Backend Development– Complement: NoSQL and MongoDB

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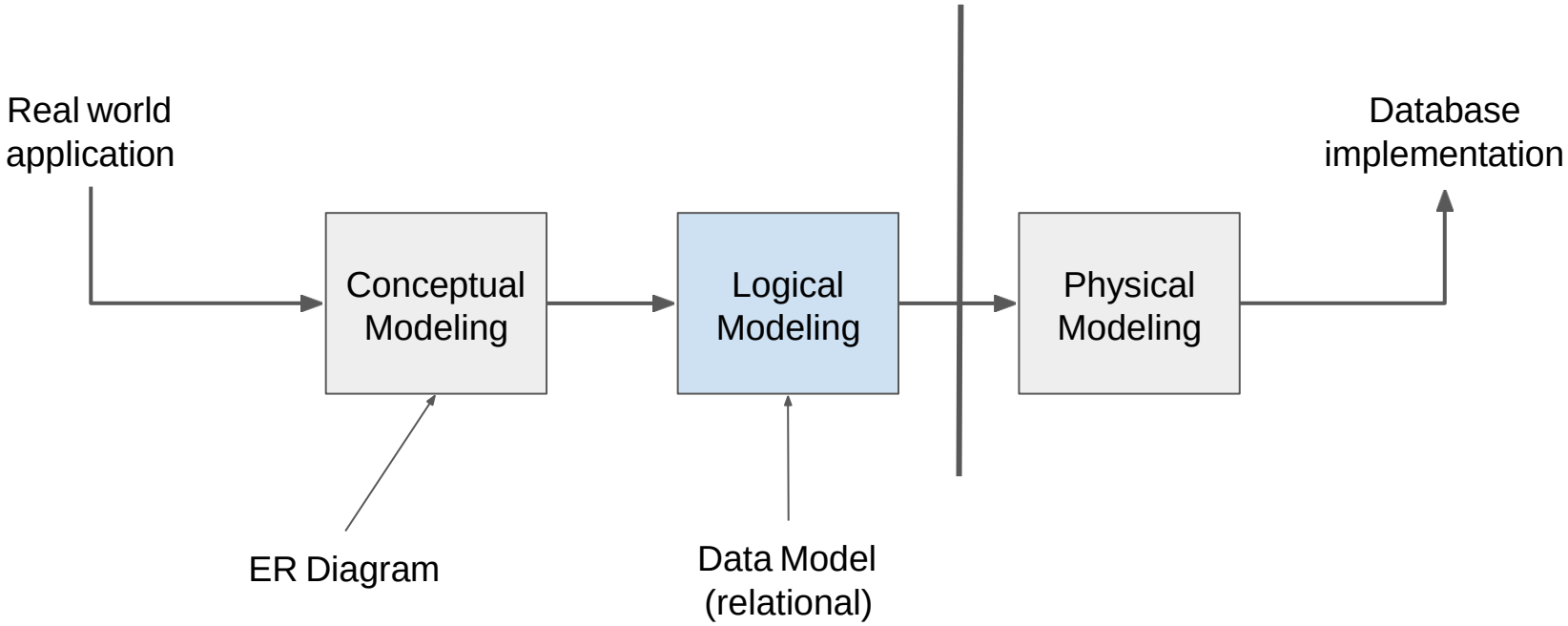
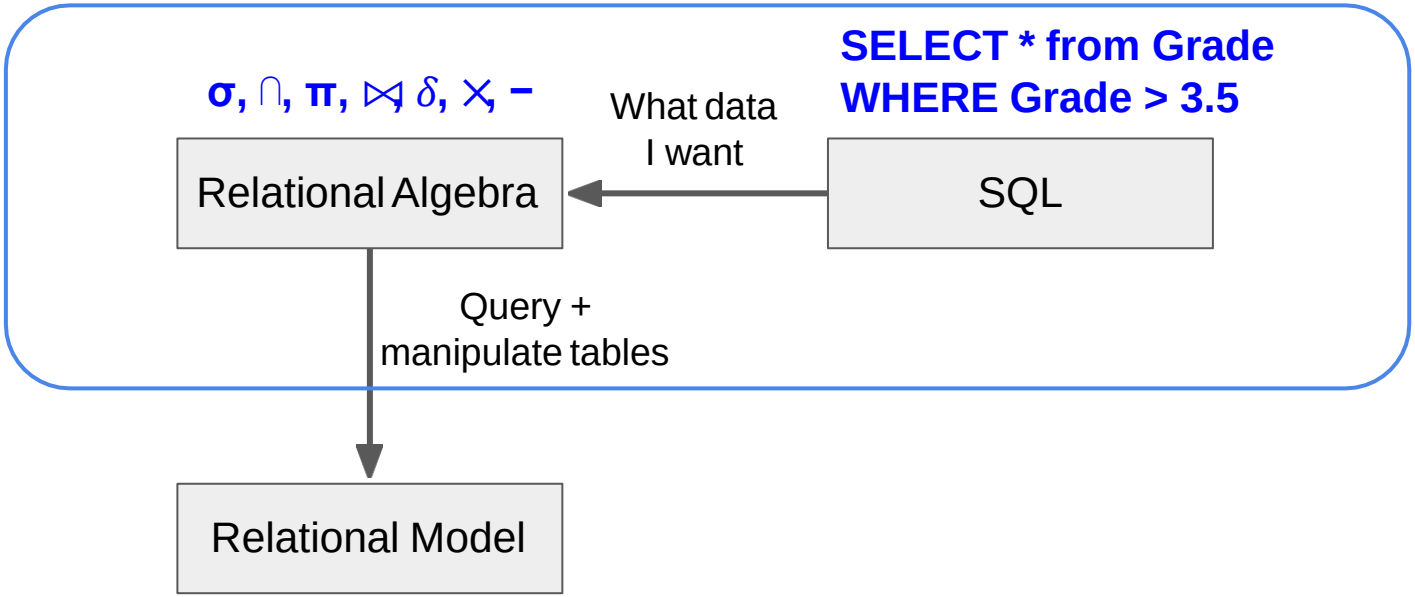
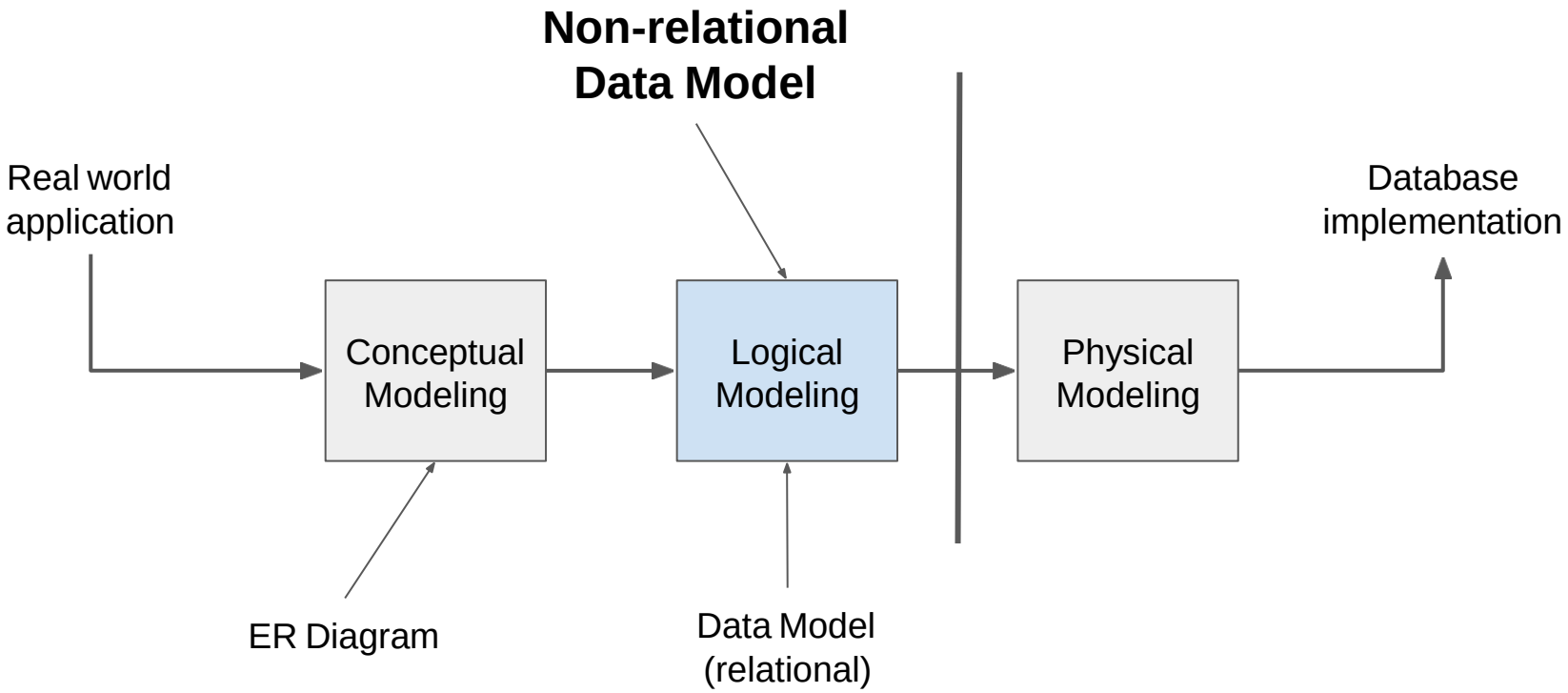
Recap


Real world application
Conceptual Modeling
Logical Modeling
Physical Modeling
Database implementation
ER Diagram
Data Model (relational)

Recap

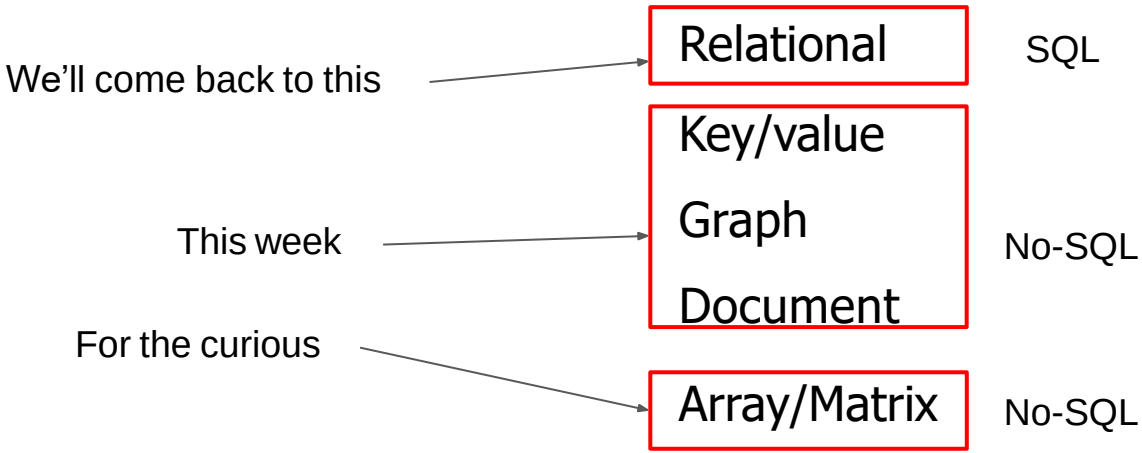

SELECT * from Grade
WHERE Grade > 3.5
σ, ∩, π, ⋈, δ, ×, −
What data
I want
Relational Algebra
SQL
Query +
manipulate tables
Relational Model

In these slides



Data Model - Recap

- Data model:
  - How do you **describe** the data to the database?
  - In a language the database understand



Data Model - Recap

Relational

Key/value

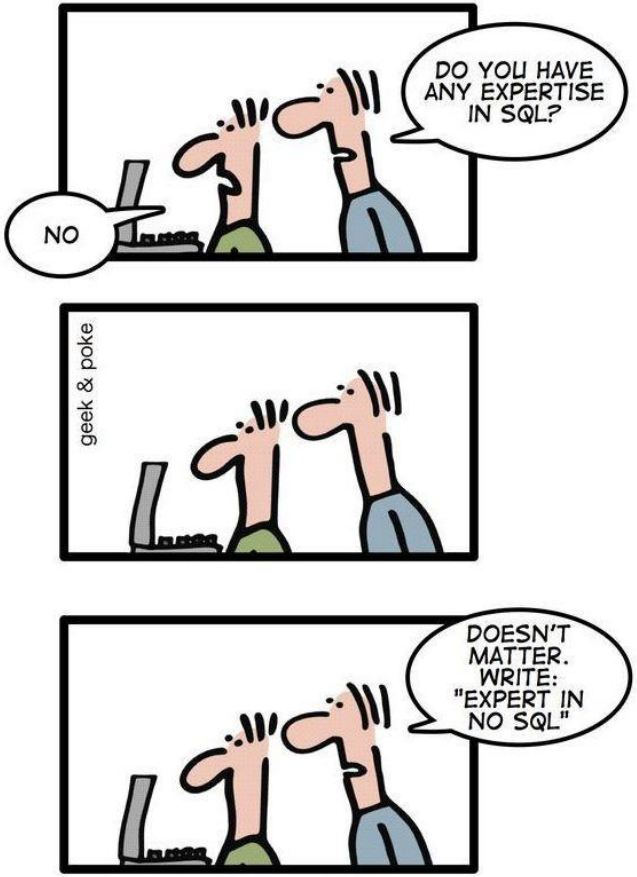
Graph

Document

Array/Matrix

NoSQ

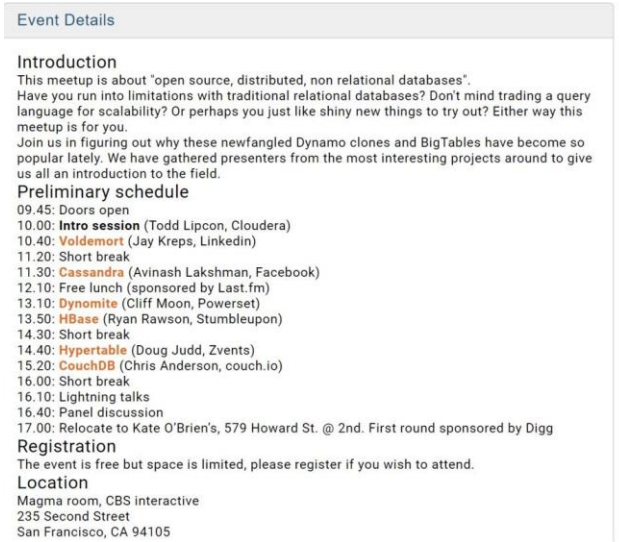
HOW TO WRITE A CV



Leverage the NoSQL boom

NoSQL

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



Event Details

Introduction

This meetup is about "open source, distributed, non relational databases".
Have you run into limitations with traditional relational databases? Don't mind trading a query language for scalability? Or perhaps you just like shiny new things to try out? Either way this meetup is for you.
Join us in figuring out why these newfangled Dynamo clones and BigTables have become so popular lately. We have gathered presenters from the most interesting projects around to give us all an introduction to the field.

Preliminary schedule

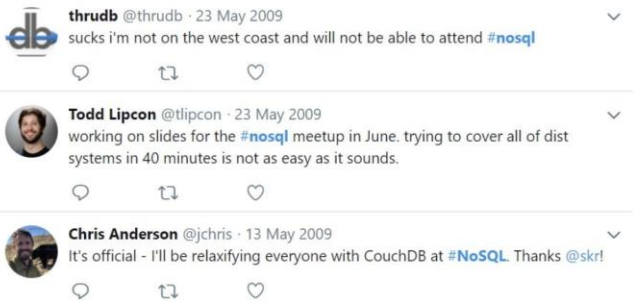
09.45: Doors open
10.00: **Intro session** (Todd Lipcon, Cloudera)
10.40: **Voldemort** (Jay Kreps, Linkedin)
11.20: Short break
11.30: **Cassandra** (Avinash Lakshman, Facebook)
12.10: Free lunch (sponsored by Last.fm)
13.10: **Dynomite** (Cliff Moon, Powerset)
13.50: **HBase** (Ryan Rawson, Stumbleupon)
14.30: Short break
14.40: **Hypertable** (Doug Judd, Zvents)
15.20: **CouchDB** (Chris Anderson, couch.io)
16.00: Short break
16.10: Lightning talks
16.40: Panel discussion
17.00: Relocate to Kate O'Brien's, 579 Howard St. @ 2nd. First round sponsored by Digg

Registration

The event is free but space is limited, please register if you wish to attend.

Location

Magma room, CBS interactive
235 Second Street
San Francisco, CA 94105



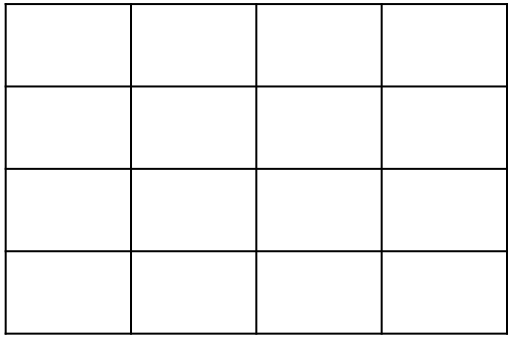
thrudb @thrudb · 23 May 2009
sucks i'm not on the west coast and will not be able to attend #nosql

Todd Lipcon @tlipcon · 23 May 2009
working on slides for the #nosql meetup in June. trying to cover all of dist systems in 40 minutes is not as easy as it sounds.

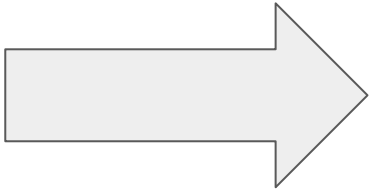
Chris Anderson @jchris · 13 May 2009
It's official - I'll be relaxifying everyone with CouchDB at #NoSQL. Thanks @skr!

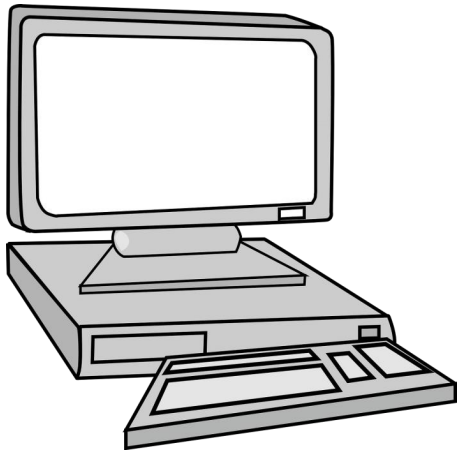
Why NoSQL?

- You have a few tables
- Few GBs max



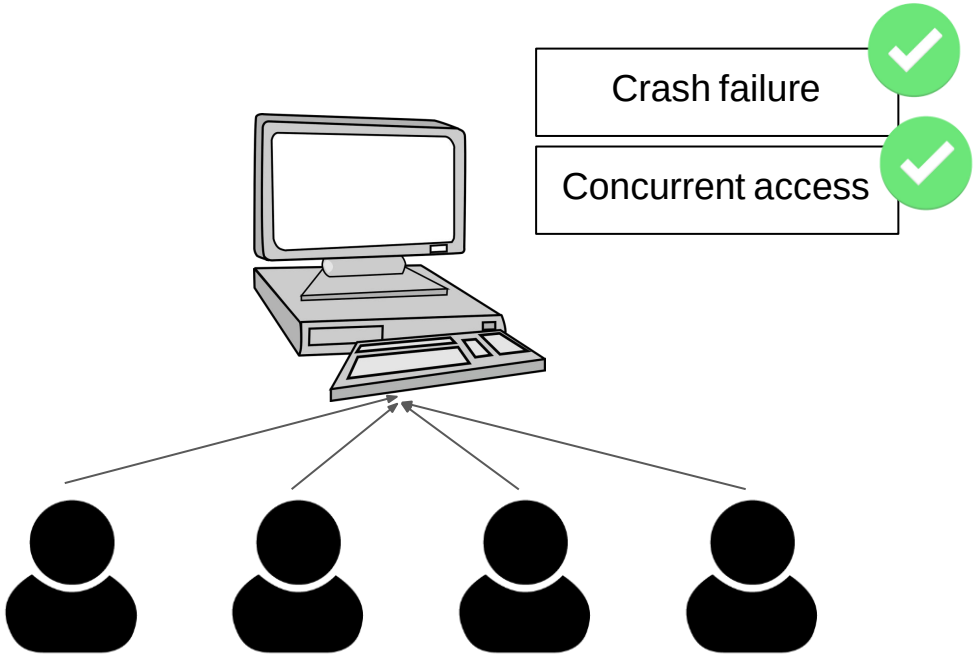
Your data fit on a **single machine**





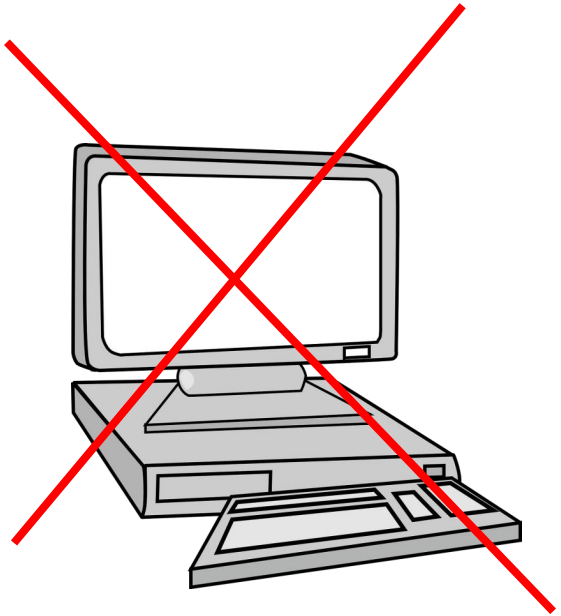
Why NoSQL?

- Or a single server



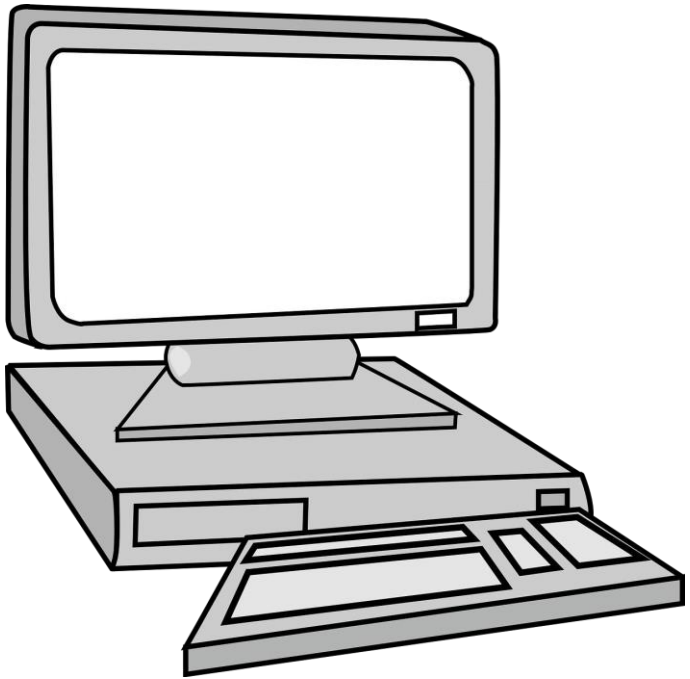
Why NoSQL?

Petabytes do not fit on a single machine



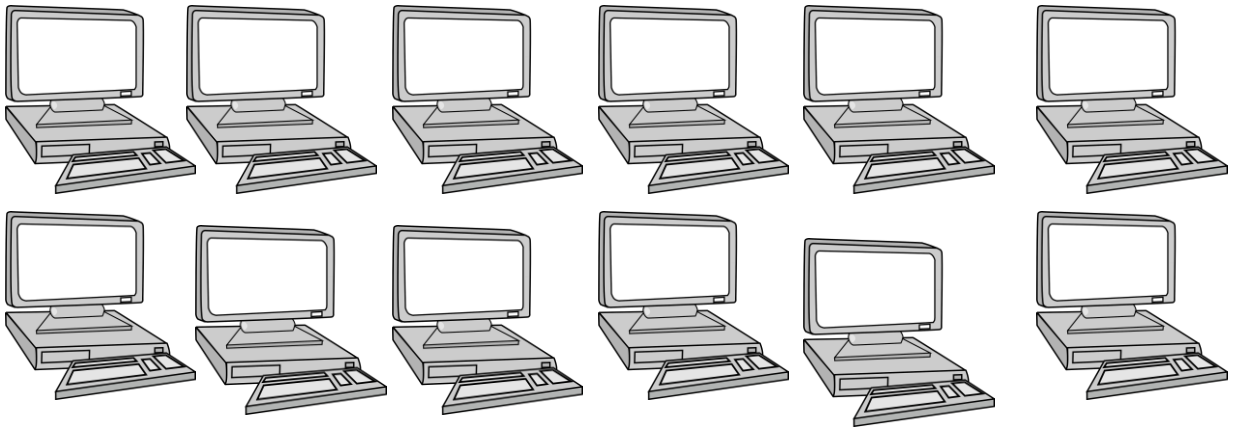
Why NOSQL

- 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

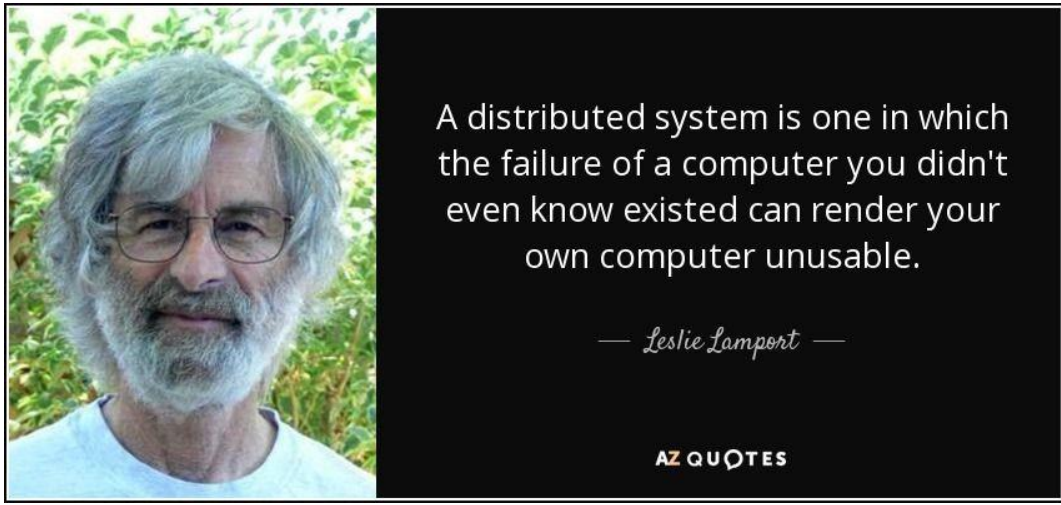


Why NOSQL

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

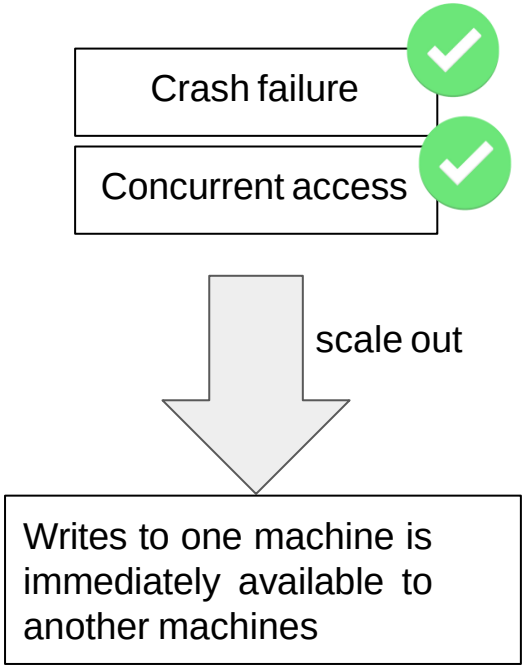
“You can have a second computer **once you’ve shown you know how to use the first one**”

(Paul Barham)



Why NoSQL

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

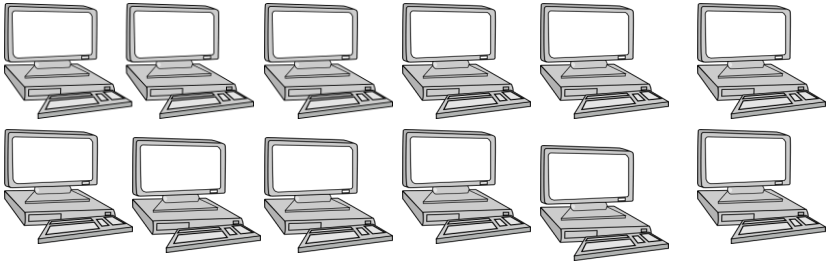


Why NoSQL

- 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

Big companies are obsessed about this

NOSQL



Relational Database

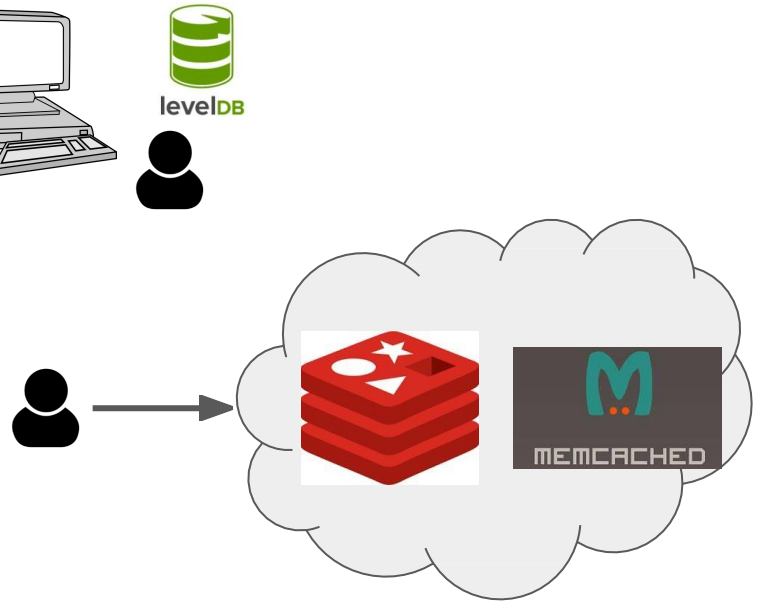
NoSQL Database

Schema: tables	Scale out	No Schema: just blobs
SQL: join, select, ect.		Simple API: put/get
Correctness		Not always correct
Speed*		Scalability

Key-Value Store

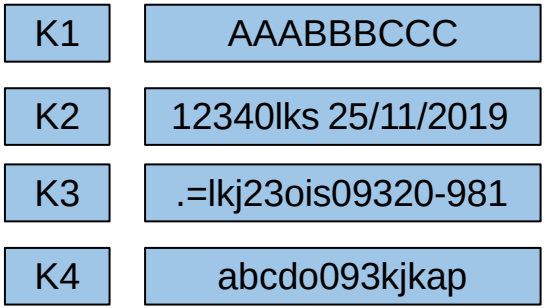
Key-Value Store

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



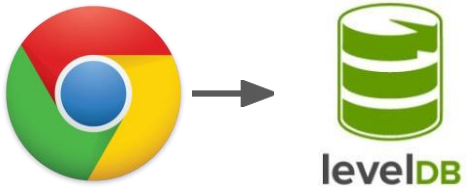
Key-Value Store

- Schema
  - key → value
  - Like a hash table
- Language:
  - put(key, value)
  - get(key)



LevelDB

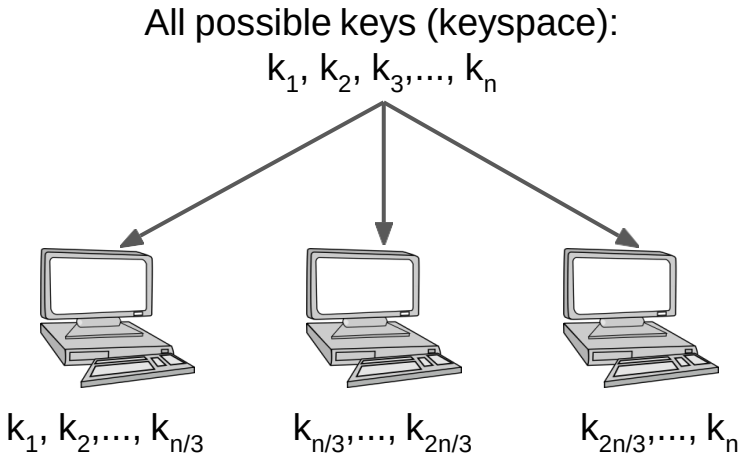
- Backend of Chrome!
- What happen:
  - Put(.) on existing key?
  - Get(.) on non-existent keys?
- Demo with ldb tool




```
%./ldb db
>ls
apple1   apple2   apple3   apple4   apple5   apple6   apple7   apple8   apple9   foobar1  foobar2
foobar3  foobar4  foobar5  foobar6  foobar7  foobar8  foobar9  quxx     zebra1   zebra2   zebra3
zebra4   zebra5   zebra6   zebra7   zebra8   zebra9
>
zebra1   zebra2   zebra3   zebra4   zebra5   zebra6   zebra7   zebra8   zebra9
>get zebra1
```

Key-Value Store

- 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



Key-Value Store

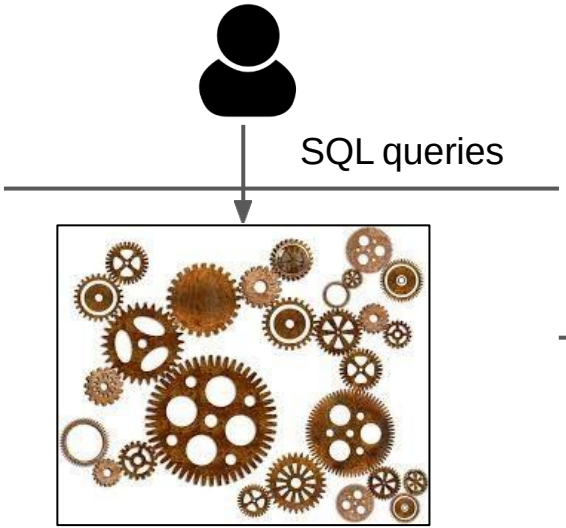
- The bad
  - You only have Put(.) and Get(.)
  - Range queries, like `select * from Grade where gpa > 3.5;`
  - Join queries not supported.

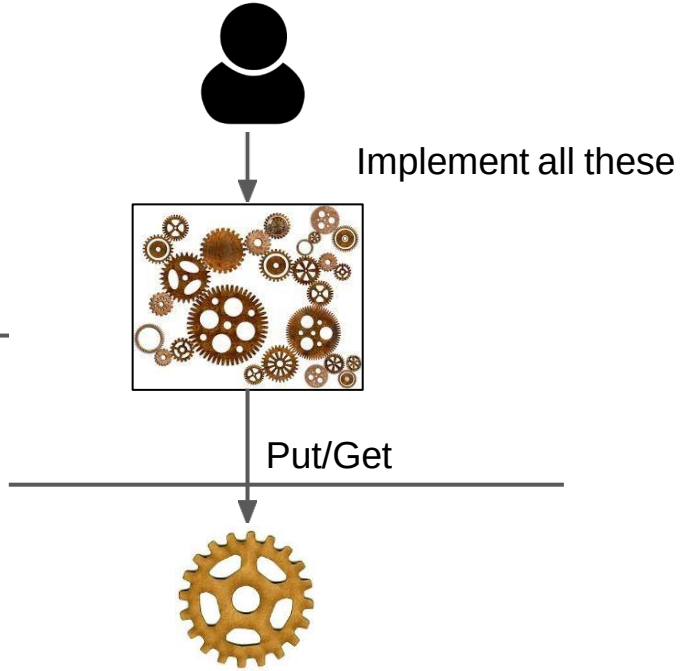
```
for each key in keyspace:
  if (int)key[value] > 3.5:
    print(key)
```

Key-Value Store

- 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

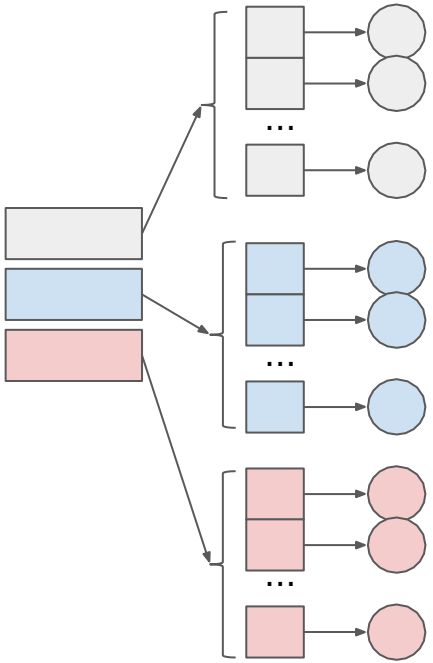
Key-Value Store





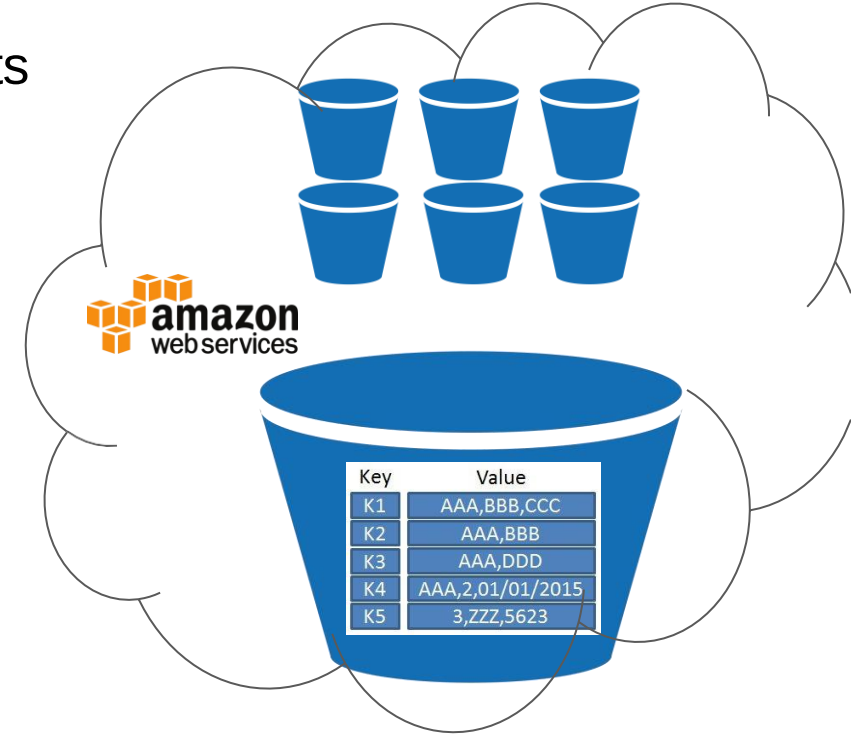
More Structured Key-Value Store

- LevelDB:
  - 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



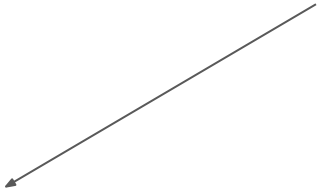
S3

- 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))`



S3

Loss of 1 object per 10K years



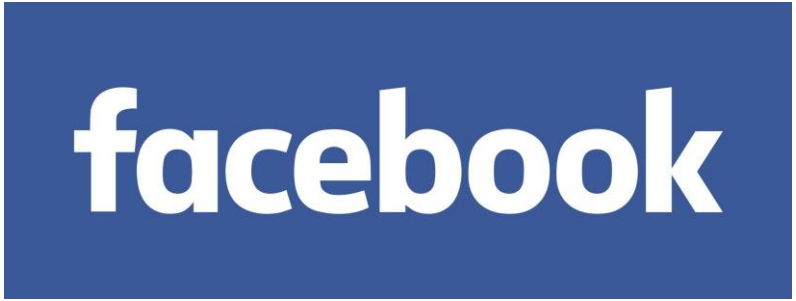
Durability & Data Protection

Q: How durable is Amazon S3?

Amazon S3 Standard, S3 Standard–IA, S3 One Zone-IA, and S3 Glacier are all designed to provide 99.999999999% durability of objects over a given year. This durability level corresponds to an average annual expected loss of 0.000000001% of objects. For example, if you store 10,000,000 objects with Amazon S3, you can on average expect to incur a loss of a single object once every 10,000 years. In addition, Amazon S3 Standard, S3 Standard-IA, and S3 Glacier

RocksDB

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



Redis

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

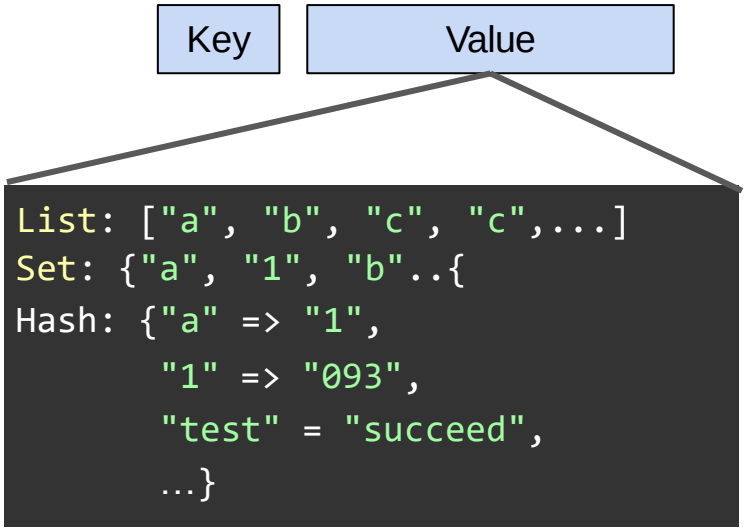
Numbers Everyone Should Know

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 1K 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

Google

Redis

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



Redis

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

Redis

- Demo with redis-cli

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

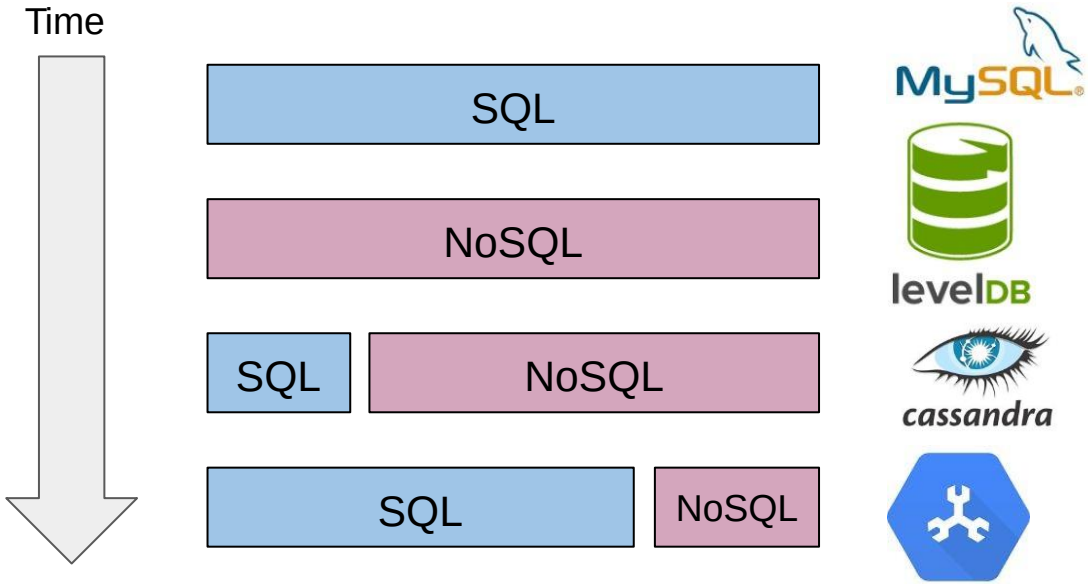
```
hset student age 25
hset student gpa 3.5
hvals student
hkeys student
```

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

MongoDB

A database that everybody ***loves to hate***

mongoDB.

But ends up using it anyway

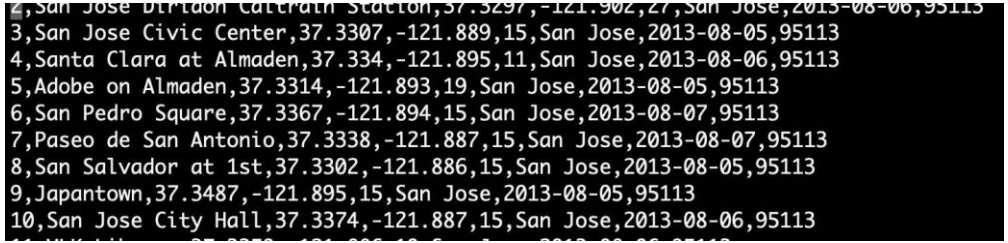
- Most vocal proponent of the NoSQL movement
  - Among the earliest (2009)
- But the database community weren’t impressed

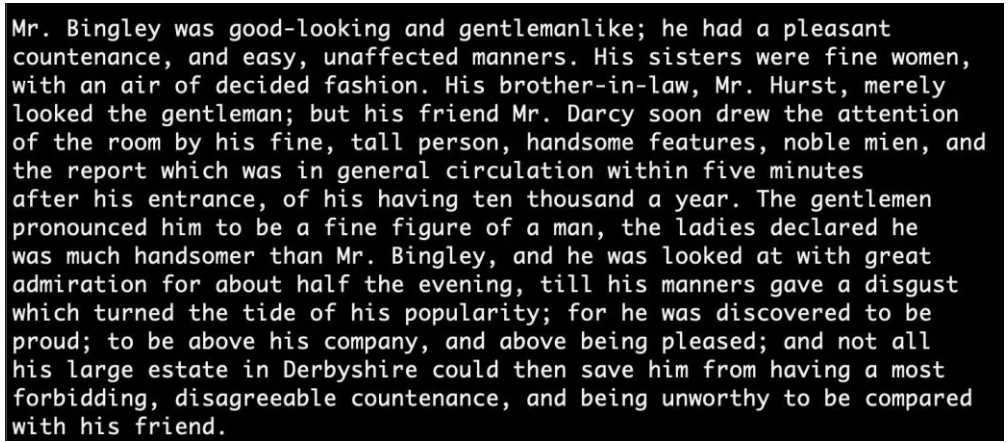
MongoDB

- Regardless, still the most popular NoSQL

Document Store

- What is a document?
  - Loose term
  - Anything that is *parse-able*
  - Examples: csv, text, etc.

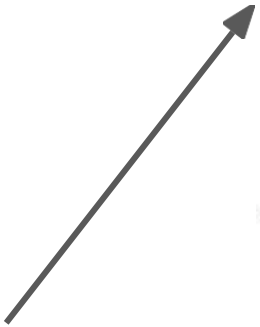




Document Store

- We want semi-structure document:
  - Self-explaining documents
  - Use tags to capture semantics
  - Examples:
    - XML, JSON
    - ProtoBuffer

```
<?xml version="1.0" standalone="yes"?>
<BankAccount>
      <Number>1234</Number>
      <Type>Checking</Type>
      <OpenDate>11/04/1974</OpenDate>
      <Balance>25382.20</Balance>
      <AccountHolder>
            <LastName>Singh</LastName>
            <FirstName>Darshan</FirstName>
      </AccountHolder>
</BankAccount>
```



Tell ***us*** what the enclosed data means

Document Store

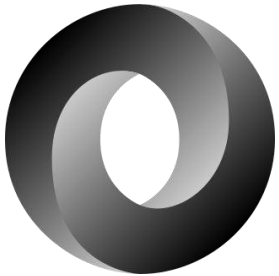
- Semi-structured documents


```
message Block {
    uint32 version = 1;
    google.protobuf.Timestamp timestamp = 2;
    repeated Transaction transactions = 3;
    bytes stateHash = 4;
    bytes previousBlockHash = 5;
    bytes consensusMetadata = 6;
    NonHashData nonHashData = 7;
}

// Contains information about the blockchain le
// block hash, and previous block hash.
message BlockchainInfo {

    uint64 height = 1;
    bytes currentBlockHash = 2;
    bytes previousBlockHash = 3;

}
```

Document Store

- Semi-structured documents

JavaScript Object Notation (JSON)

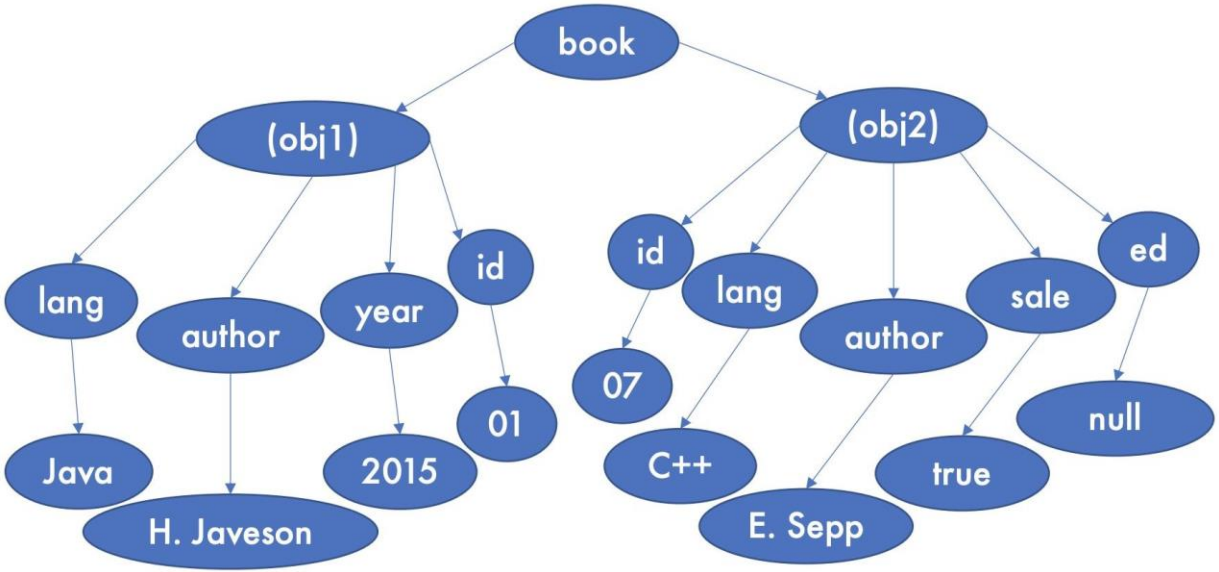


Many applications phasing out XML in favor of JSON

```
{
    "orders": [
        {
            "orderno": "748745375",
            "date": "June 30, 2088 1:54:23 AM",
            "trackingno": "TN0039291",
            "custid": "11045",
            "customer": [
                {
                    "custid": "11045",
                    "fname": "Sue",
                    "lname": "Hatfield",
                    "address": "1409 Silver Street",
                    "city": "Ashland",
                    "state": "NE",
                    "zip": "68003"
                }
            ]
        }
    ]
}
```

Semi-Structured Documents

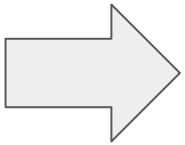
- It's a **tree**



Semi-Structured Documents

- From tables to trees
  - Not hard, because tables = flat trees

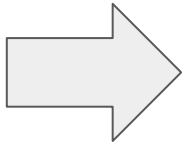
Name	Phone
Anh	12345
Dan	23093
Leo	09470



```
{
  "person":[
      {
        "name": "Anh",
        "phone": "12345"
      },
      {
        "name": "Dan",
        "phone": "23093"
      },
      {
        "name": "Leo",
        "phone": "09470"
      }
  ]
}
```

Semi-Structured Documents

Name	Phone
Anh	12345
Dan	23093
Leo	09470
Ben	NULL



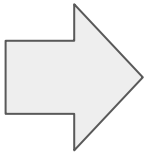
```
{
  "person":[
      {
        "name": "Anh",
        "phone": "12345"
      },
      {
        "name": "Dan",
        "phone": "23093"
      },
      {
        "name": "Leo",
        "phone": "09470"
      },
      {
        "name": "Ben"
      }
  ]
}
```

Field missing = NULL

Semi-Structured Documents

- But can we fit any tree into a table?
- Non-flat data:
  - **Array**
  - Multi-part

Name	Phone
Anh	???
Dan	23093
Leo	09470

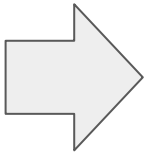


```
{
  "person":[
      {
        "name": "Anh",
      },
      {
        "name": "Dan",
        "phone": "23093"
      },
      {
        "name": "Leo",
        "phone": "09470"
      }
  ]
}
```

Semi-Structured Documents

- But can we fit any tree into a table?
- Non-flat data:
  - Array
  - **Multi-part**

Name	Phone
???	12345
Dan	23093
Leo	09470



```
{
  "person":[
      {
        "phone": "12345"
      },
      {
        "name": "Dan",
        "phone": "23093"
      },
      {
        "name": "Leo",
        "phone": "09470"
      }
  ]
}
```

Semi-Structured Documents

- Relational data model isn’t designed for nested data
  - Tables vs. trees
- Term: **impedance mismatch**

Document Store

- Handle trees
- Many implementations

mongoDB.

Apache CouchDB relax

elasticsearch

AsterixDB™

MongoDB Language

- Remember for SQL, we have:
  - Data Definition Language (DDL): create, delete, etc.
  - Data Manipulation Language (DML): insert, update, etc.
  - Query Language: select … from … where
- MongoDB supports same categories
  - **Tables** in SQL → **Collections** in MongoDB

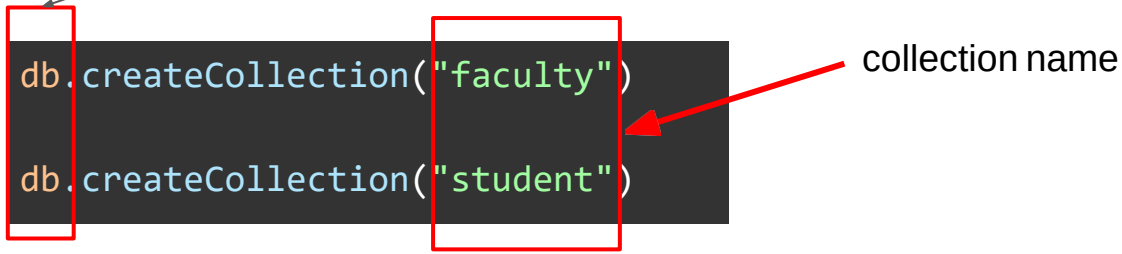
MySQL	**MongoDB**
Database	Database
Tables	Collections


{
"Name": "Anh"
"Phone": [
"12345",
"93932"
]
}
{
"Name": "Dan"
"Phone": "93752"
}
Collection
{
"Name": {
"First": "Albert",
"Last": "Einstein"
},
"Theory": "Particle Physics"
}
{
"Name": {
"First": "Sheldon",
"Last": "Copper"
},
}
Document
Database

MongoDB

- Create table

- Create collections



MongoDB

- Insert new document
  - Duplicates are allowed

```
db.faculty.insert({"Name": "Einstein",
"Theory": "Relativity"})
```

```
db.faculty. update({"Name": "Einstein"},
                   {"$set": {"Country": "Germany"}})
```

Guess what this does

```
{
  "Name": {
      "First": "Albert",
      "Last": "Einstein"
  },
  "Theory": "Particle Physics"
}

{
  "Name": "Einstein",
  "Theory": "Relativity"
}

…
…
```

MongoDB

- Read
  - Select all documents

```
db.faculty.find({})
```

```
db.faculty.find()
```

SELECT * from University;

(SQL CheatSheet)

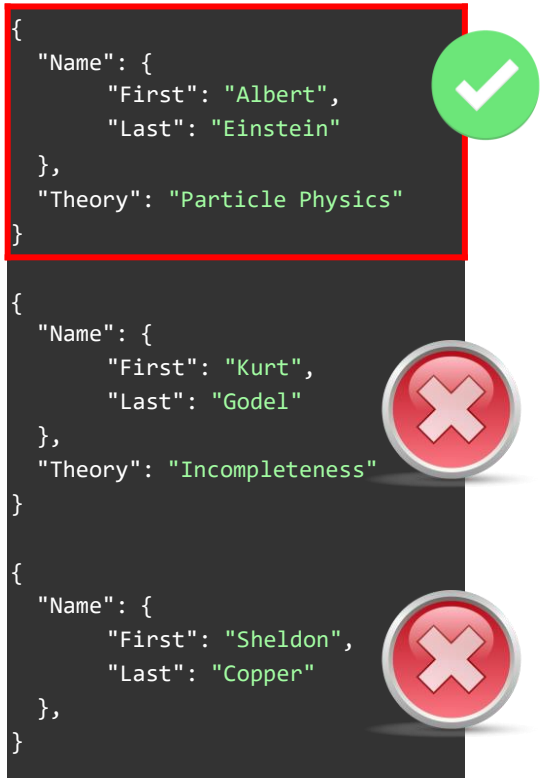
MongoDB

- Read condition

```
db.faculty.find(
      {"Theory": "Particle Physics"}
)
```

SELECT * from Faculty
WHERE Theory = “Particle Physics”

(SQL CheatSheet)



MongoDB

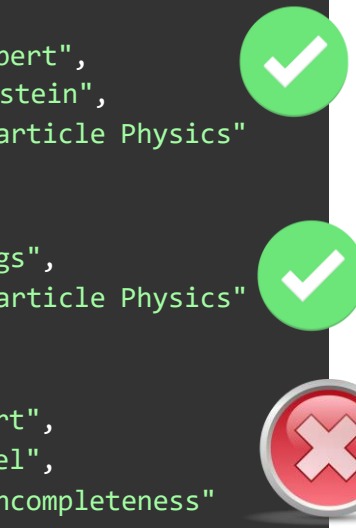
- Read projection

```
db.faculty.find(
      {"Theory": "Particle Physics"},
      {"Name": 1, "Last": 1}
)
```

SELECT Name, Last from Faculty
WHERE Theory = “Particle Physics”

(SQL CheatSheet)

```
{
  "First": "Albert",
  "Last": "Einstein",
  "Theory": "Particle Physics"
}
{
  "Name": "Higgs",
  "Theory": "Particle Physics"
}
{
  "First": "Kurt",
  "Last": "Godel",
  "Theory": "Incompleteness"
}
```

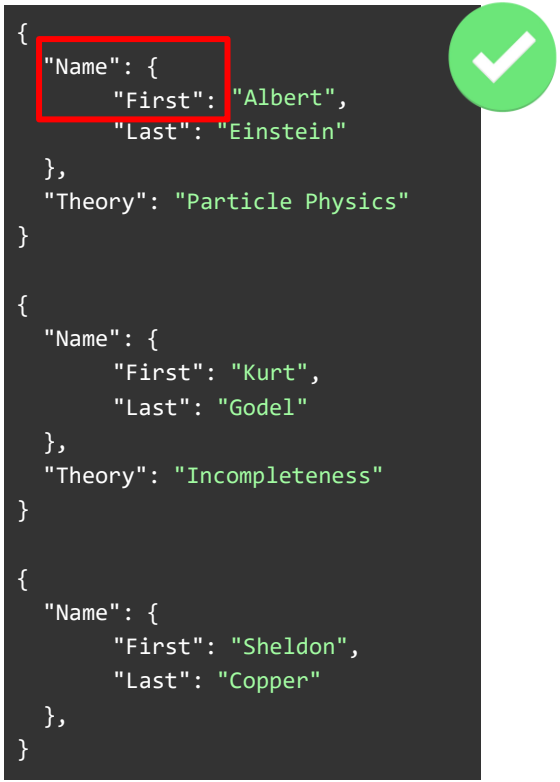


{ “Last”: “Einstein”}
{ “Name”: “Higgs”}

MongoDB

- Read
  - Nested field

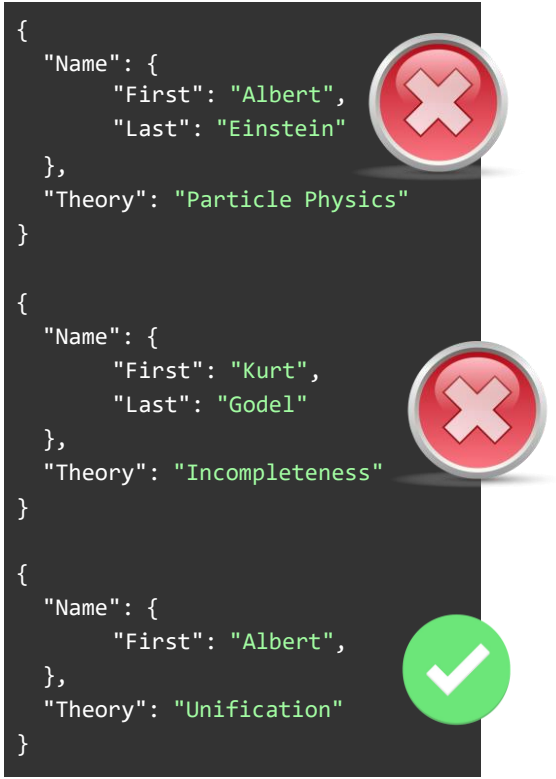
```
db.faculty.find({"Name.First": "Albert"})
```



MongoDB

- Read
  - Nested field

```
db.faculty.find({"Name": {"First": "Albert"}})
```



MongoDB

- Read
  - List matching

```
db.faculty.find({"Theory": “Special relativity”})
```

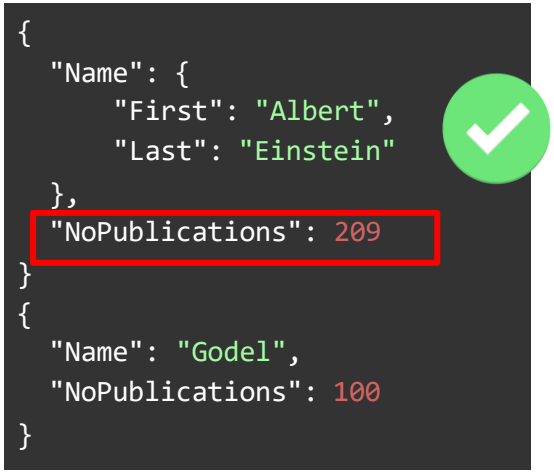
```
{
  "Name": {
      "First": "Albert",
      "Last": "Einstein"
  },
  "Theory": [
      "Special relativity",
      "General relativity"
  ]
}
{
  "Name": "Godel",
  "Theory": "Incompleteness"
}
```

MongoDB

- Read operators

Operator	Meaning
$gte, $eq, $ne, $gt, $lt, $lte	>=, =, !=, >, <, <=
$in, $nin	∈, ∉

```
db.faculty.find( {
      "NoPublications": {"$gte": 120}
})
```



MongoDB

- Read operators

Operator	Meaning
$gte, $eq, $ne, $gt, $lt, $lte	>=, =, !=, >, <, <=
$in, $nin	∈, ∉

```
db.faculty.find( {
      "University": {"$in": ["NUS”, “SUTD"]}
})
```

```
{
  "Name": {
      "First": "Albert",
      "Last": "Einstein"
  },
  "NoPublications": 209
  "University": "SUTD"
}
{
  "Name": "Godel",
  "NoPublications": 100
  "University": "SMU"
}
```

MongoDB

- Read
  - Count
  - Sort
  - Limit
  - Duplicates

```
db.faculty.find({"Theory": "Particle Physics"}).count()
```

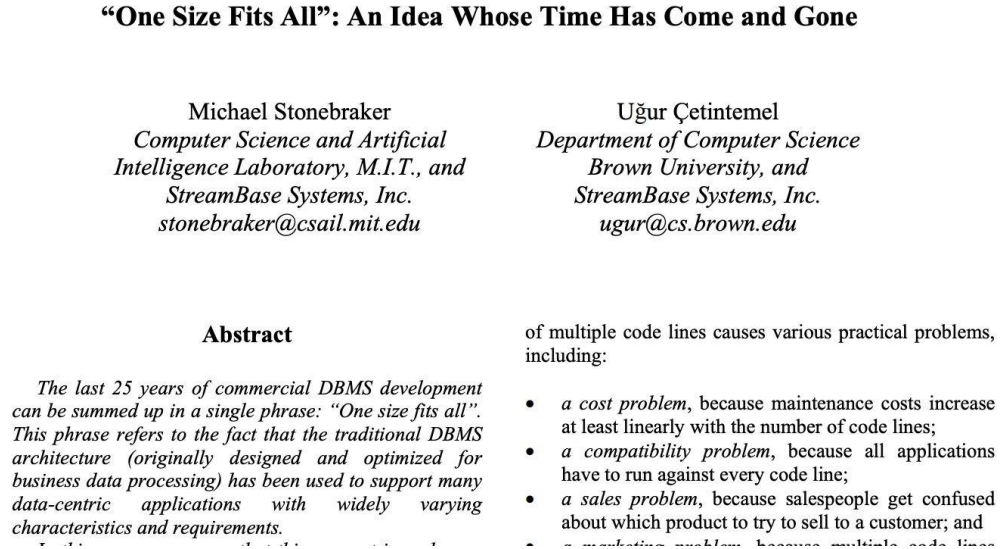
```
db.faculty.find({"Theory": "Particle Physics"}).sort({"Age": -1})
```

```
db.faculty.distinct(“Name”, {"Theory": "Particle Physics"})
```

- Check them out yourself

Summary

- MongoDB most popular for
  - Semi-structured data model
  - Or trees
- One size doesn’t fit all



“One Size Fits All”: An Idea Whose Time Has Come and Gone

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Abstract

The last 25 years of commercial DBMS development can be summed up in a single phrase: “One size fits all”. This phrase refers to the fact that the traditional DBMS architecture (originally designed and optimized for business data processing) has been used to support many data-centric applications with widely varying characteristics and requirements.

of multiple code lines causes various practical problems, including:

- *a cost problem*, because maintenance costs increase at least linearly with the number of code lines;
- *a compatibility problem*, because all applications have to run against every code line;
- *a sales problem*, because salespeople get confused about which product to try to sell to a customer; and