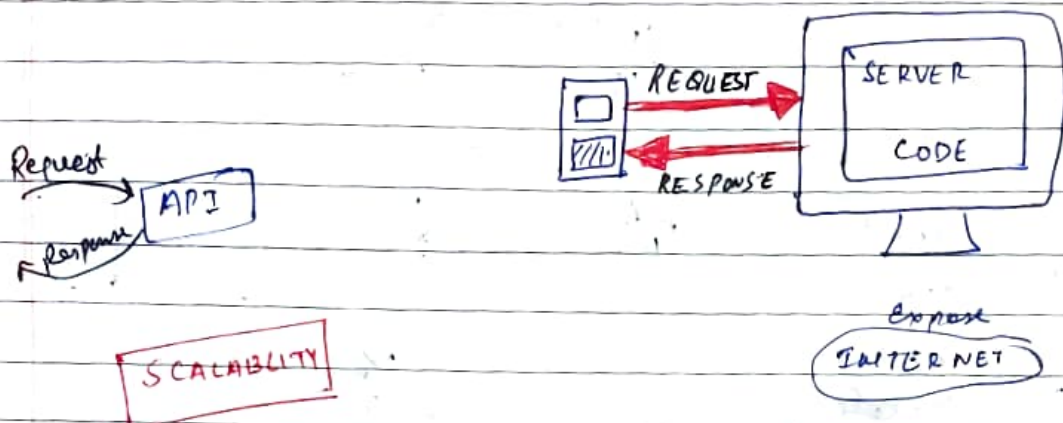


- Task 2
3. ATIR database #
3. ATIR Signup
- Atishi no sha
- Controller → Signup → await User. create data → as this user is
- Signup → is ATIR Signup & Signup → not ATIR → submit = Signup (event)
- same has change

1 SYSTEM DESIGN BASICS : HORIZONTAL vs VERTICAL SCALING



SCALABILITY

- Buy Bigger m/c - VERTICAL SCALING
- Buy MORE m/c - HORIZONTAL SCALING

Horizontal Scaling

- 1
- 2
- 3
- 4
- 5

Vertical Scaling

Huge Buy

- | | |
|---|--|
| <ol style="list-style-type: none"> Load Balancing Required RESILIENT - if 1 m/c fails it may be transferred to other m/c Network calls (Remote procedure calls) - slow Data Inconsistency - If there is transaction then we have to log all the servers i.e. all the DB which is impractical. we have some sort of loose transactional guarantee Scales well as USER increases | <ol style="list-style-type: none"> N/A Single point of failure Inter process communication - fast Consistent as one system where data reside Hardware limit - we cannot make m/c bigger & bigger what we want |
|---|--|

Hybrid scaling is Horizontal Scaling where each m/c have big box

2 WHY DO DATABASE FAILS? ANTIPATTERNS TO AVOID

Frequent Polling for load on DB / low interval for sufficient

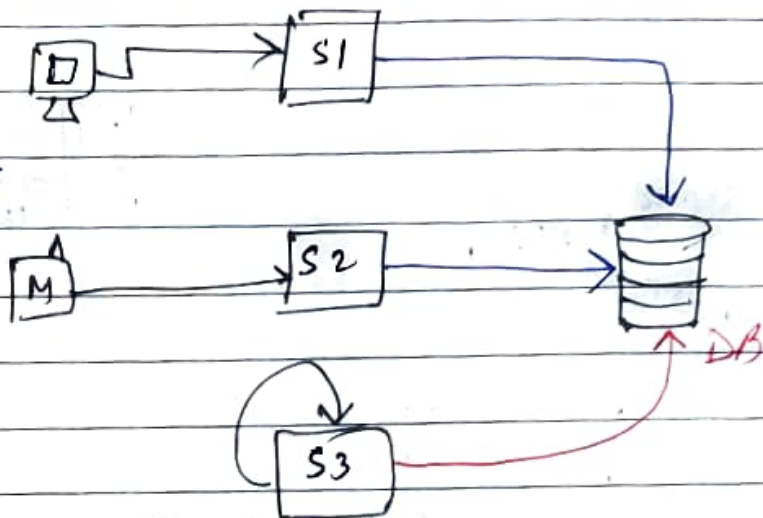
ANTI-PATTERN

DATABASES AS MESSAGE QUEUES

Ant

C	→ Create
R	→ Read
U	→ Update
D	→ Delete

INSERT



"Using DB as message queues".

"Using DB as message queues". If,

→ If we are using DB to send messages from one server to another that's considered as antipattern and we should avoid them

→ DB is optimised for either reading or writing not both

→ when we are going to do both operation at a time, it will have issues like locking, dead lock

→ Something like CRUD model

→ In this CRUD mode we do study that how much time do each of these operations take

→ DB are rarely designed such that the read & write DB optimal Data Structure. (Not possible to Design)

Here, interesting point is that what if our server are talking frequently i.e. set of message S1, S2 & S3 are send to each other. So the DB will filled up with lot of entries

so, we need to store all the data or we can start clearing it. If we are not going to delete the data then we need to update it & mark it as completed which is an expensive operation.

or we have to delete it manually or using cron jobs, which in both cases of deletion is going to be using script or using complicated logic

→ So 3rd problem comes for large system is scalability.

(S4) } also in this way DB cannot handle so many read operations. so if
(S5) } we are adding new DB which handle S4 & S5, then how will S4 talk to S2. For this we have to make complicated system which will act as broker b/w 2 DB. There is no need to do all this when we have system design concept **MESSAGE QUEUES**

• **ANTIPATTERN** here is that we are using **DATABASES AS MESSAGE QUEUES**. Instead we will use specialized message queues when we have very large systems.

Message Queue will avoid all these problems

→ polling interval is not going to come into consideration because message queues push the message to other server. ~~so in~~

→ so instead of server asking, message queue is going to give the message.

→ They are optimised so there's not too many read

operation. So writing to it is easy while reading is really the message queues ~~operation~~ responsibility. So it's going to do it in its own free time.

Scalability is also taken care bcoz if we need more message queues we can just add them.

Drawbacks of message queues.

- when system is not very large bcoz if we don't have those many servers then the DB will be able to handle loads that we throw on it.
- Cost of setting up the message queue along with training is large.
- Maybe overkill for a small service.

Summary

- ~~DB~~ DB are often used to store various types of information, but one case where it becomes an as a problem is when being used as a message broker.
- DB is rarely designed to deal with messaging features, & hence a poor substitute of a specialized message queue. When debugging a system, this pattern is considered an anti pattern.

Here are the possible drawbacks

- polling interval have to be set correctly. Too long makes the system inefficient, Too short makes the

DB undergo heavy read load

- Read & write operation heavy DB. usually, they are good at one of the two
- Manual delete procedures to be written to remove read messages
- Scaling is difficult conceptually & physically.

3 What is microservice architecture & its advantage

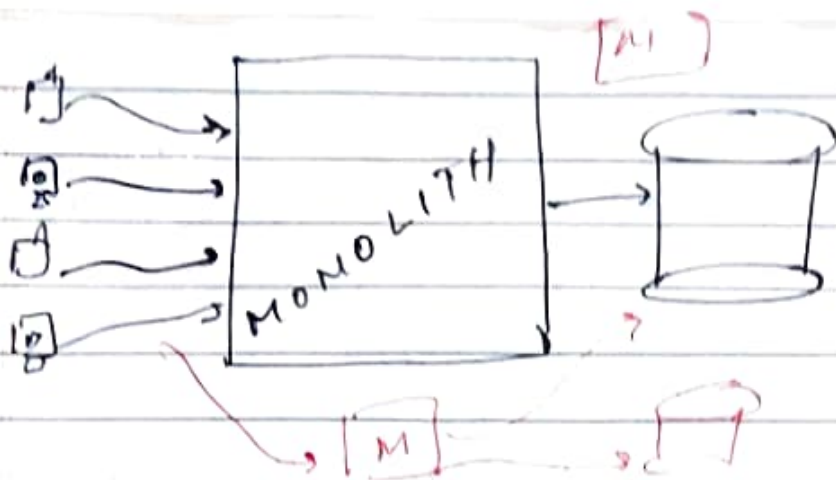
- we use microservices instead of monolithic system. short answer is: Scalability.

• Advantage (Microservices architecture favorable)

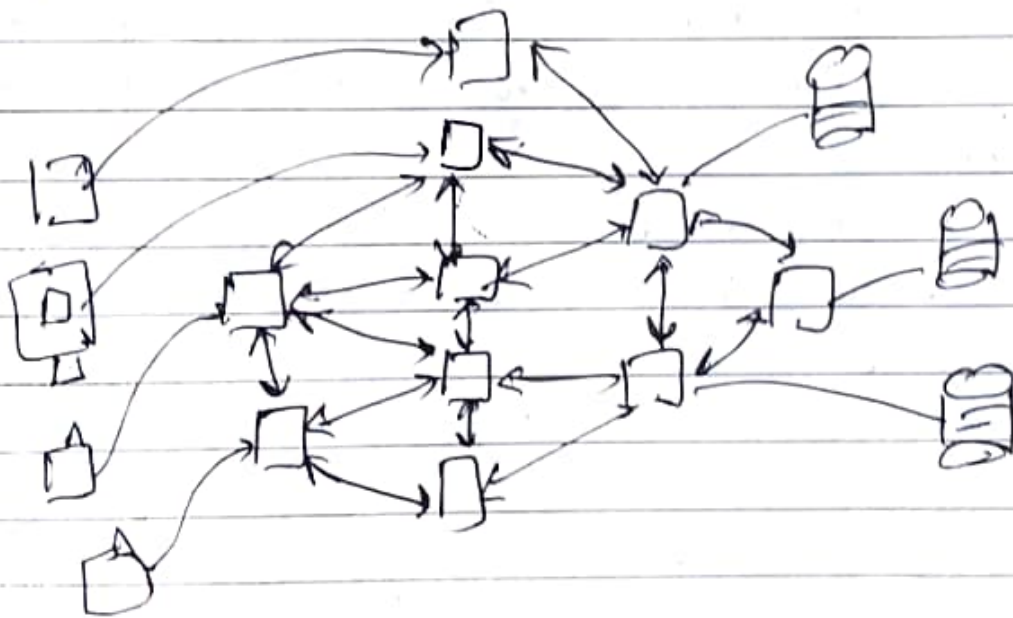
- Microservice architecture is easy to design for complicated system
- Allow new members to train for shorter periods & have less context before touching a system
- Deployments are fluid & continuous for each service.
- Allow decoupling service logic on the basis of business responsibility
- They are more available as a ^{single} service having a bug doesn't bring down the entire system. This is called single point of failure
- Individual services can be written in diff languages
- Developer teams can talk to each other through API sheets instead of working on the same repository, which requires the conflict resolution
- New service can be tested easily & individually, Testing structure is close to unit testing compared to monolith.

• Disadvantage (Monolith architecture favorable)

- Technical/developer team is very small
- Service is simple to think of as a whole.
- Service requires very high efficiency, where network are avoided as much as possible
- All developers must have context of all services

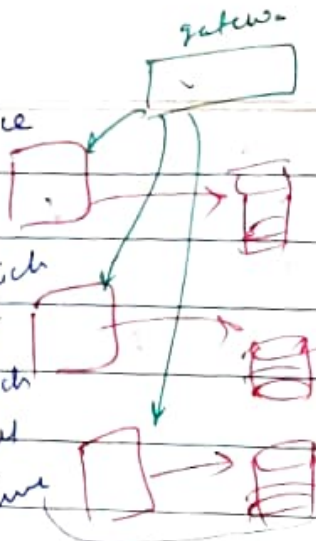


- Monolith doesn't need to be single in case of no. of m/c we are running bcoz there might be multiple m/c to the same monolith & these client can connect to them & they can connect to DB we can horizontally scale out even with monolith



- Microservices is a single business unit, all data, all fn which are relevant to a service are put into one service. If we can separate it with lot of concerns being separated out then we probably want to separate single service into pieces

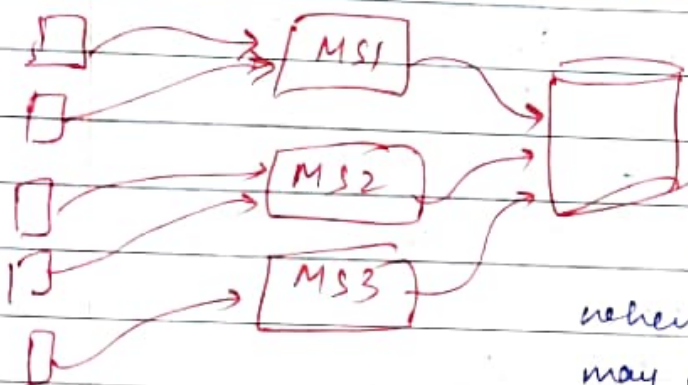
microservice is not a tiny m/c which interact with each other all the time



There may be 3 microservices in an entire architecture depending on what our system is. & they usually talk to their own DB.

client may ~~not~~ be talking to microservices they might be talking to gateway. & this gateway might be talking to all the microservices

• ADVANTAGE OF MONOLITH



① Good for small team.
↳ when put under lot of load, monolith architecture scales out

when we have small teams we may not afford time & efforts to break them into small microservices. If our team is cohesive go for monolith architecture.

② Less complex

↳ lesser moving part around this architecture, we don't need to wonder that how we break into pieces & once we break into pieces how we maintain diff. servers at diff times. Deployment is easy as everything is same

③ Less Duplication for every service we create

↳ we might have some code which are for setting up

④ This is faster (Procedure call is faster)

↳ bcoz we are not making any call to network
It's not the RPC (Remote procedure call). It's
actually in the same box. we just need to make the
local call.

DISADVANTAGES

More context required

① If there are new members in team, they need to
have lot of context on what they are developing
to give them monolith which contains all
the logic they have to go through and
understands the whole system.

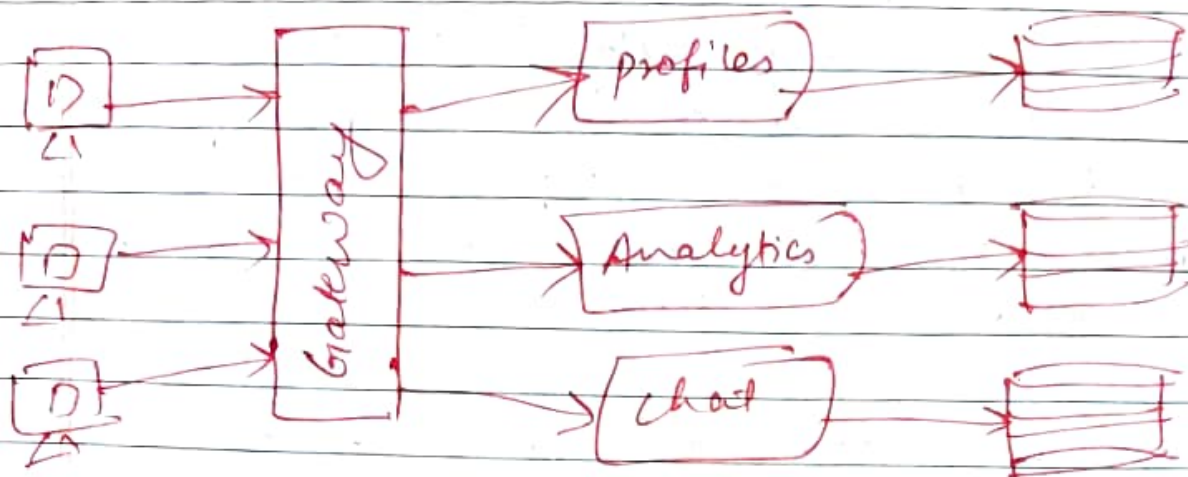
② Complicated Deployment

↳ bcoz any change in the code requires a
new deployment. Here code is going to be
deployed very frequently & it has to be monitored
everytime that it is being deployed.

③ Single point of failure

↳ There is too much responsibility on each server.
If there is a mistake bcoz of which server crashes
all of these server will crash & whole system
will collapse. ~~Instead if we have something~~
~~called server profiles~~

Microservices



① Scalability

↳ as we can look entire set of data as a set of services. Each concerns with only its data & interacting with each other, so it's easier to design the system in that way

② Easier for new team member

↳ whenever there is new developer coming in the team, we can assign them the task which ~~can~~ concerns to the particular service, so they need to know the context of 1 service for eg: chat service instead of entire ~~service~~ monolith

③ Working in Parallel

↳ parallel development is easy bcoz there is lesser dependency for the chat developers on the analytics developers now bcoz they can develop at the same time in the monolith, maybe one ~~for~~ is calling the other for & its changing. so there is lot of tight coupling not just in code, but also in developer time

④ Easier to Scale out

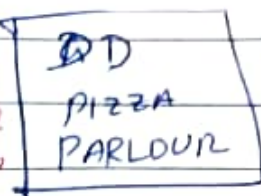
- ↳ There are lesser part which are hidden, when we are actually deploying this service. If there is lot of load on chat server, we can easily scale that out by putting more m/c for that chat code, with monoliths, it's more difficult to save what is being used a lot & what is being useless. So we are more likely to put more server directly instead here we can have more streamlined approach for the problem.

Disadvantage of Microservices

- ↳ Not easy to design so that are being broken into far more parts which are not required.
- ↳ Needs small architect to architect well for a microservice architecture

4 How to Start with Distributed System

① optimise process & increase throughput using ^{vertical} ^{scaling} same resource



1 chef
+
1 backup
chef

② preparing before hand during non peak hours

Processing & examples

③ keep backup & avoid single points of failure

Backup

④ Have more resources i.e. more m/c

Horizontal Scaling

⑤ Microservice architecture

⑥ Distributed system (partitions)

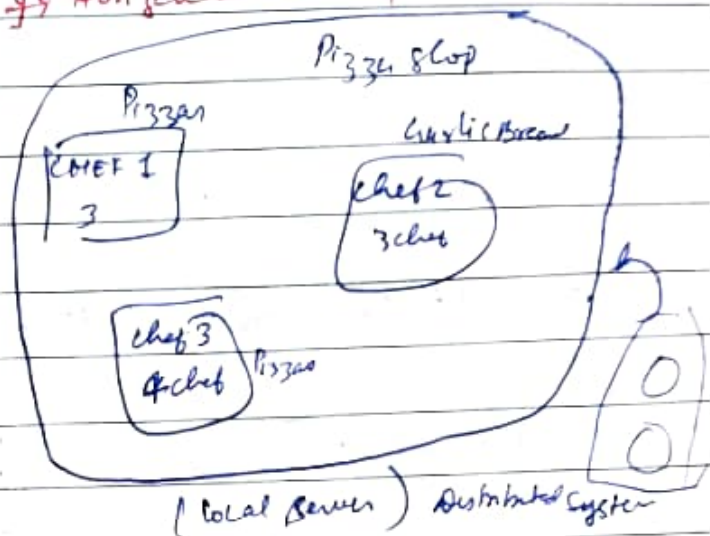
⑧ decoupling

↳ Separation of responsibilities

⑨ logging & metrics calculation

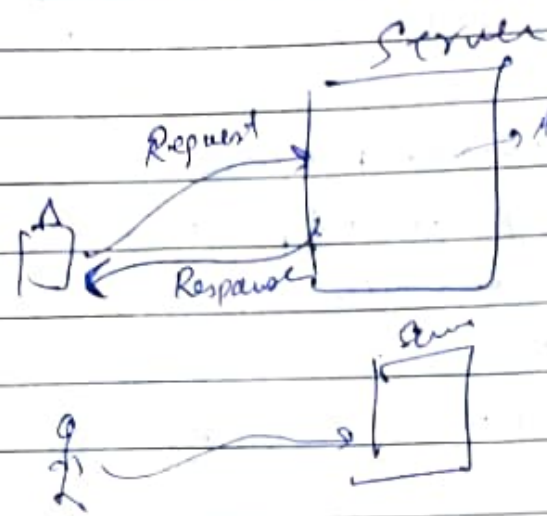
⑩ Extensible

⑦ Load Balances



5 What is Load Balancing

2 WHAT IS LOAD BALANCING



Algorithm runs by understanding the algorithm & sends the response

we have N servers & we have to balance the load ^{evenly} on each server. Concept of this balancing is known as load balancing.

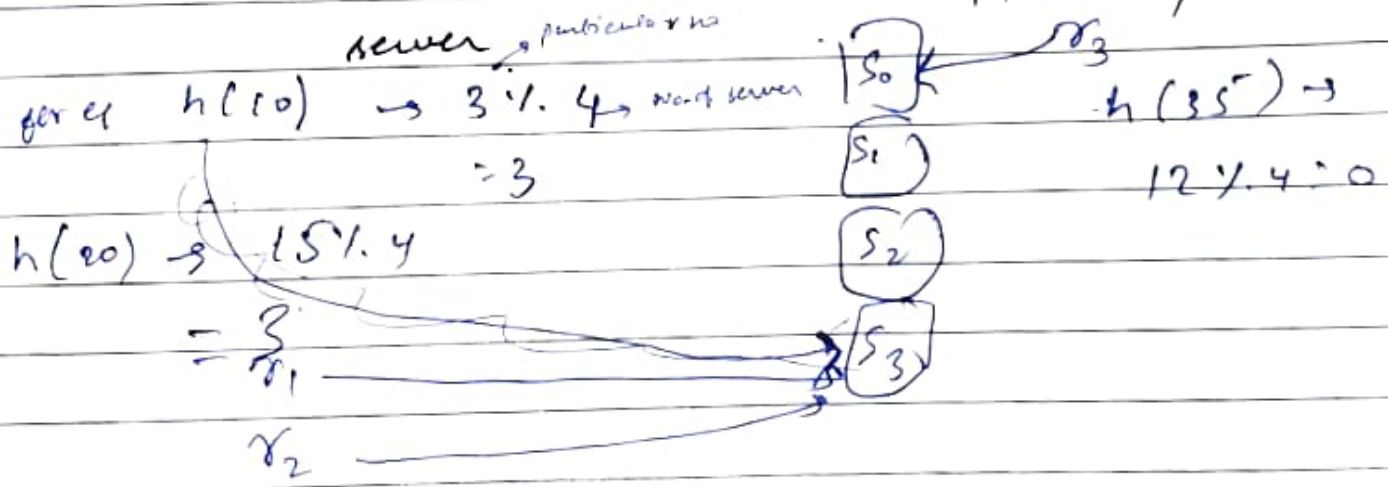
Concept of Consistent Hashing helps us to do that -

- Request ID is generated from each server.

Let we have Request ID $\rightarrow 0$ to $M-1$
 \rightarrow take this request id & hash it, we get particular

$$h(x_i) \rightarrow [m, 1, n]$$

no. & this no can be mapped to particular server _{particular no}



X request n load i.e X/n load & load factor is $1/n$

fault tolerance means machine crashes & then server 11
request allocation means request to the server

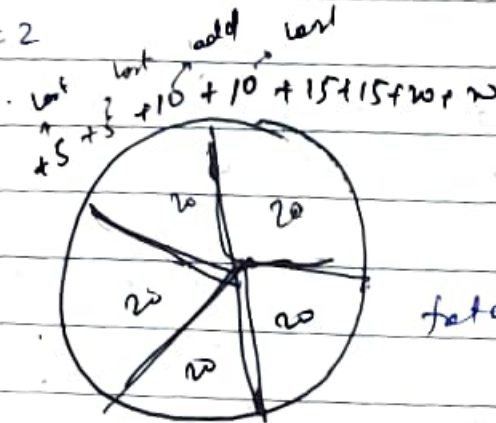
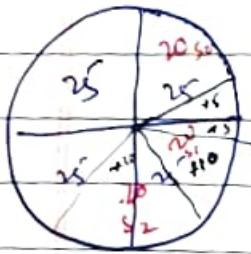
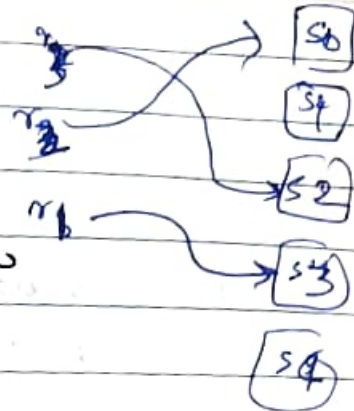
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if we had 4 more server

$$r_1 \rightarrow h(r_1) = 3 \cdot 1.5 = 3$$

$$r_2 \rightarrow h(r_2) = 15 \cdot 1.5 = 0$$

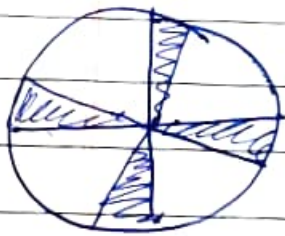
$$r_3 \rightarrow h(r_3) = 12 \cdot 1.5 = 2$$



total change = 100 = 11

and case :- overall change should be min^m

In new pie chart
we will do min^m change in all 4 such that
addition them gives 100.



In practice request ID is so rarely random,
instead it encapsulates some user ID.

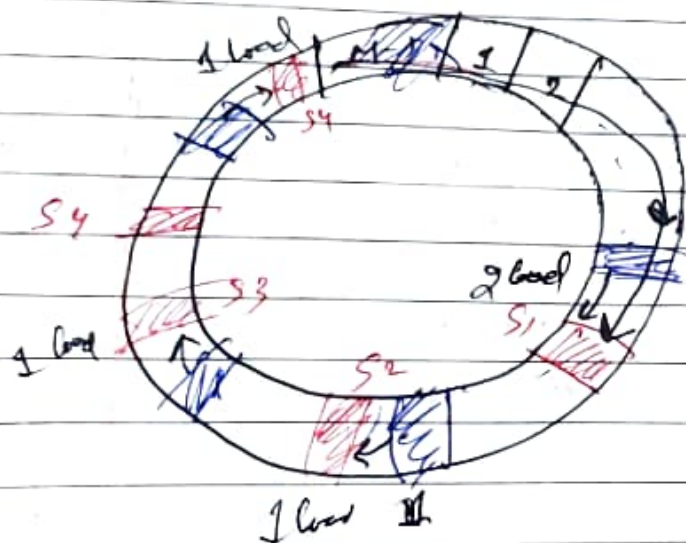
$h(r_i) \rightarrow$ This hash should give same result again & again
Depending on user ID we will send it to specific server
once we send there we will store relevant information
in form of cache but in previous one entire system changes
& all useful information in form of cache is ~~lost~~ & useless
as the no's what we were serving before to server is
completely changed. & we see huge change in range
of servers we are serving.

So old hashing don't work in this case & so we need
more advance cases from where consistent hashing comes

WHAT IS CONSISTENT HASHING 2 WHERE TO USE IT

- Problem is not the load balancing but adding or removing servers that we ~~lose~~ saw, which completely changes the local data that we have

request ID $\rightarrow h(m)$



load factor
expected $= \frac{1}{N}$

It will always go in clockwise
direction & see
the nearest 0 $m-1$
server to it

Instead of hashing
in array we will
do it in ring.

$h(0) \rightarrow 1.M$
 $h(1) \rightarrow 1.M$
 $h(2) \rightarrow 1.M$
 $h(3) \rightarrow 1.M$
 $h(4) \rightarrow 1.M$
 $h(5) \rightarrow 1.M$

on adding new server the change what we see
is less.

If we lost S1 server then all load will go to S2.
problem here is that theoretically load factor is $1/N$ but it is not uniformly distributed but it is skewed distributed.

like here only we will see that half of the load is on S2 which is terrible.

what we can do ?

↳ make multiple hash fn. K hashes fn that we have

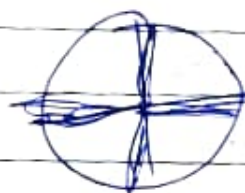
& likelihood of getting 1 K points

server load more is much much lesser

if $K=2$ then 10 on

10 = 3 - 15

If we choose K value appropriately (e.g. $\log W$), we have ~~change~~^{almost} ~~off~~ entirely removing skewed distribution chance.



→ add or remove so this skewed distribution change will be minim^m as equally all server \exists add \exists remove hgh .

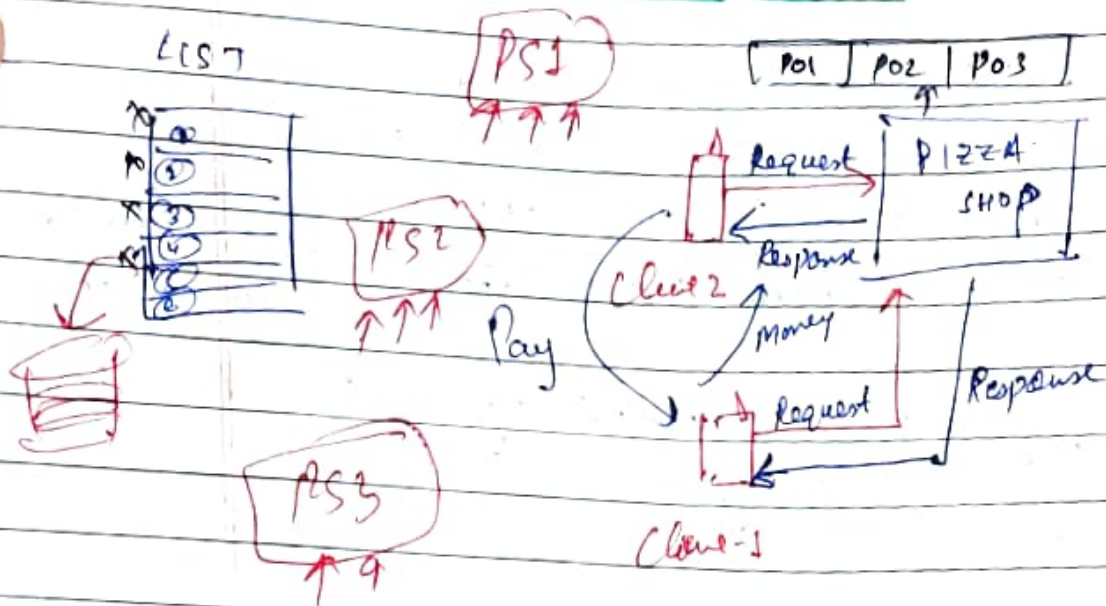
LOAD BALANCING is

Used in distributed system extensively. used by Databases,

consistent Hashing is something that gives flexibility & load balancing in very clear & efficient way. Such that load is balanced & remain close to equal

- MQ are widely used in asynchronous systems. Message processing is an asynchronous fashion allows the client to relieve itself from waiting for a task to complete & hence can do other jobs during that time. It also allows server to process its jobs in the order it wants to.

WHAT IS MESSAGE QUEUE AND WHERE IT IS USED



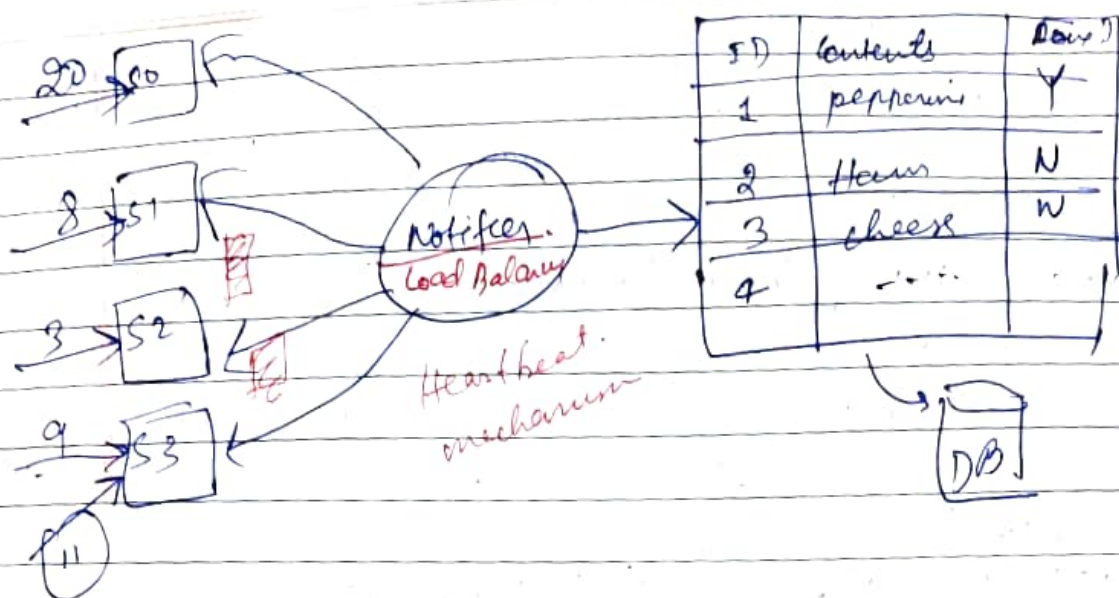
- Pizza bn rha & order n h ad ja rhe aur jaise hi client ko pizza chye use bole pay krne & money wo dea also list of uska order h gya.

→ we can manipulate the queue like some work can be done instant like (eg:- flip coke). This can be done by asynchronous processing.

→ like 3 pizza shop ki outlet h aur ek outlet band ho gya. then the delivery of that order will be fulfilled by other outlet. This cannot be done by list for this we need some persistent in data for which we need DB. & this list will be stored in DB.

consistent Hashing is the technique by which we get rid of Hashing.

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for ex [S3] gets fault. one way is to check DB which orders belongs to S3 & we can note down the server ID which is handling the order everytime when it is making the entry. this is complicated

INSTEAD

Let there be a notifier which is going to check ^{same set of} heartbeat on each server after every 10 sec (ex) asks whether the server is alive if server will not respond ^{notifier} it will ^{assume that server} ~~check~~ it as dead & then it ~~see which~~ all can't handle orders.

↓
then it will query the DB to find all of those orders which are not done. once they are not done it picks those orders & distributes it among remaining servers.

Problem

what if there is duplication & for ex:- order 3 is not yet done & is picked by query that it is not done & get distributed to server 1

System having MQ can move to higher level requirements while abstracting implementation details of message delivery & event handling to the message queue.

Then order 3 will be in both server 1 & server 2.
Then there will be big loss & lots of confusion

We can do

↳ Some sort of load balancing seems like sending right amount of load to each server

But principles of load balancing ensures that you don't have duplicates request to same server

Consistent Hashing

These principles will take care of two things

① Balancing the load

② Not sending duplicates to same server

Reason is that S1 is going to handle some set of buckets
S2 is going to handle some set of bucket.

Once this server crashes S2 won't lose its bucket. It will get new buckets added to it. S1 will also get new bucket added to it & therefore 3rd order will now come bcz they belong to S2 even now & other orders like 9 8 11 come to it

Through load balancing & some sort of heartbeat mechanism we can notify some sort of failed orders to newer server.

• If we want all features of assignment or notification, load balancing, a heartbeat & persistent in one thing. That would be a message queue / Task queue

• 2x as m/c crashes \rightarrow Fault Tolerance

• Scalability \rightarrow m/c need to be added to process more requests

Principles
used in
distributed
system

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what it does it that it takes tasks, persists them, assign them to correct server & waits for them to complete & if it takes too long for server to give an acknowledgement, it feels that server is dead & then assign it to the next server.

Using messaging queue or task queue we get work done easily so that we can encapsulate all the complexity into just one thing.

messaging Queue

eg:- Rabbit MQ, Zero MQ, JMS (Java messaging Service)

MQ are really good encapsulation for complexities in the server side

• Consistent Hashing allows request to be mapped into hash buckets while allowing the system to add & remove nodes flexibly so as to maintain a good load factor on each m/c

• Standard way to hash object is to map them to a search space & then transfer the load to the mapped computer. A system using this policy is likely to suffer when new nodes are added or removed from it.

• Consistent Hashing maps servers to the key space & assigns request (mapped to relevant buckets, called load) to the next clockwise server. Servers can then store relevant data in them while allowing the system flexibility & scalability.

What is Database Sharding

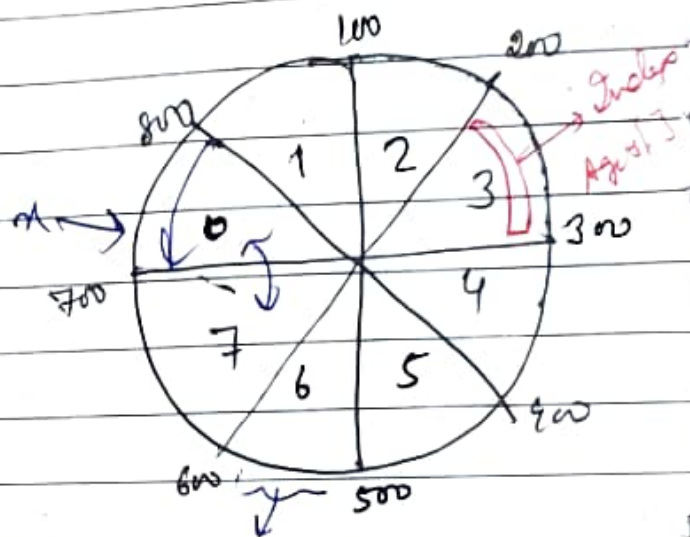
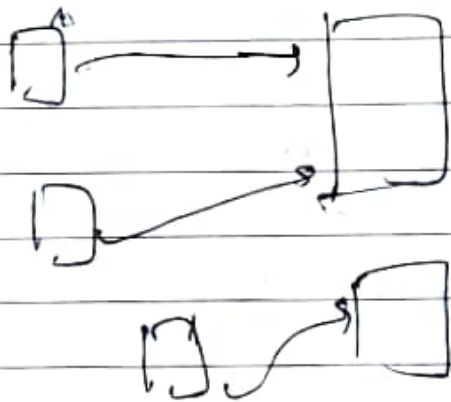
→ Sharding a Database

Scalability strategy used when designing server side systems. It uses concepts like sharding to make system more scalable, reliable & performant

Sharding is horizontal partitioning of data according to

Shard Key:

→ It determines which DB the entry to be persisted is sent to. Some common strategies for this are reverse proxies



Server 6

- partitioning which uses some sort of key to break the data into pieces & allocate them to different servers is called Horizontal partitioning
- Horizontal Partitioning depends upon 3 key which are the attribute of the data which we are storing to partition data
- Vertical Partitioning which uses columns to partition data effectively

• Sharding (Horizontal Partitioning)

↳ Sharding is taking one attribute in the data & partitioning the data such that each ~~data~~^{server} gets one chunk. Server here is DB server

Sharding
↓
Database Server

• SHARDING

→ Consistency

one of the key attribute of any DB that whatever data you persist in it is what we can read out of it later on & there is some sort of synchronization that if a person is going to make update then new request is going to read that update

→ Availability

means DB should not crash & should stay down but we want our application to be running all the time

like if X fall in 7th shard, we just need to read through this shard which is what this DB Server can do.

Advantage

- ↳ shard is going to be smaller in size,
- ↳ easier to maintain
- ↳ faster performance

Disadvantage

① Joins across shard

Query needs to go to two diff. shards. They need to pull out the data then join the data across the network. & this is going to be extremely expensive.

② Shards are inflexible

but we want our DB server to be flexible in numbers. So one of the good algorithm is consistent hashing.

To overcome this problem we do is

take a shard which has too much data in it & then dynamically break them into pieces.

There will be some sort of managers for every particular shard which is going to map the request to the correct mini ~~store~~ pieces in shard ~~also~~ using this hierarchical sharding we can ~~also~~ get rid of inflexibility

One smart thing we will do is to create an index of these shards

assuming that our query requires this index could be completely different attribute compared to user ID

Find all people in } New York lives in server 4. we will
New York 750 } find all users within given range.

All our queries are fast ~~is~~ fast bcoz our read & write performance goes up bcoz all of our queries fall on one particular point.

what happens if shard fails

↳ we have something like master/slave architecture



we have multiple slaves which are copying the master whenever there's a write request it's always on master while slaves continuously pull the master & read from it



If there is a read request it can be distributed across the slaves while if there is write request it goes to master

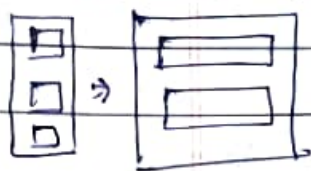


In case master fails the slave choose one master among themselves so there is good single point of failure tolerance over here

Conceptually it's very easy to take the data, break it into ~~pieces~~ ranges & then persist in different places but practically it is tough bcoz consistency is difficult to do.

Two diff ways of increasing the capacity of system or application

Vertical Scaling



involves adding more resource to single m/c

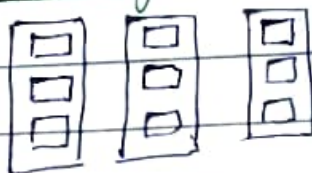
- ↳ adding more resource (such as CPU, RAM & storage) to single server or m/c. (also known as scaling up)

- ↳ eg:- adding up more ~~RAM~~ ^{RAM} to server & its processing power or we might upgrade its CPU to make faster

- ↳ can be quick & easy way to increase the capacity of system, but there is a limit how much a ^{single m/c} system can be scaled

Problem } we can't fit infinitely much CPU power into single m/c

Horizontal Scaling



add more servers & merge data into one database (scaling out)

- ↳ eg:- instead of adding more resource to single server, we might add more servers to cluster or network to distribute the workload which allows system to handle more requests & users at the same time

- ↳ setup is more complex than vertical scaling

- ↳ offers unlimited scalability & high availability by distributing the workload across multiple m/c

- involves adding more m/c to a system
- or application