### **Big Data**

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## Beyond Relational Data noSQL databases

- Document DBs
  - MongoDB
- ✓ Graph databases
  - Neo4j

### What is MongoDB?

- Developed by 10gen: "humongous" DB
  - Founded in 2007
- A document-oriented NoSQL database
  - Hash-based, schema-less database
    - No Data Definition Language
    - can store hashes with any keys and values that you choose
      - Keys are a basic data type but in reality stored as strings
      - Document Identifiers (\_id) will be created for each document, field name reserved by system
    - Uses BSON format
      - Based on JSON B stands for Binary
  - Supports APIs (drivers) in many computer languages
    - JavaScript, Python, Ruby, Perl, Java, Java Scala, C#, C++, Haskell, Erlang

### **Document DB**

- Documents are the main concept.
- A Document-oriented database stores and retrieves documents (XML, JSON, BSON and so on).
- ✓ Documents are:
  - Self-describing
  - Hierarchical tree data structures (maps, collection and scalar values)

### What is a Document DB?

- Document databases store documents in the value part of the key-value store where:
  - Documents are indexed using a BTree
  - queried using e.g, JavaScript query engine

```
name: "sue",

age: 26,

status: "A",

groups: [ "news", "sports" ] 

field: value

field: value

field: value
```

### What is a Document DB?

```
{
    "name": "Phil",
    "age": 26,
    "status": "A"
}
    "citiesVisited" : ["Chicago", "LA", "San Francisco"]
}
```

- Documents may have different attributes
- ✓ But belongs to the same collection
- ✓ Different from relational databases where columns:
  - Stores the same type of values
  - Or null

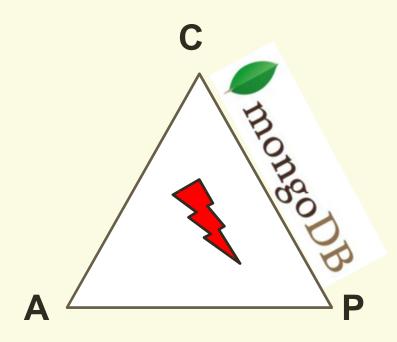
### **MongoDB**

- ✓ MongoDB Features
  - Document-Oriented storage
  - Index Support
  - Replication & Availability
  - Auto-Sharding
  - Ad-hoc Querying
  - Fast In-Place Updates
  - Map/Reduce functionality
- ✓ Application need
  - Simple queries
  - •Functionality provided applicable to most web applications
  - Easy and fast integration of data
  - •Not well suited for heavy and complex transactions systems

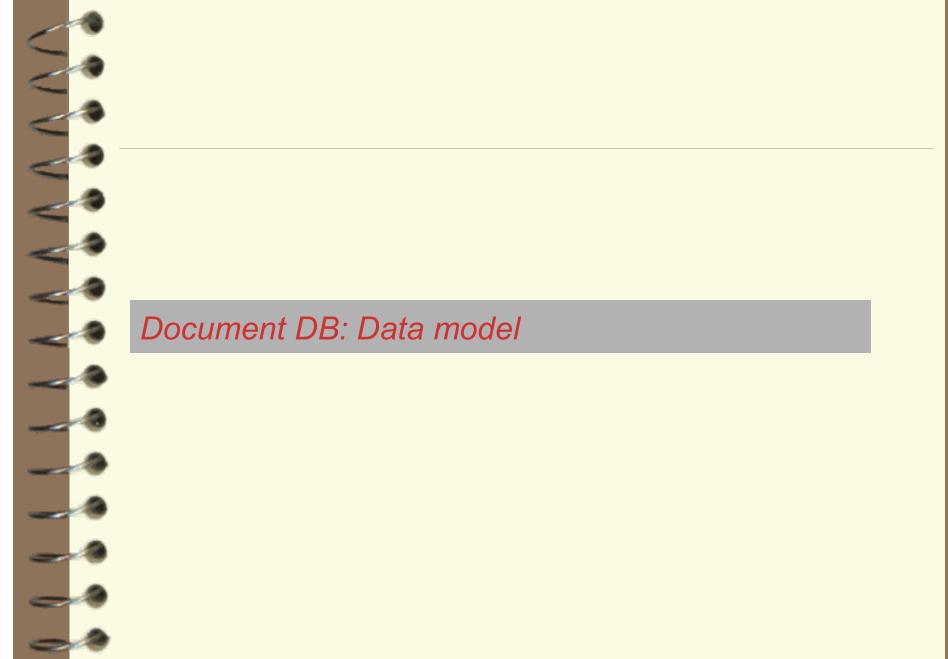
### MongoDB: CAP approach

## Focus on Consistency and Partition tolerance

- Consistency
  - all replicas contain the same version of the data
- Availability
  - system remains operational on failing nodes
- Partition tolarence
  - multiple entry points
  - system remains operational on system split

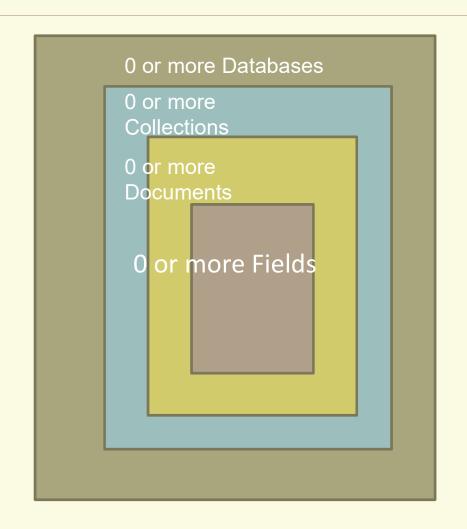


CAP Theorem: satisfying all three at the same time is impossible



### **MongoDB: Hierarchical Objects**

- A MongoDB instance may have zero or more 'databases'
- A database may have zero or more 'collections'.
- A collection may have zero or more 'documents'.
- A document may have one or more 'fields'.
- MongoDB 'Indexes' function much like their RDBMS counterparts.



### **Documents: Data Model**

- Data has a flexible schema
- This helps in matching document to objects
  - Each document can match the fields of a document also if with different structure
- Data is represented as a map
- Relations can be represented as: references and embedded documents

### **BSON**

- "Binary JSON"
- Binary-encoded serialization of JSON-like docs
- Also allows "referencing"
- Embedded structure reduces need for joins
- Goals
  - Lightweight
  - Traversable
  - Efficient (decoding and encoding)

http://bsonspec.org/

### The \_id Field

- By default, each document contains an \_id field. This field has a number of special characteristics:
  - Value serves as primary key for collection.
  - Value is unique, immutable, and may be any non-array type.
  - Default data type is ObjectId, which is "small, likely unique, fast to generate, and ordered."

http://docs.mongodb.org/manual/reference/bson-types/

### **BSON Example**

### **Documents: Structure References**

```
contact document
                                   _id: <0bjectId2>,
                                   user_id: <ObjectId1>,
                                   phone: "123-456-7890",
user document
                                   email: "xyz@example.com"
 _id: <0bjectId1>,
  username: "123xyz"
                                 access document
                                   _id: <0bjectId3>,
                                   user_id: <ObjectId1>,
                                   level: 5,
                                   group: "dev"
```

### **Documents: Structure Embedded**

```
_id: <0bjectId1>,
username: "123xyz",
contact: {
                                           Embedded sub-
            phone: "123-456-7890",
                                           document
            email: "xyz@example.com"
access: {
           level: 5,
                                           Embedded sub-
           group: "dev"
                                           document
```

### **RDB Concepts to Document DB**

RDBMS		MongoDB
Database	$\Rightarrow$	Database
Table, View	$\Rightarrow$	Collection
Row	$\Rightarrow$	Document (BSON)
Column	$\Rightarrow$	Field
Index	$\Rightarrow$	Index
Join	$\Rightarrow$	Embedded Document
Foreign Key	$\Rightarrow$	Reference
Partition	$\Rightarrow$	Shard

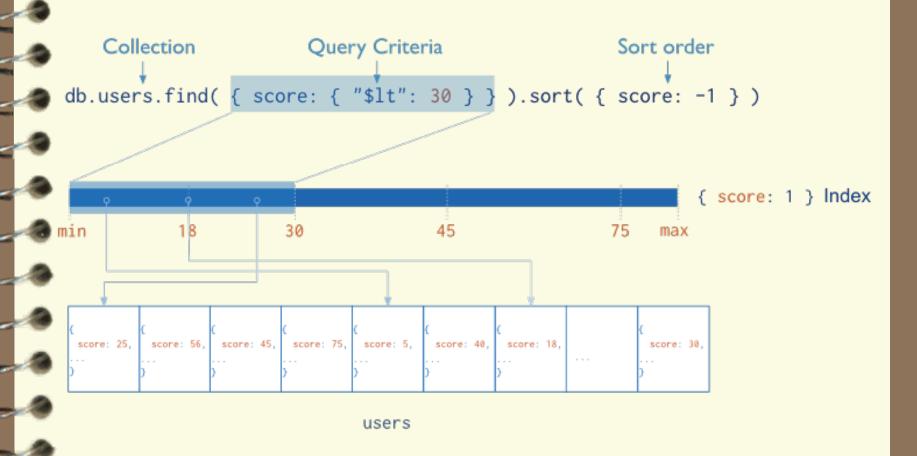
# Document DB: Indexing

### **Documents: Indexing**

- Indexes allows efficient queries on MongoDB.
- They are used to limit the number of documents to inspect (Otherwise, scan every document in a collection)
- ✓ By default MongoDB create indexes only on the \_id field
- ✓ Indexes are created using B-tree and stores data of fields ordered by values.
- ✓ In addition MongoDB returns sorted results by using the index.

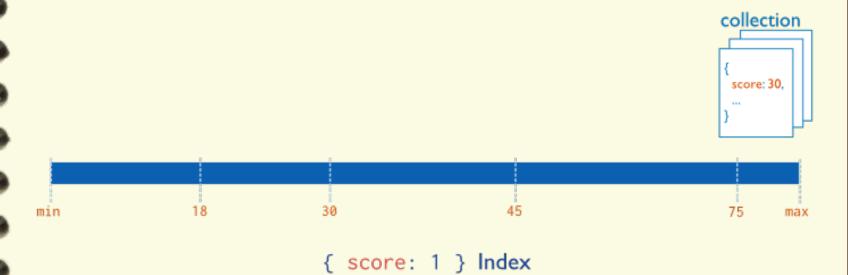
7 16

### **Documents: Indexing**



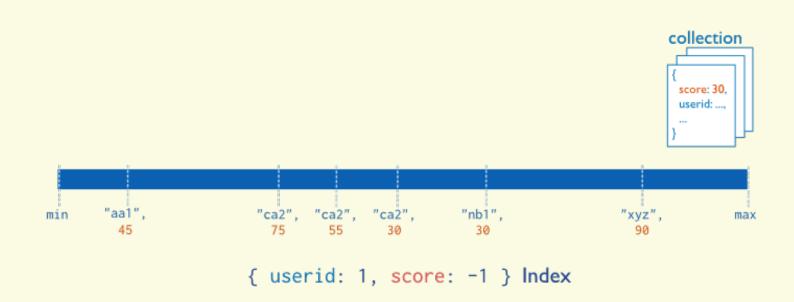
### **Documents: Index Types**

✓ Single Field



### **Documents: Index Types**

✓ Compound Index: indexed by attributes (left to right)



### **Documents: Index Types**

### Multikey Index:

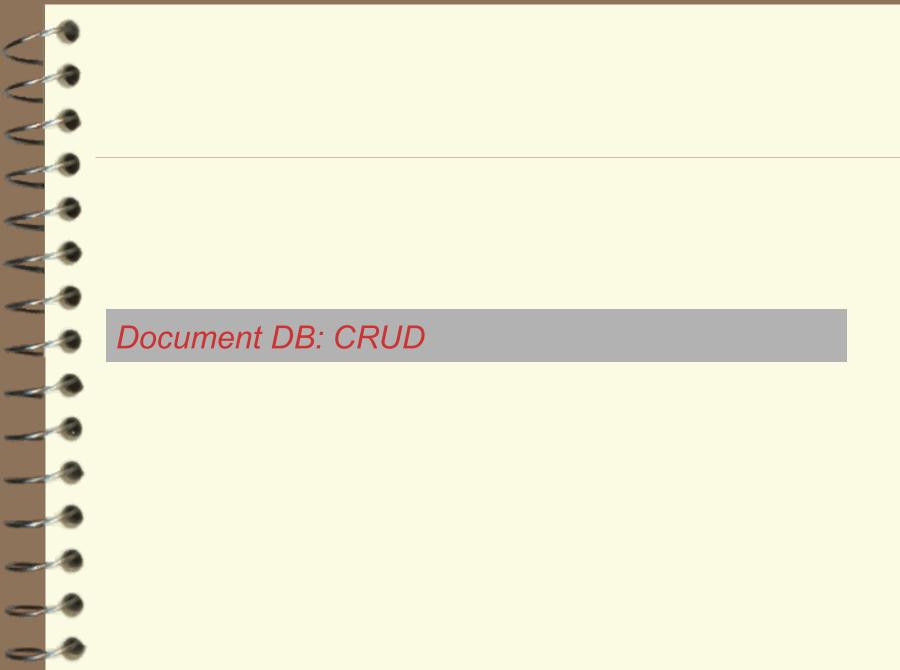
to index content in arrays

### collection

```
{
    userid: "xyz",
    addr:
        [
            { zip: "10036", ... },
            { zip: "94301", ... }
        ],
    ...
}
```

```
min "10036" "78610" "94301" max
```

```
{ "addr.zip": 1 } Index
```



### **CRUD**

- ✓ Create
  - db.collection.insert( <document> )
  - db.collection.save( <document> )
  - db.collection.update( <query>, <update>, { upsert: true } )
- ✓ Read
  - db.collection.find( <query>, , projection> )
  - db.collection.findOne( <query>, , projection> )
- ✓ Update
  - db.collection.update( <query>, <update>, <options> )
- ✓ Delete
  - db.collection.remove( <query>, <justOne> )

### **CRUD** example

```
> db.user.insert({
    first: "John",
    last : "Doe",
    age: 39
})
```

```
> db.user.find ()
{
    "_id": ObjectId("51..."),
    "first": "John",
    "last": "Doe",
    "age": 39
}
```

```
> db.user.remove({
    "first": /^J/
})
```

### **Query Interface**

The same query in SQL

### **Query Statements**

```
Query Criteria
                                                                      Modifier
    Collection
db.users.find( { age: { $gt: 18 } } ).sort( {age: 1 } )
  { age: 18, ...}
  { age: 28, ...}
                                                                    { age: 21, ...}
                                   { age: 28, ...}
  { age: 21, ...}
                                   { age: 21, ...}
                                                                   { age: 28, ...}
  { age: 38, ...}
                                   { age: 38, ...}
                                                                   { age: 31, ...}
                                                     Modifier
                  Query Criteria
  { age: 18, ...}
                                   { age: 38, ...}
                                                                    { age: 38, ...}
  { age: 38, ...}
                                   { age: 31, ...}
                                                                    { age: 38, ...}
  { age: 31, ...}
                                                                       Results
      users
```

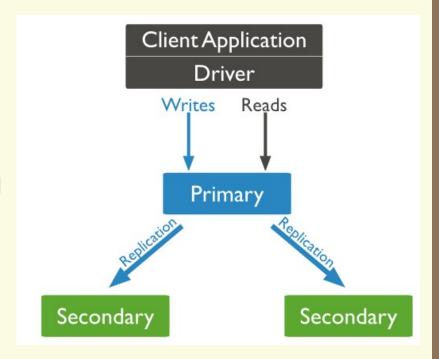
### **Projections**

```
Collection Query Criteria
                                                 Projection
db.users.find( { age: 18 }, { name: 1, _id: 0 } )
 { age: 18, ...}
 { age: 28, ...}
 { age: 21, ...}
                             { age: 18, ...}
                                                         { name: "al" }
 { age: 38, ...}
                             { age: 18, ...}
                                                         { name: "bob" }
               Query Criteria
                                            Projection
 { age: 18, ...}
                                                             Results
 { age: 38, ...}
 { age: 31, ...}
     users
```

## Document DB: Scaling

### Replication of data

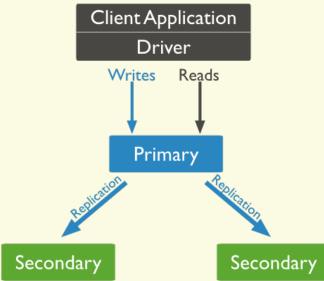
- Ensures redundancy, backup, and automatic failover
  - Recovery manager in the RDMS
- Replication through groups of servers known as replica sets
  - Primary set set of servers that client tasks direct updates to
  - Secondary set set of servers used for duplication of data
  - If the primary set fails the secondary sets 'vote' to elect the new primary set



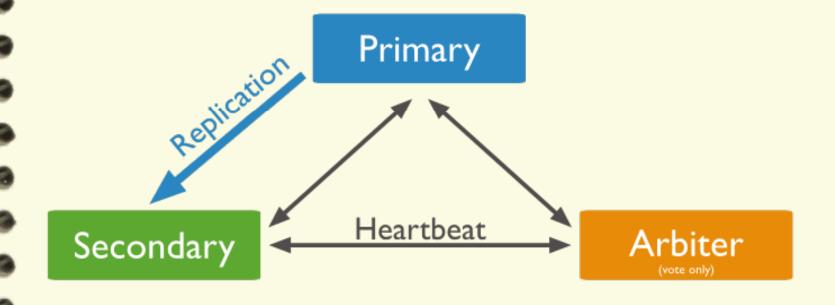
### **Scaling: Heavy Reads**

- Scaling is achieved by adding more read slaves
- All the reads can be directed to the slaves.
- ✓ When a node is added it will sync with the other nodes.
- ✓ The advantage of this setting is that we do not need to stop the cluster.

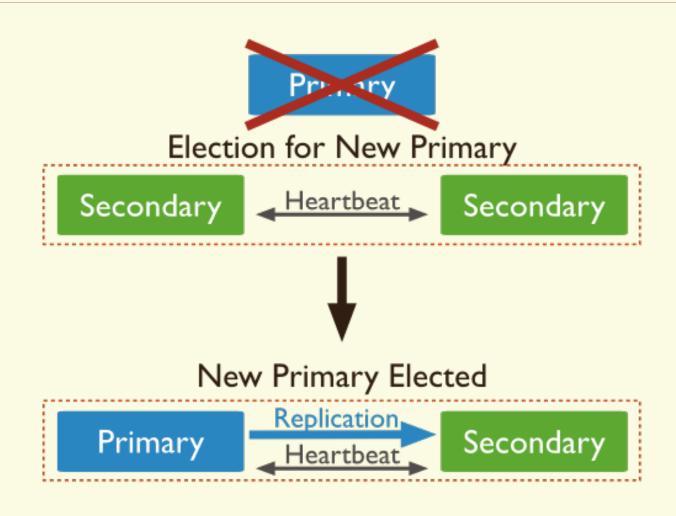
rs.add("mongo\_address:27017")



### **Data Replication**

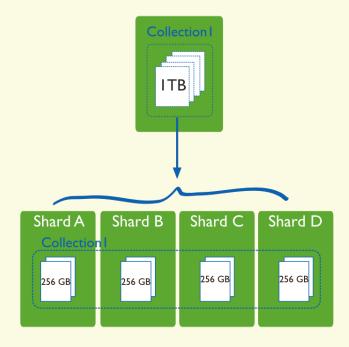


### **Automatic Failover**



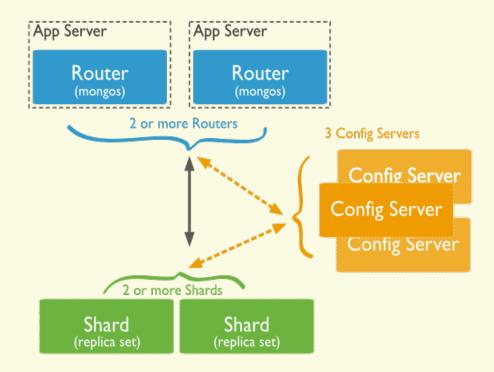
### **Scaling: Heavy Writes**

- ✓ Sharding, or horizontal scaling divides the data set and distributes the data over multiple servers.
- ✓ Each shard is an independent database, and collectively, the shards make up a single logical database.



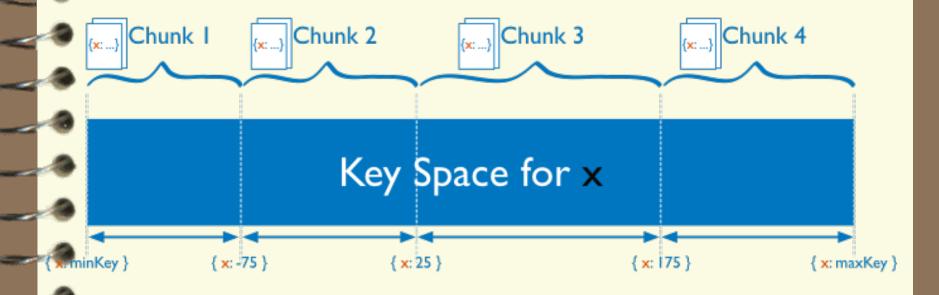
### **Sharding in MongoDB**

- ✓ Shard: store the data
- ✓ Query Routers: interface to client and direct queries
- Config Server: store cluster's metadata.



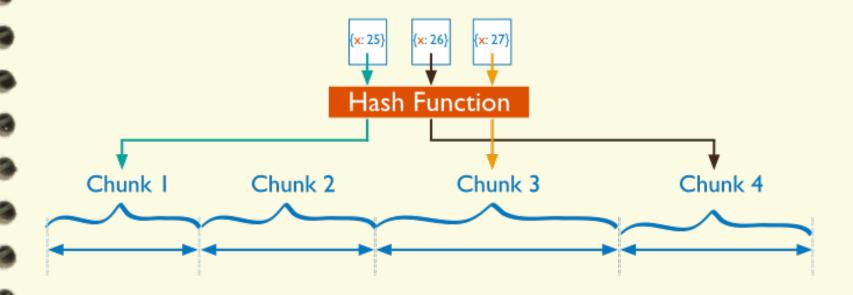
## **Range Based Sharding**

divides the data set into ranges determined by the shard key values to provide range based partitioning.



#### **Hash Based Sharding**

 computes a hash of a field's value, and then uses these hashes to create chunks



## **Performance Comparison**

- ✓ Range based partitioning supports more efficient range queries.
- ✓ However, range based partitioning can result in an uneven distribution of data.
- ✓ Hash based partitioning, by contrast, ensures an even distribution of data at the expense of efficient range queries.

#### **Document Store: Advantages**

- Documents are independent units
- ✓ Application logic is easier to write. (JSON).
- ✓ Schema Free:
  - Unstructured data can be stored easily, since a document contains whatever keys and values the application logic requires.
  - In addition, costly migrations are avoided since the database does not need to know its information schema in advance.

#### **Suitable Use Cases**

- ✓ Event Logging: where we need to store different types of event (order\_processed, customer\_logged).
- ✓ Content Management System: because the schema-free approach is well suited
- ✓ Web analytics or Real-Time Analytics: useful to update counters, page views and metrics in general.

#### When Not to Use

- ✓ Complex Transactions: when we need atomic cross-document operations (other options could be RavenDB or RethinkDB)
- ✓ Queries against Varying Aggregate Structure: i.e., when the structure of the aggregates vary because of continuous data evolutions

# Graph databases

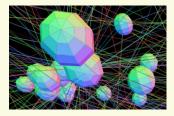












#### **Motivations**

- ✓ The necessity to represent, store and manipulate complex data make RDBMS somewhat obsolete
- ✓ Problem 1: Violations of the 1NF
  - Multi-valued attributes
  - Complex attributes
- Problem 2 : Accommodate Changes
  - acquiring data from autonomous dynamic sources or Web
  - RDBMS require schema renormalization
- ✓ Problem 3: Unified representation for:
  - Data
  - Knowledge (Schemas are a subset of this)
  - Queries [results + def]
  - Models (Concepts)

#### **Existing Approaches**

- ✓ RDBMS –need schema renormalization
- Approaches that try to fix the above mentioned problems:
  - OO Databases [P1], [P2] graphs [but procedural]
  - XML Databases [P1] (somewhat [P3]) trees
  - OORDBMS [P1] graphs with foreign keys
  - RDF triple stores [P1, P2], somewhat [P3]
- ✓ Others
  - Datalog more efficient fragment of Prolog
  - Network Models graphs
  - Hierarchical Models trees

## What is a Graph Database?

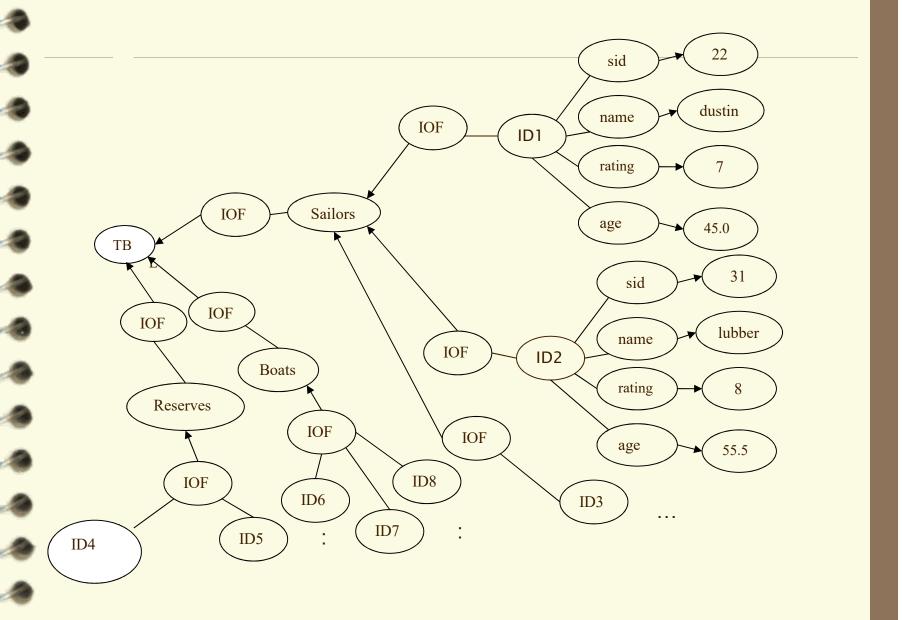
- ✓ A database with an explicit graph structure: Each node knows its adjacent nodes
- ✓ As the number of nodes increases, the cost of a local step (or hop) remains the same; Plus an Index for lookups
- Express Queries as Traversals. Fast deep traversal instead of slow SQL queries that span many table joins.
- ✓ Very natural to express graph related problem with traversals (recommendation engine, find shortest path etc..)
- Seamless integration with various existing programming languages.
  - Two design principle: Declarativity & Change
- Distinguish between "Database for graph as object"!

# **Database Representation**

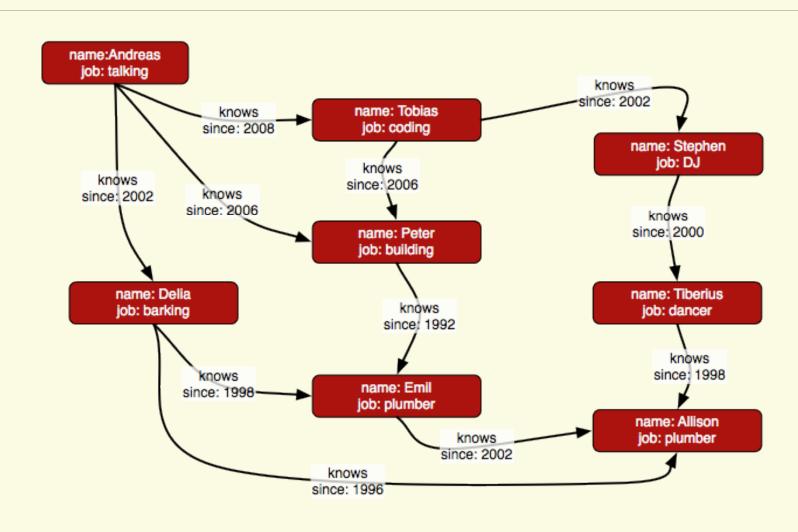
- ✓ Sailors(sid:integer, sname:char(10), rating: integer, age:real)
- ✓ Boats(bid:integer, bname:char(10), color:char(10))
- ✓ Reserve(sid:integer, bid:integer, day:date)

<b>,</b>		Sailors			Reserves			Boats		
_								_		
9	<u>sid</u>	sname	rating	age	sid	<u>bid</u>	day	bid	bname	color
	22	dustin	7	45.0	22	101	10/10/9	101	Interlake	red
3	31	lubber	8	55.5	F.0	102	6	102	Clipper	green
	58	rusty	10	35.0	58	103	11/12/9 6	103	Marine	red

## **Graph Representation**

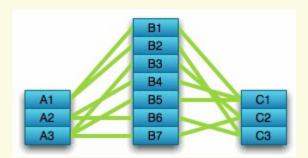


## **Property Graph**

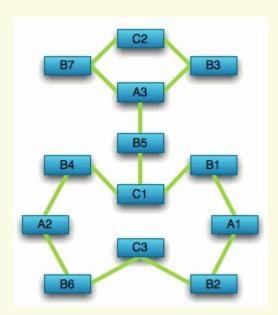


## **Compared to Relational Databases**

Optimized for aggregation



Optimized for connections

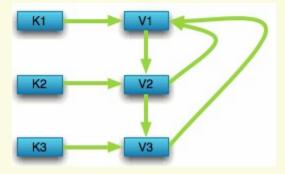


## **Compared to Key Value Stores**

Optimized for simple look-ups



Optimized for traversing connected data

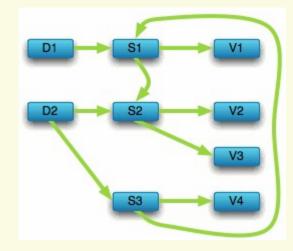


## **Compared to Document Stores**

Optimized for "trees" of data



Optimized for seeing the forest and the trees, the branches, and the trunks



# Social Network "path exists" Performance

#### Experiment:

- ~1k persons
- Average 50 friends per person
- pathExists(a,b) limitedto depth 4

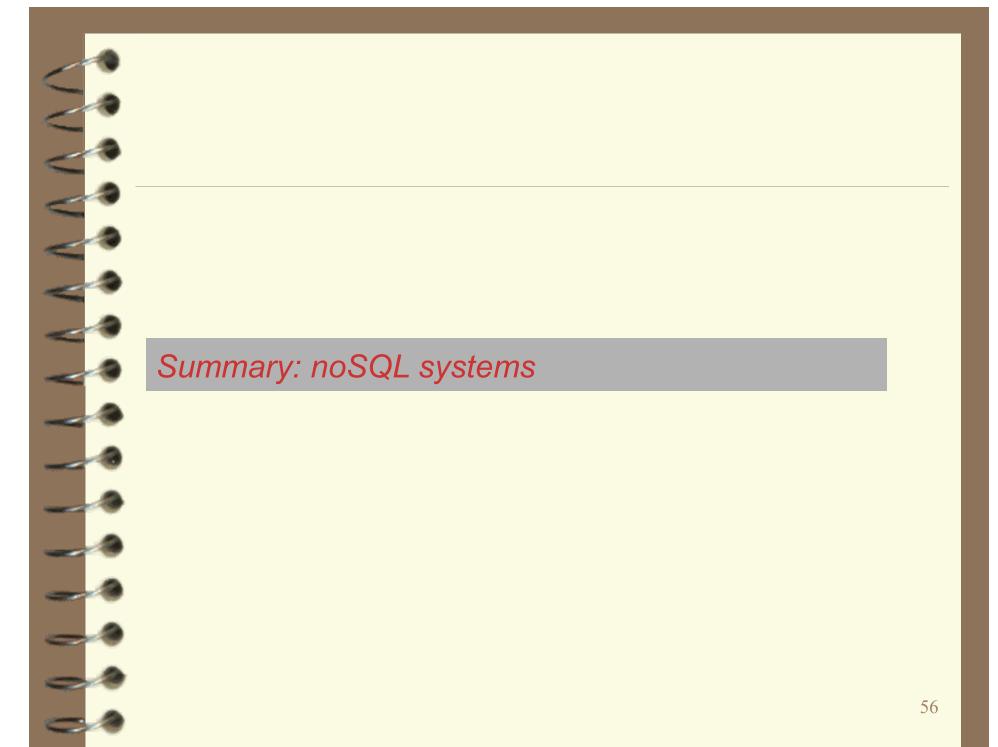
	# persons	query time
Relational database	1000	2000ms
Neo4j	1000	2ms
Neo4j	1000000	2ms

## Neo4j?

- ✓ A Graph Database + Lucene Index
- Property Graph
- ✓ Full ACID (atomicity, consistency, isolation, durability) (?)
- ✓ High Availability (with Enterprise Edition)
- ✓ 32 Billion Nodes, 32 Billion Relationships, 64 Billion Properties
- ✓ Embedded Server
- ✓ REST API

#### **Good For**

- Highly connected data (social networks)
- ✓ Recommendations (e-commerce)
- ✓ Path Finding (how do I know you?)
- ✓ A\* (Least Cost path)
- ✓ Data First Schema (bottom-up, but you still need to design)



#### Summary

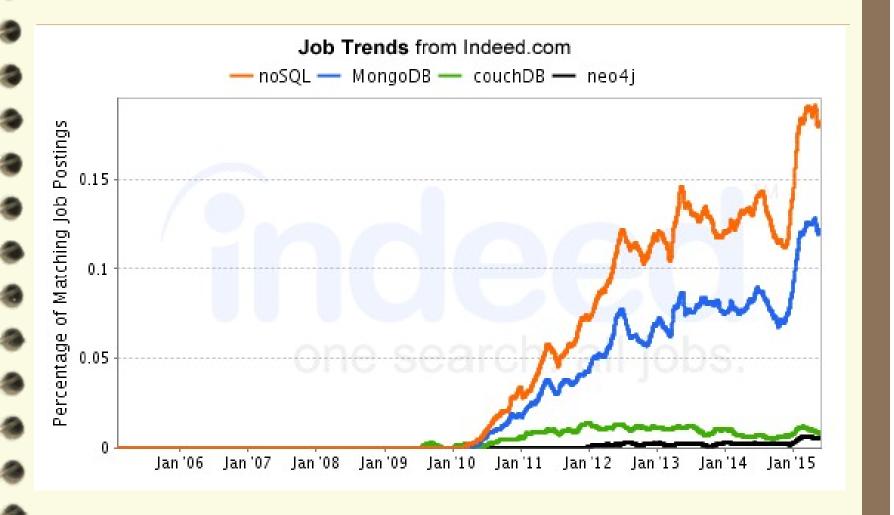
- ✓ SQL Databases
  - Predefined Schema
  - Standard definition and interface language
  - Tight consistency (ACID)
  - Well defined semantics
- NoSQL Database
  - No predefined Schema
  - Per-product definition and interface language
  - Getting an answer quickly is more important than getting a correct answer (BASE)

## **Summary: noSQL Common Advantages**

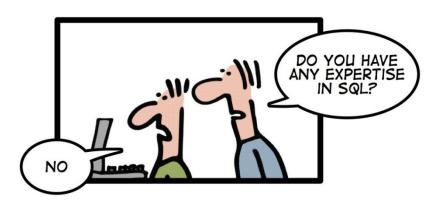
- ✓ Cheap, easy to implement (open source)
- ✓ Data are replicated to multiple nodes (therefore identical and fault-tolerant) and can be partitioned
  - Down nodes easily replaced
  - No single point of failure
- Easy to distribute
- ✓ Don't require a schema
- Can scale up and down
- ✓ Relax the data consistency requirement (CAP)

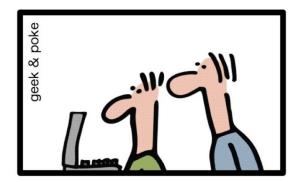
## Summary: What are we giving up?

- ✓ joins
- ✓ group by
- ✓ order by
- ✓ ACID transactions (none are strict ACID!)
- SQL as a sometimes frustrating but still powerful query language
- easy integration with other applications that support SQL



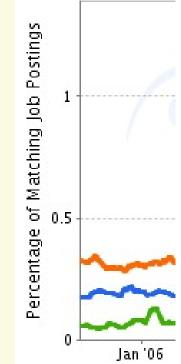
# HOW TO WRITE A CV











Jan '14

Jan '15