partitioned tables

High availability and scalability: eventual consistency

- Data fetched are not guaranteed to be up-to-date
- Updates are guaranteed to be propagated to all nodes eventually

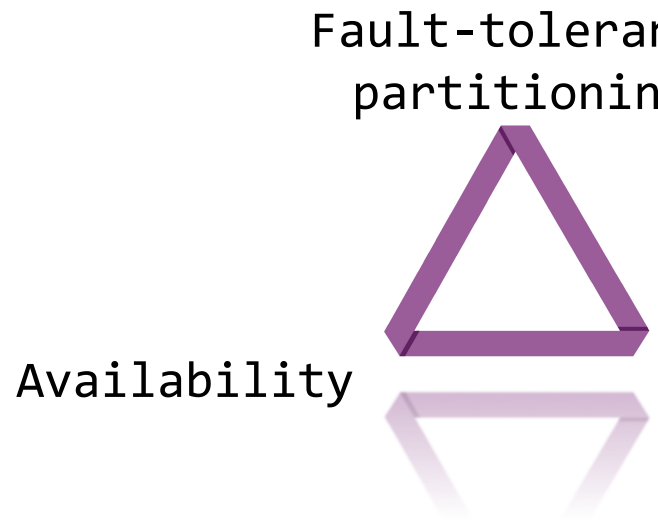
Shared nothing horizontal scaling

- Replicating and partitioning data over many servers
- Support large number of simple read/write operations per second (OLTP)

No ACID guarantees

- Updates eventually propagated but limited guarantees on reads consistency
- BASE: basically available; soft state, eventually consistent
- Multi-version concurrency control

PROBLEM STATEMENT: HOW MUCH TO GIVE UP?



Strict
The changes to the data are atomic and appear to take effect instantaneously. This is the highest form of consistency.

Sequential
Every client sees all changes in the same order they were applied.

Causal
All changes that are causally related are observed in the same order by all clients.

Eventual
When no updates occur for a period of time, eventually all updates will propagate through the system and all replicas will be consistent.

Weak
No guarantee is made that all updates will propagate and changes may appear out of order to various clients.

CAP theorem[1]: a system can have two of the three properties

NoSQL systems sacrifice **consistency**

1 Eric Brewer, "Towards robust distributed systems." PODC. 2000 http://www.cs.berkeley.edu/~brewer/cs262b-2004/PODC-keynote.pdf

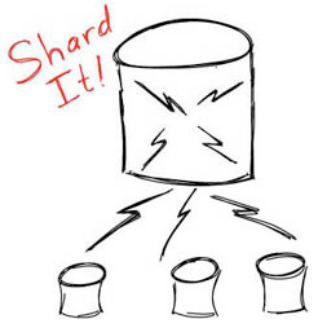
AVAILABILITY & PERFORMANCE

Replication

- Copy data across multiple servers (each bit of data can be found in multiple servers)
- Increase data availability
- Faster query evaluation

Sharding

- Distribute different data across multiple servers
- Each server acts as the single source of a data subset



Orthogonal techniques

REPLICATION: PROS & CONS

Data is more available

- Failure of a site containing E does not result in unavailability of E if replicas exist

Performance

- Parallelism: queries processed in parallel on several nodes
- Reduce data transfer for local data

Increased updates cost

- Synchronisation: each replica must be updated

Increased complexity of concurrency control

- Concurrent updates to distinct replicas may lead to inconsistent data unless special concurrency control mechanisms are implemented

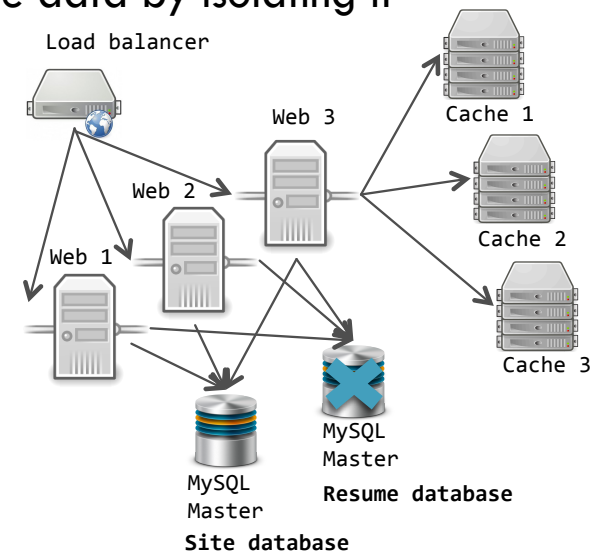
SHARDING WHY IS IT USEFUL?

Scaling applications by reducing data sets in any single databases

Segregating data

Sharing application data

Securing sensitive data by isolating it



Improve read and write performance

- Smaller amount of data in each user group implies faster querying
- Isolating data into smaller shards accessed data is more likely to stay on cache
- More write bandwidth: writing can be done in parallel
- Smaller data sets are easier to backup, restore and manage

Massively work done

- Parallel work: scale out across more nodes
- Parallel backend: handling higher user loads
- Share nothing: very few bottlenecks

Decrease resilience improve availability

- If a box goes down others still operate
- But: Part of the data missing

SHARDING & REPLICATION

Sharding with no replication: unique copy, distributed data sets

- (+) Better concurrency levels (shards are accessed independently)
- (-) Cost of checking constraints, rebuilding aggregates
- Ensure that queries and updates are distributed across shards

Replication of shards

- (+) Query performance (availability)
- (-) Cost of updating, of checking constraints, complexity of concurrency control

Partial replication (most of the times)

- Only some shards are duplicated

Genoveva Vargas-Solar

CR1, CNRS, LIG-LAFMIA

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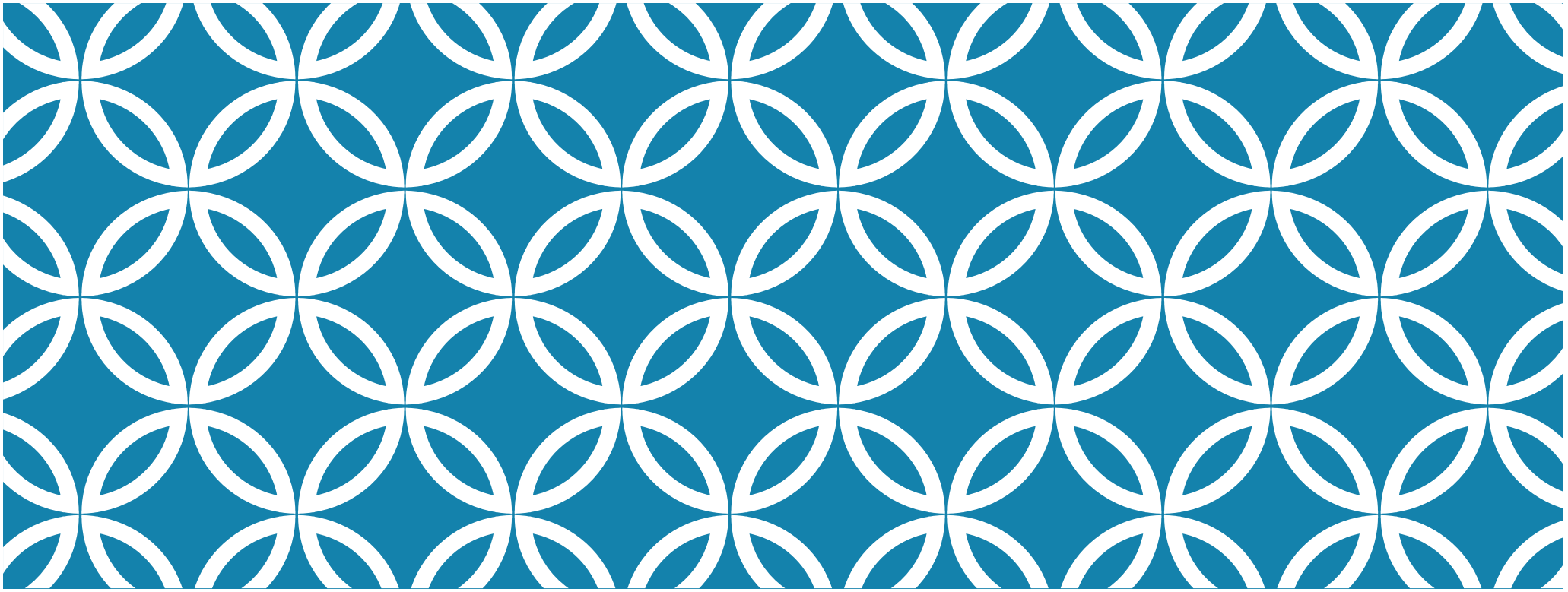
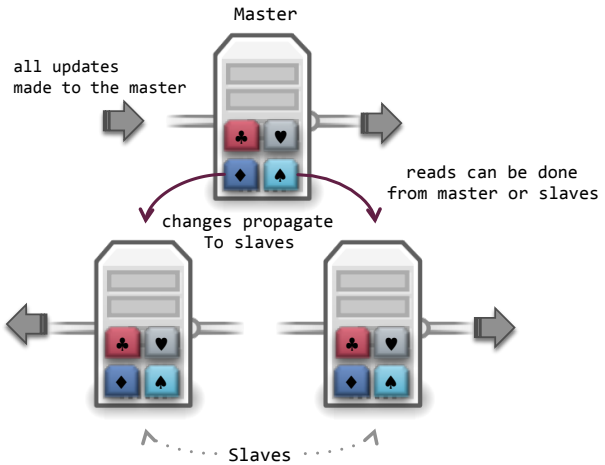


AVAILABILITY AND PERFORMANCE

REPLICATION MASTER - SLAVE



Makes one node the authoritative copy/replica that handles writes while replica synchronize with the master and may handle reeds

All replicas have the same weight

- Replicas can all accept writes
- The lose of one of them does not prevent access to the data store

Helps with read scalability but does not help with write scalability

Read resilience: should the master fail, slaves can still handle read requests

Master failure eliminates the ability to handle writes until either the master is restored or a new master is appointed

Biggest complication is consistency

- Possible write – write conflict
- Attempt to update the same record at the same time from to different places

Master is a bottle-neck and a point of failure

MASTER-SLAVE REPLICATION MANAGEMENT

Masters can be appointed

- Manually when configuring the nodes cluster
- Automatically: when configuring a nodes cluster one of them elected as master. The master can appoint a new master when the master fails reducing downtime

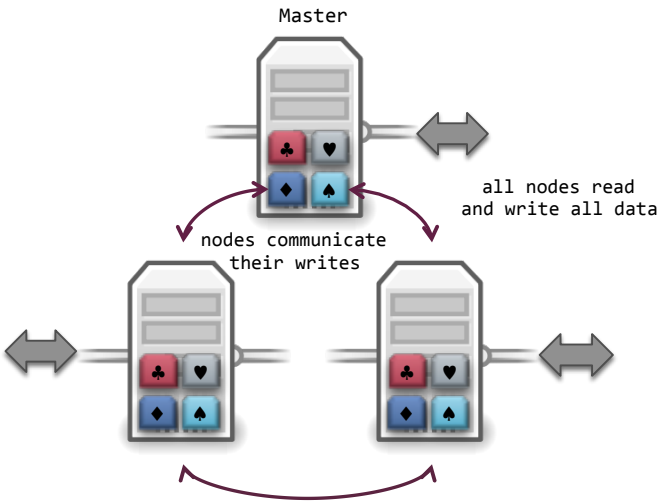
Read resilience

- Read and write paths have to be managed separately to handle failure in the write path and still reads can occur
- Reads and writes are put in different database connections if the database library accepts it

Replication comes inevitably with a dark side: **inconsistency**

- Different clients reading different slaves will see different values if changes have not been propagated to all slaves
- In the worst case a client cannot read a write it just made
- Even if master-slave is used for hot backups, if the master fails any updates on to the backup are lost

REPLICATION: PEER-TO-PEER



Allows writes to any node; the nodes coordinate to synchronize their copies

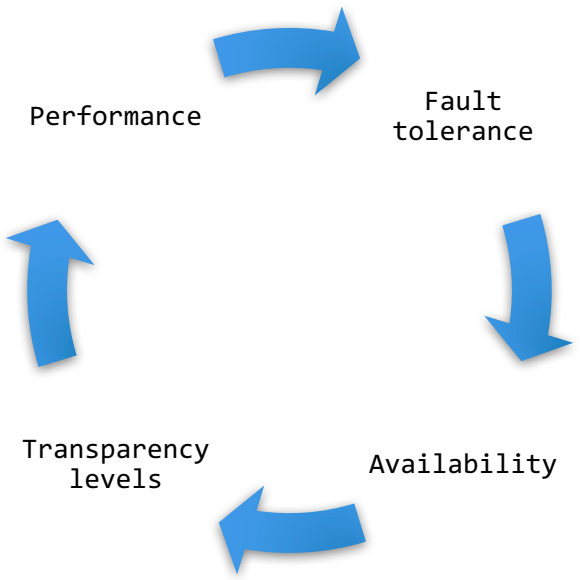
The replicas have equal weight

Deals with inconsistencies

- Replicas coordinate to avoid conflict
- Network traffic cost for coordinating writes
- Unnecessary to make all replicas agree to write, only the majority
- Survival to the loss of the minority of replicas nodes
- Policy to merge inconsistent writes
- Full performance on writing to any replica

REPLICATION: ASPECTS TO CONSIDER

Conditioning



Important elements to consider

- Data to duplicate
- Copies location
- Duplication model (master – slave / P2P)
- Consistency model (global – copies)

→ **Find a compromise !**

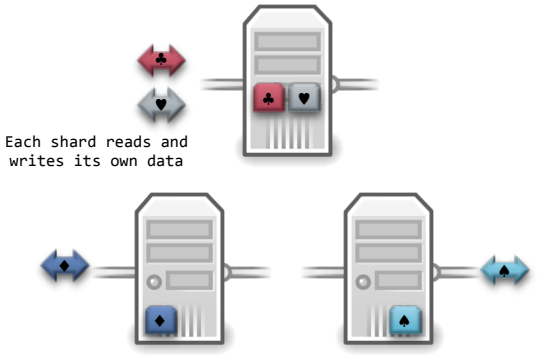
SHARDING

Puts different data on separate nodes

- Each user only talks to one servicer so she gets rapid responses
- The load should be balanced out nicely between servers

Ensure that

- data that is accessed together is clumped together on the same node
- that clumps are arranged on the nodes to provide best data access



Ability to distribute both data and load of simple operations over many servers, with no RAM or disk shared among servers

A way to horizontally scale **writes**

Improve **read** performance

Application/data store support

SHARDING

Database laws

Small databases are fast

Big databases are slow

Keep databases small

Principle

Start with a big monolithic database

- Break into smaller databases
- Across many clusters
- Using a key value

Instead of having one million customers information on a single big machine ….

*100 000 customers on **smaller** and **different** machines*

SHARDING CRITERIA

Partitioning

- Relational: handled by the DBMS (homogeneous DBMS)
- NoSQL: based on ranging of the k-value

Federation

- Relational
  - Combine tables stored in different physical databases
  - Easier with denormalized data
- NoSQL:
  - Store together data that are accessed together
  - Aggregates unit of distribution

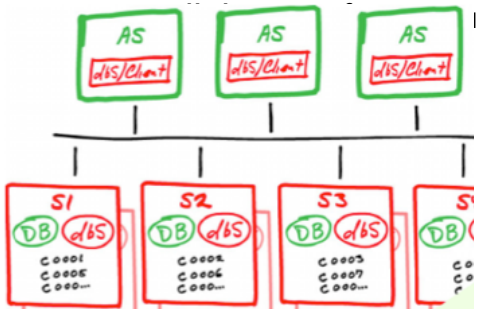
SHARDING

Architecture

Each application server (AS) is running DBS/client

Each shard server is running

- a database server
- replication agents and query agents for su lity



Process

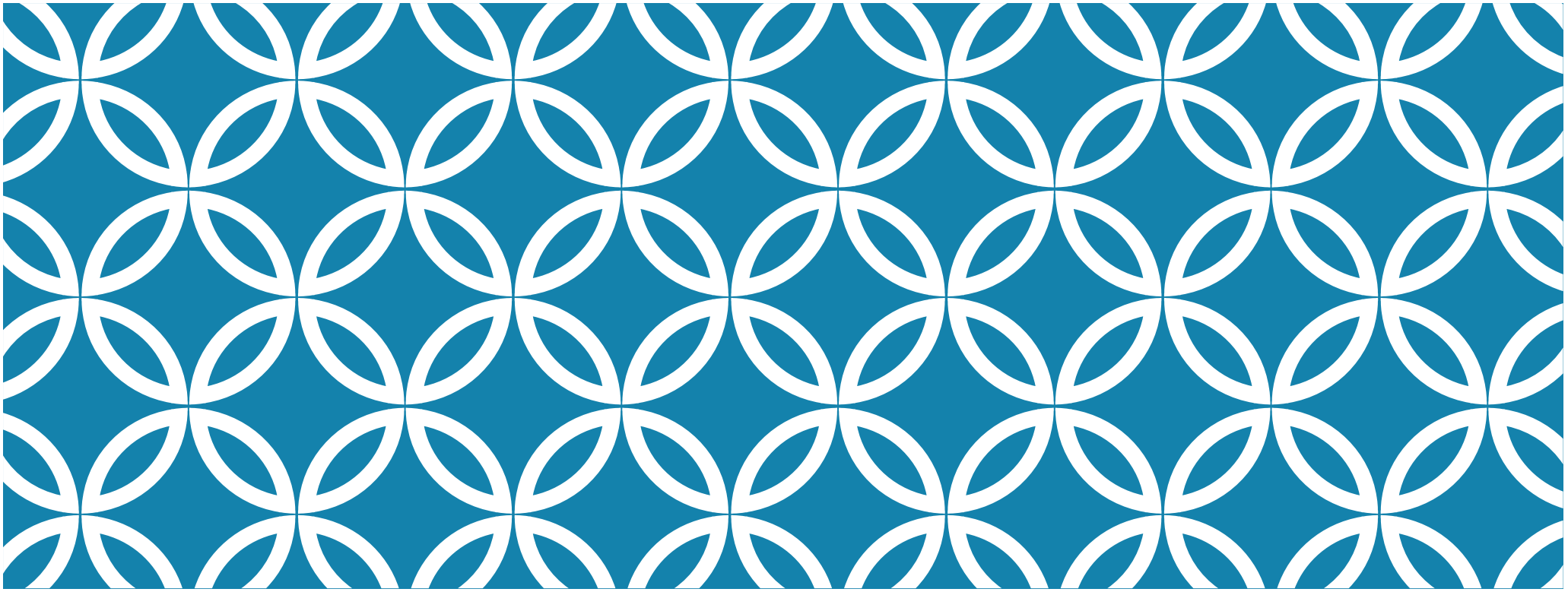
Pick a dimension that helps sharding easily (customers, countries, addresses)

Pick strategies that will last a long time as repartition/re-sharding of data is operationally difficult

This is done according to two different principles

- **Partitioning:** a partition is a structure that divides a space into tow parts
- **Federation:** a set of things that together compose a centralized unit but each individually maintains some aspect of autonomy

Customers data is partitioned by ID in shards using an algorithm d to determine which shard a customer ID belongs to



PARTITIONING

A partition is a structure that divides a space into tow parts

BACKGROUND: DISTRIBUTED RELATIONAL DATABASES

External schemas (views) are often subsets of relations (contacts in Europe and America)

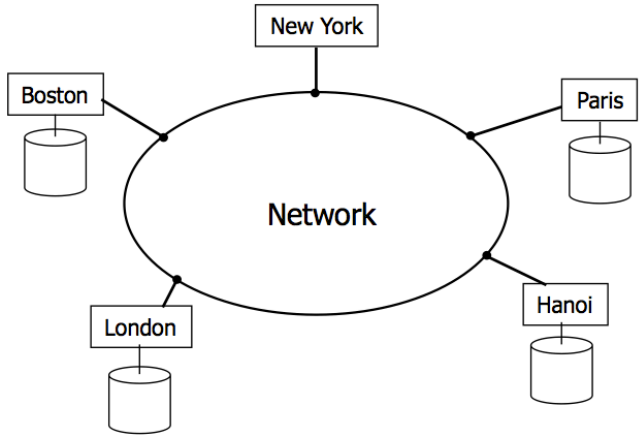
Access defined on subsets of relations: 80% of the queries issued in a region have to do with contacts of that region

Relations partition

- Better concurrency level
- Fragments accessed independently

Implications

- Check integrity constraints
- Rebuild relations



- EMP(ENO, ENAME, TITLE)
- PROJ(PNO, PNAME, BUDGET, LOC)
- PAY(TITLE, SAL)
- ASG(ENO, PNO, DUR, RESP)

FRAGMENTATION

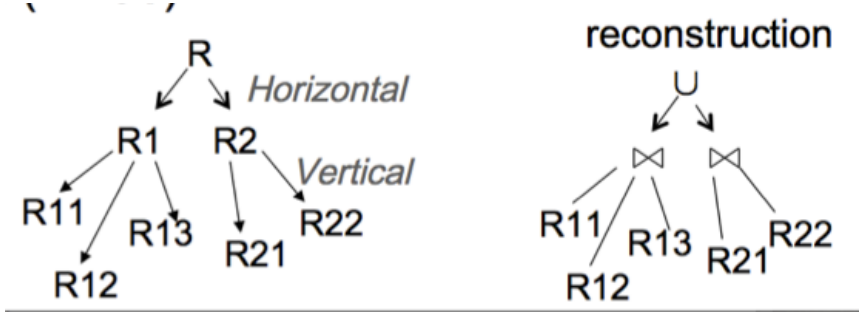
Horizontal

- Groups of tuples of the same relation
- Budget < 300 000 or >= 150 000
- Not disjoint are more difficult to manage

Vertical

- Groups attributes of the same relation
- Separate budget from loc and pname of the relation project

Hybrid



FRAGMENTATION: RULES

Vertical

Clustering

- Grouping elementary fragments
- Budget and location information in two relations

Splitting

- Decomposing a relation according to affinity relationships among attributes

Horizontal

Tuples of the same fragment must be statistically homogeneous

- If t1 and t2 are tuples of the same fragment then t1 and t2 have the same probability of being selected by a query

Keep important conditions

- Complete
  - Every tuple (attribute) belongs to a fragment (without information loss)
  - If tuples where budget >= 150 000 are more likely to be selected then it is a good candidate
- Minimum
  - If no application distinguishes between budget >= 150 000 and budget < 150 000 then these conditions are unnecessary

SHARDING: HORIZONTAL PARTITIONING

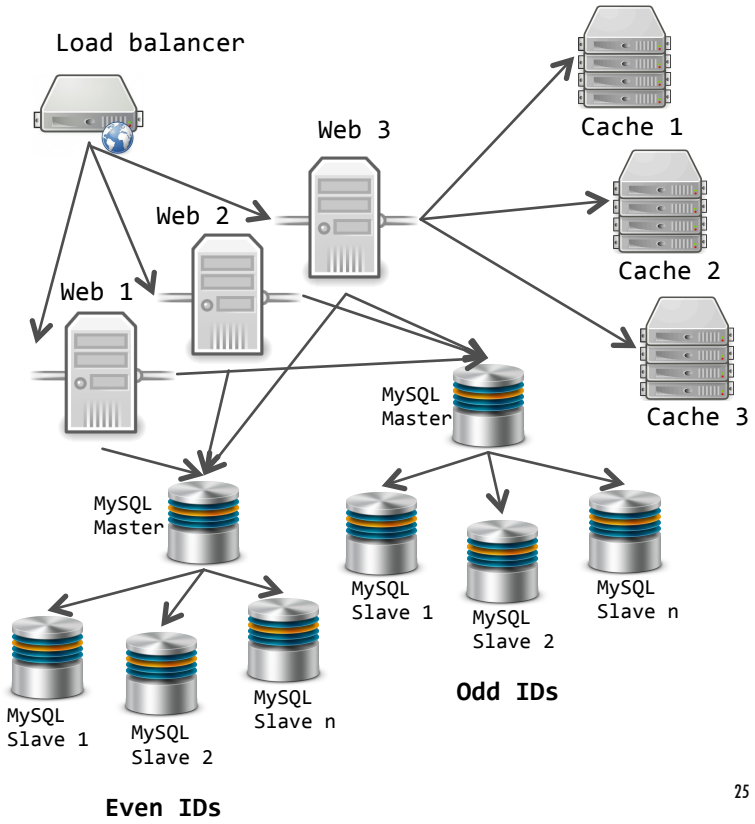
The entities of a database are split into two or more sets (by row)

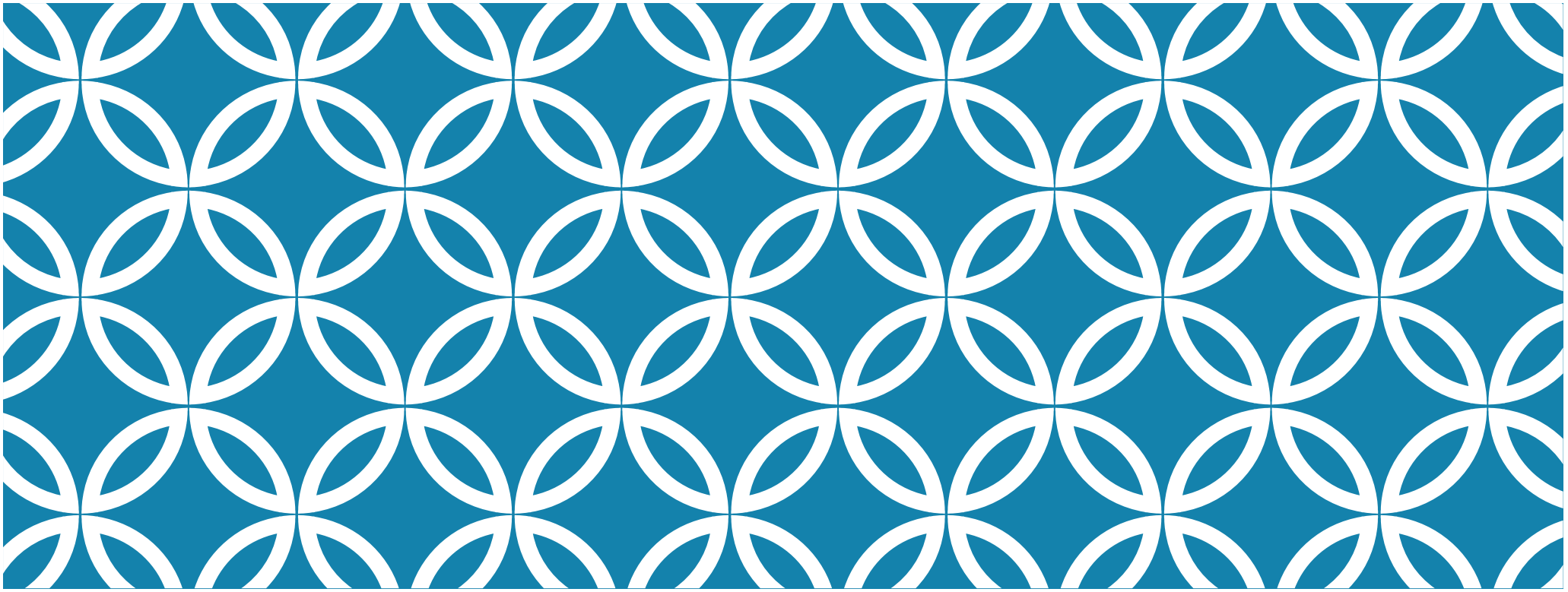
In relational: same schema several physical bases/servers

- Partition contacts in Europe and America shards where they zip code indicates where the will be found
- Efficient if there exists some robust and implicit way to identify in which partition to find a particular entity

Last resort shard

- Needs to find a sharding function: modulo, round robin, hash – partition, range - partition





FEDERATION

A federation is a set of things that together compose a centralized unit but each individually maintains some aspect of autonomy

FEDERATION: VERTICAL SHARDING

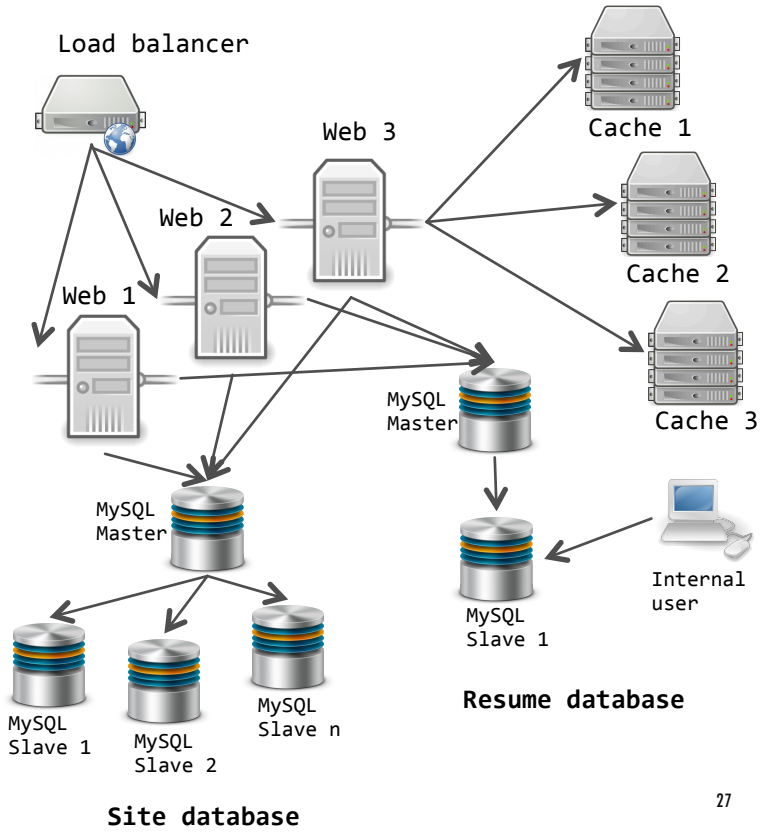
Principle

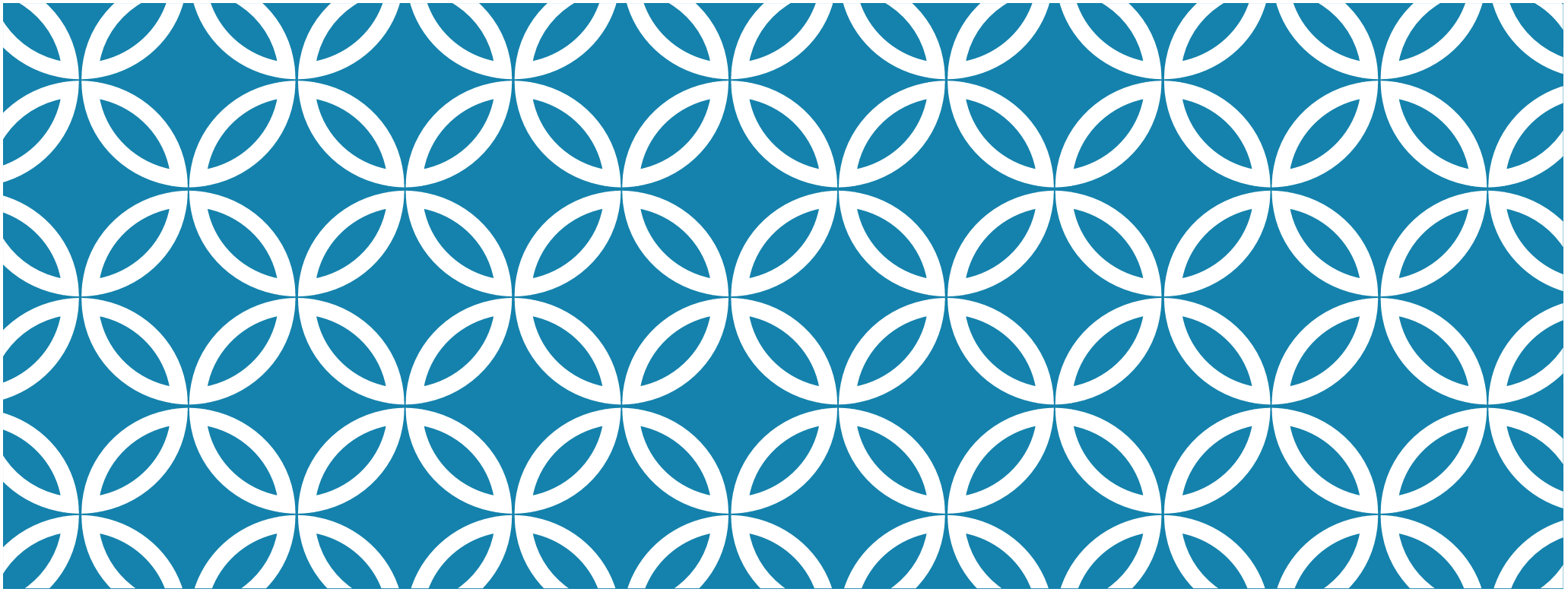
- Partition data according to their logical affiliation
- Put together data that are commonly accessed

The search load for the large partitioned entity can be split across multiple servers (logical and physical) and not only according to multiple indexes in the same logical server

Different schemas, systems, and physical bases/servers

Shards the components of a site and not only data





NOSQL STORES: PERSISTENCY MANAGEMENT

«*MEMCACHED*»

«*memcached*» is a memory management protocol based on a cache:

- Uses the key-value notion
- Information is completly stored in RAM

«*memcached*» protocol for:

- Creating, retrieving, updating, and deleting information from the database
- Applications with their own «*memcached*» manager (Google, Facebook, YouTube, FarmVille, Twitter, Wikipedia)

STORAGE ON DISC (1)

For efficiency reasons, information is stored using the RAM:

- Work information is in RAM in order to answer to low latency requests
- Yet, this is not always possible and desirable

➢**The process of moving data from RAM to disc is called "*eviction*"; this process is configured automatically for every bucket**

STORAGE ON DISC (2)

NoSQL servers support the storage of key-value pairs on disc:

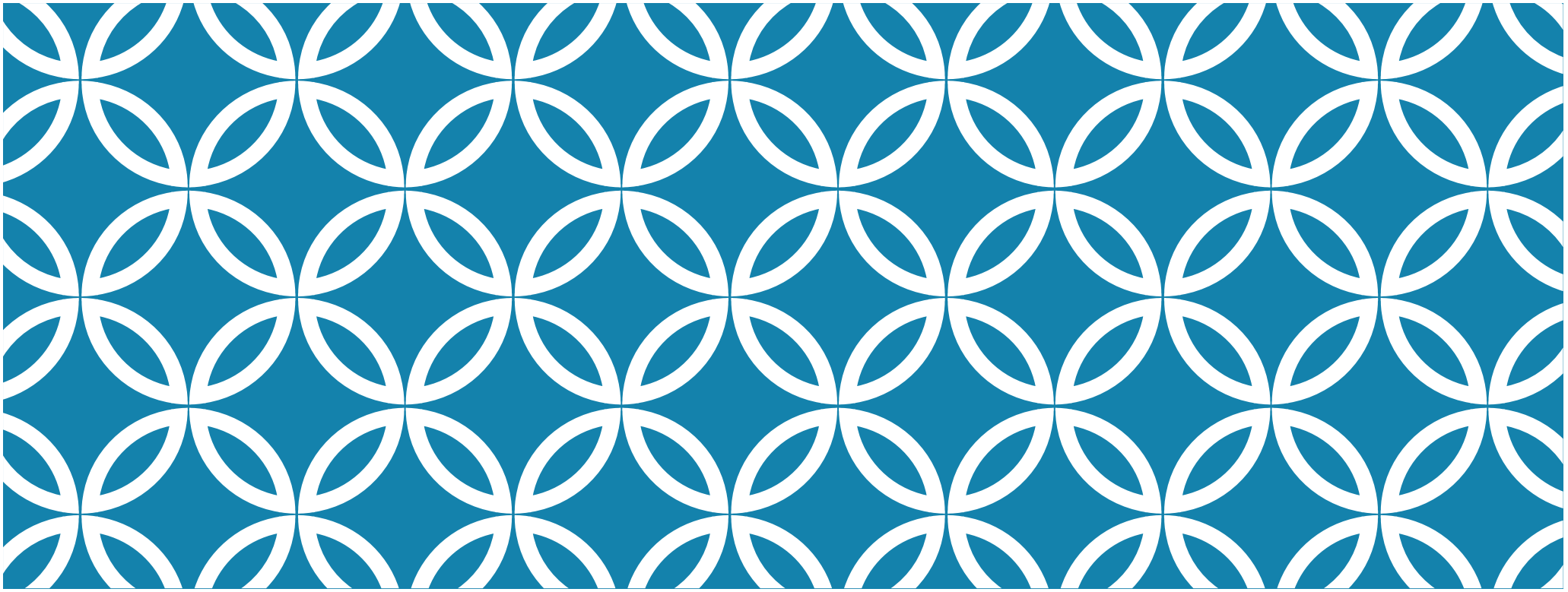
- **Persistency**–can be executed by loading data, closing and reinitializing it without having to load data from another source
- **Hot backups**– loaded data are sotred on disc so that it can be reinitialized in case of failures
- **Storage on disc**– the disc is used when the quantity of data is higher thant the physical size of the RAM, frequently used information is maintained in RAM and the rest es stored on disc

STORAGE ON DISC (3)

Strategies for ensuring:

- Each node maintains in RAM information on the key-value pairs it stores. Keys:
  - may not be found, or
  - they can be stored in memory or on disc
- The process of moving information from RAM to disc is asynchronous:
  - The server can continue processing new requests
  - A queue manages requests to disc

➢ **In periods with a lot of writing requests, clients can be notified that the server is termporaly out of memory until information is evicted**



CONCURRENCY CONTROL

MULTI VERSION CONCURRENCY CONTROL (MVCC)

Objective: Provide concurrent access to the database and in programming languages to implement transactional memory

Problem: If someone is reading from a database at the same time as someone else is writing to it, the reader could see a half-written or inconsistent piece of data.

Lock: readers wait until the writer is done

MVCC:

- Each user connected to the database sees a snapshot of the database at a particular instant in time
- Any changes made by a writer will not be seen by other users until the changes have been completed (until the transaction has been committed
- When an MVCC database needs to update an item of data it marks the old data as obsolete and adds the newer version elsewhere → multiple versions stored, but only one is the latest
- Writes can be isolated by virtue of the old versions being maintained
- Requires (generally) the system to periodically sweep through and delete the old, obsolete data objects