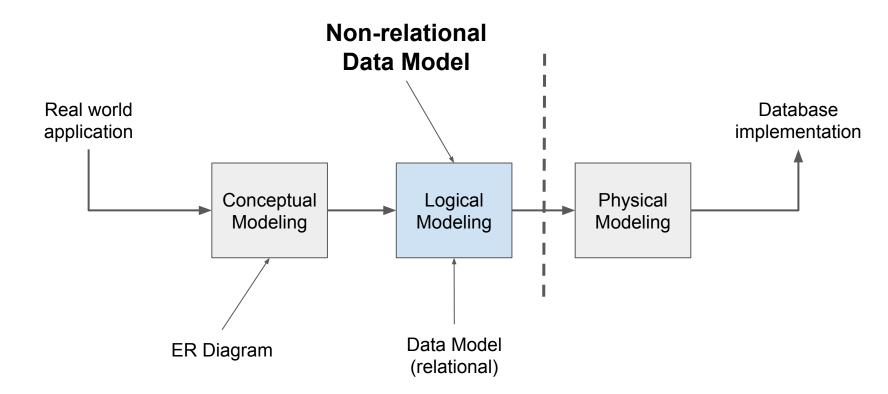
Databases and Big Data

NoSQL

This week



Data Model - Recap



Relational

Key/value

Graph

Document





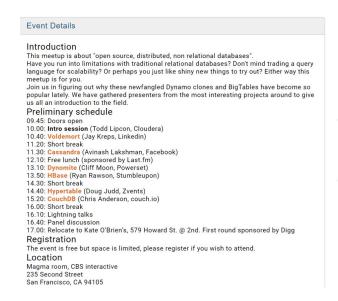






NoSQL

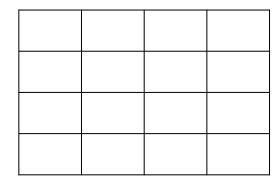
- Started as: no SQL
- Now: Not Only SQL

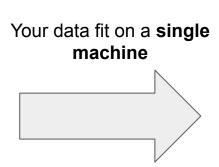


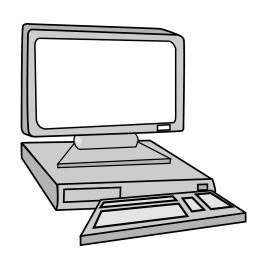


Why NoSQL?

- You have a few tables
- Few GBs max

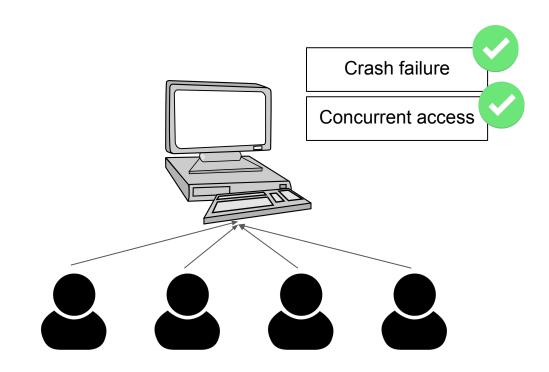






Why NoSQL?

Or a single server



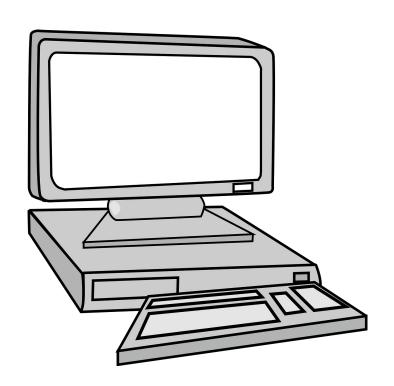
Why NoSQL?

Petabytes do not fit on a single machine



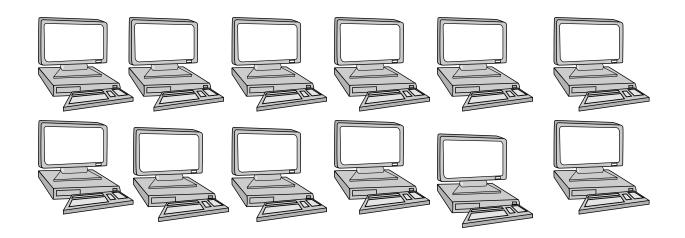
Why NOSQL

- Scale up:
 - Bigger machines
 - Cannot count on it forever
 - Moore Law is slowing down (dead!)



Why NOSQL

- Scale out
 - Many more machines that act as a single machine

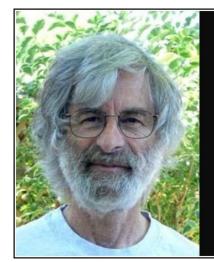


Why NOSQL

- Scale-out is adopted in practice
 - More difficult that it looks.

"You can have a second computer once you've shown you know how to use the first one"

(Paul Barham)



A distributed system is one in which the failure of a computer you didn't even know existed can render your own computer unusable.

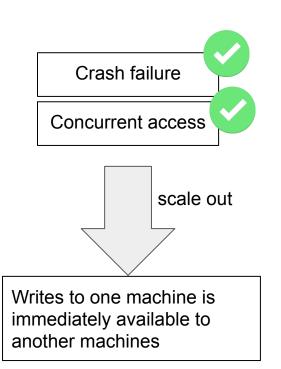
— Leslie Lamport —

AZ QUOTES

Why NoSQL

- <Massive hand waving>
 - Very difficult to implement the Relation Model over many machines
 - Check out 50.041 if interested





Why NoSQL

• Facebook:

- Don't care if I don't see photos in real-time
- Care if I can't upload a photo

Amazon:

- Don't care if my cart forgot an item
- Care if I can't add an item



NOSQL



Relational Database

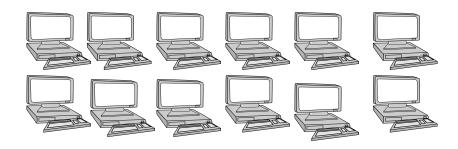
Schema: tables

SQL: join, select, ect.

Scale out

Correctness

Speed*



NoSQL Database

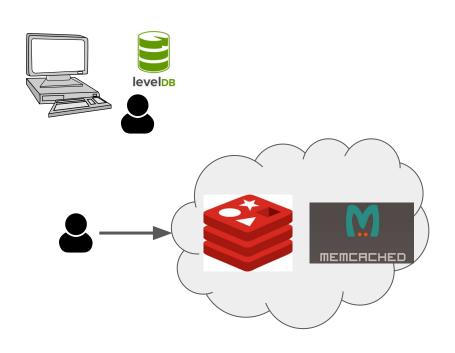
No Schema: just blobs

Simple API: put/get

Not always correct

Scalability

- Absolute opposite of relational database
- Local:
 - o LevelDB
- Remote services
 - Redis
 - Memcached



- Schema
 - \circ key \rightarrow value
 - Like a hash table
- Language:
 - put(key, value)
 - get(key)

K1 AAABBBCCC

K2 | 12340lks 25/11/2019

K3 .=lkj23ois09320-981

K4 abcdo093kjkap

LevelDB

- Backend of Chrome!
- What happen:
 - Put(.) on existing key?
 - Of Get(.) on non-existent keys?
- Demo with ldb tool

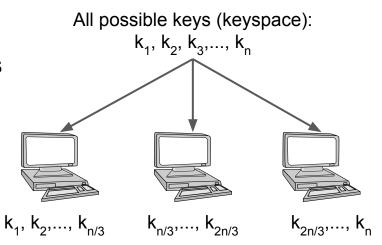




don't need to search through column, like a hashmap

The good

- Very simple to use, behave like a map
- Very handy in many types of applications (Web applications)
- Very fast, like a map
- Very scalable
 - Partition the keys into disjoint sets
 - Then distribute them over many machines



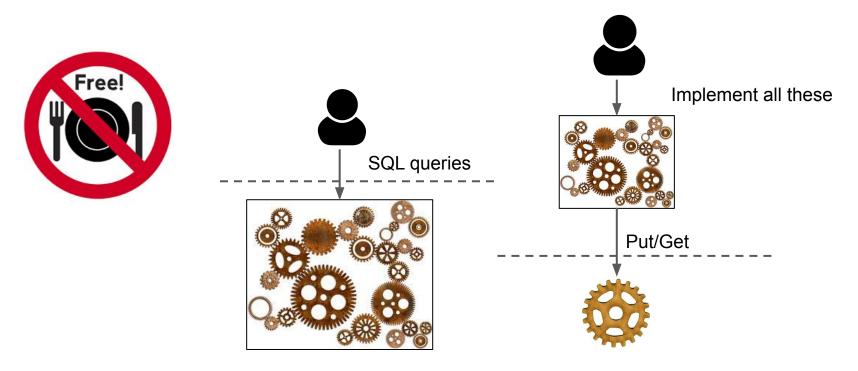
The bad

- You only have Put(.) and Get(.)
- Range queries, like select * from Grade where gpa > 3.5;
- Join queries not supported.

```
for each key in keyspace:
   if (int)key[value] > 3.5:
     print(key)
```

no discrimination on the data types

- The ugly:
 - Treat everything as a binary string (blob, no structure)
 - If your value contains structure, e.g., (age, name, salary)
 - Too bad for you
 - Push complexity to someone else!
 - Price will be paid, question is where, and by whom
 - Joins and range queries are costly

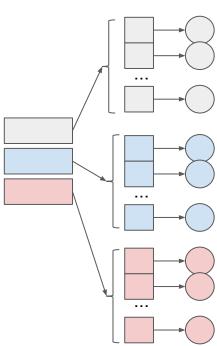


More Structured Key-Value Store

LevelDB:

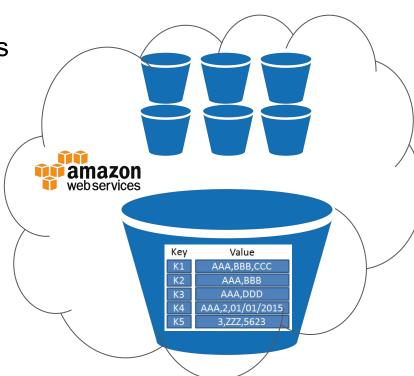
All keys are equal, just like hashmap

- Flat keyspace
- An object is identified by a single key
- Like a typical hashtable
- (A bit) more structure
 - Organize keys into groups
 - An object is identified by (group key, object key)
 - Like a 2-level hash table



S3

- Amazon service
- All data divided in independent buckets
- Each bucket is a key-value store
- put((bucket_key, object_key), value)
- get((bucket_key, object_key))



Loss of 1 object per 10K years



Durability & Data Protection

Q: How durable is Amazon S3?

RocksDB

- Built on LevelDB
 - Used inside Facebook
- Column families
 - Grouping keys in a column family
- Demo with Rocksdb's ldb tool

Column Families provide a way to logically partition the database. Some interesting properties:

- Atomic writes across Column Families are supported. This means you can atomically execute Write({cf1, key1, value1}, {cf2, key2, value2}).
- Consistent view of the database across Column Families.
- Ability to configure different Column Families independently.
- On-the-fly adding new Column Families and dropping them. Both operations are reasonably fast.



- In memory vs. disk
 - All data DRAM
 - Very, very fast



Numbers Everyone Sho	2010 1 1110 11	
L1 cache reference	0	.5 ns
Branch mispredict	5	ns
L2 cache reference	7	ns
Mutex lock/unlock	100	ns
Main memory reference	100	ns
Compress IK bytes with Zippy	10,000	ns
Send 2K bytes over 1 Gbps network	20,000	ns
Read 1 MB sequentially from memory	250,000	ns
Round trip within same datacenter	500,000	ns
Disk seek	10,000,000	ns
Read 1 MB sequentially from network	10,000,000	ns
Read 1 MB sequentially from disk	30,000,000	ns
Send packet CA->Netherlands->CA	150,000,000	ns

More structure

- Value has richer type than just blob
- List
- Set
- Even key-value collection

```
Key
                     Value
List: ["a", "b", "c", "c",...]
Set: {"a", "1", "b"..{
Hash: {"a" => "1",
       "1" => "093",
       "test" = "succeed",
       ...}
```

- Very popular backend for Web applications
 - Because every ms counts
 - Memcached's cousin
 - Too popular that it's in Ubuntu's standard apt repositories



Demo with redis-cli

```
lpush course 12345
lpush course istd_50043
llen course
lindex course 0
```

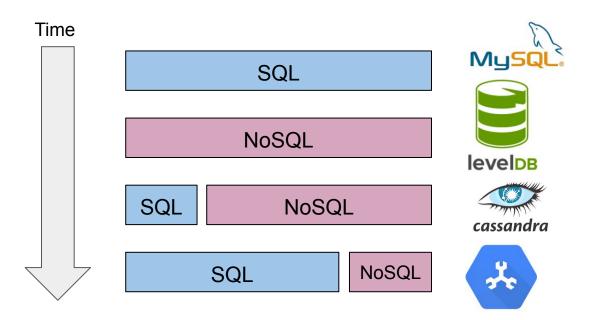
```
hset student age 25
hset student gpa 3.5
hvals student
hkeys student
```

Summary

- NoSQL chooses scalability over everything else
- Key-value store:
 - LevelDB: bare bone
 - S3/RocksDB: column families
 - Redis: in-memory + remote service + typed values.

Final Remark

On History



NoSQL still extremely useful

But people want ease-of-use

 Appreciate the value of SQL more and more