# **CSE 344**

### JULY 11<sup>TH</sup> SEMI-STRUCTURED DATA

### **ADMINISTRATIVE MINUTIAE**

- HW3 due today
- HW4 out tomorrow (Datalog)
- WQ4 & WQ5 due Friday
  - relational algebra & Datalog

#### So far we have studied the relational data model

- Data is stored in tables(=relations)
- Queries are expressions in SQL, relational algebra, or Datalog

#### **Traditional RDBMSs cannot scale to support modern web apps**

- limited to one (big) machine
- scale web servers only until the DB becomes the bottleneck

### Scale up by spreading the data or requests across machines

- BUT consistency becomes much harder
  - may need to give up on consistency checks involving multiple rows
  - e.g., no checking of foreign key constraints
  - can still check column non-null, greater than 0, etc.
- BUT joins become much harder if data is spread across machines
  - may need give up on these
  - (we will see how to do these in a couple weeks)

### Original NoSQL systems put scalability before everything else

- For <u>OLTP</u> workloads only: users read/write small amount of data
  - huge database but each request looks at only a small part
- Simplest version *efficiently* supports only get/put of (key, value) pairs
  - provide no ability to join
  - provide no ability to select on anything but key!
  - these are "distributed hash tables" not databases
- Other types support extensible columns or documents
  - can efficiently select on anything in one row, but still no joins
- Scale data by partitioning across many servers
  - primary key is hashed to choose server where row lives
  - row-level consistency takes <u>no</u> inter-machine communication

### Two standard techniques (for distributed systems in general)

- partitioning: each row lives on only one machine
  - good for OLTP
  - consistency checks can be done on one machine
- replication: each row lives on all (or some) machines
  - bad for OLTP: machines could have different versions of a row
  - good for OLAP, assuming each machine has all the data
  - have all the data needed for joins on the machine
  - does not scale to large data, just large users / requests
- modern systems use both
  - e.g., partition primary copy, replicate (stale) backups

#### Newer NoSQL systems can support full feature set

- research systems (e.g., Asterix) do this now
- expect to see production systems in the future

#### NoSQL now sometimes used to simply mean non-relational data

semi-structured data (documents)

### WHERE WE ARE

### **Today: Semistructured data model**

- Popular formats today: XML, JSon, protobuf
  - book discusses XML (out of favor now)
  - we will discuss JSon
  - (protobuf is just a binary / condensed form of this)
- semi-structured data is more flexible for users
- no free lunch: lack of structure also has costs
  - less work when writing
  - more work when querying

# **JSON - OVERVIEW**

JavaScript Object Notation = lightweight text-based open standard designed for human-readable data interchange. Interfaces in C, C++, Java, Python, Perl, etc.

The filename extension is .json.

(History: easiest data format within the browser)

### We will emphasize JSon as semi-structured data

# **JSON SYNTAX**

```
{ "book": [
   {"id":"01",
     "language": "Java",
     "author": "H. Javeson",
      "year": 2015
   },
   {"id":"07",
     "language": "C++",
     "edition": "second"
     "author": "E. Sepp",
     "price": 22.25
    }
```

# **JSON VS RELATIONAL**

### **Relational data model**

- Rigid flat structure (tables)
- Schema must be fixed in advanced
- Binary representation: good for performance, bad for exchange
- Query language based on Relational Calculus

#### Semistructured data model / JSon

- Flexible, nested structure (trees)
- Does not require predefined schema ("self describing")
- Text representation: good for exchange, bad for performance
- Most common use: Language API; query languages emerging

# **JSON TERMINOLOGY**

### Data is represented in name/value pairs.

Rows replaced by objects

### **Curly braces hold objects**

- Each object is a list of name/value pairs separated by , (comma)
- Each pair is a name is followed by ':'(colon) followed by the value

Square brackets hold arrays and values are separated by ,(comma).

# JSON DATA STRUCTURES

### **Collections of name-value pairs:**

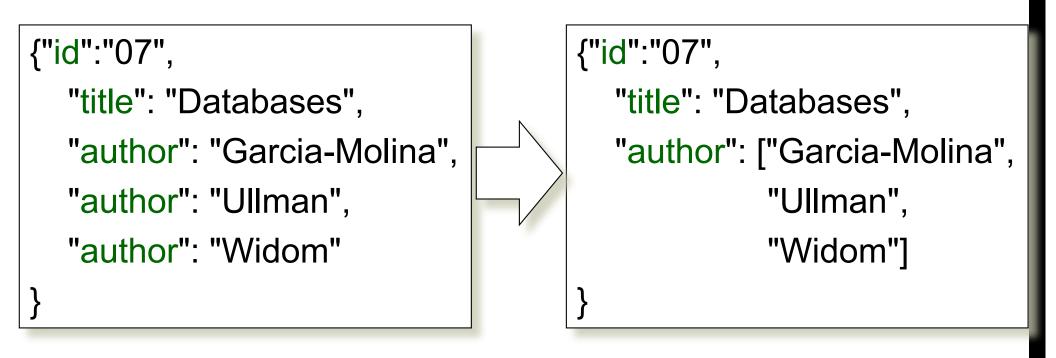
- {"name1": value1, "name2": value2, ...}
- The "name" is also called a "key" (or "field")

### **Ordered lists of values:**

• [obj1, obj2, obj3, ...]

# AVOID USING DUPLICATE KEYS

The standard allows them, but many implementations don't



# **JSON DATATYPES**

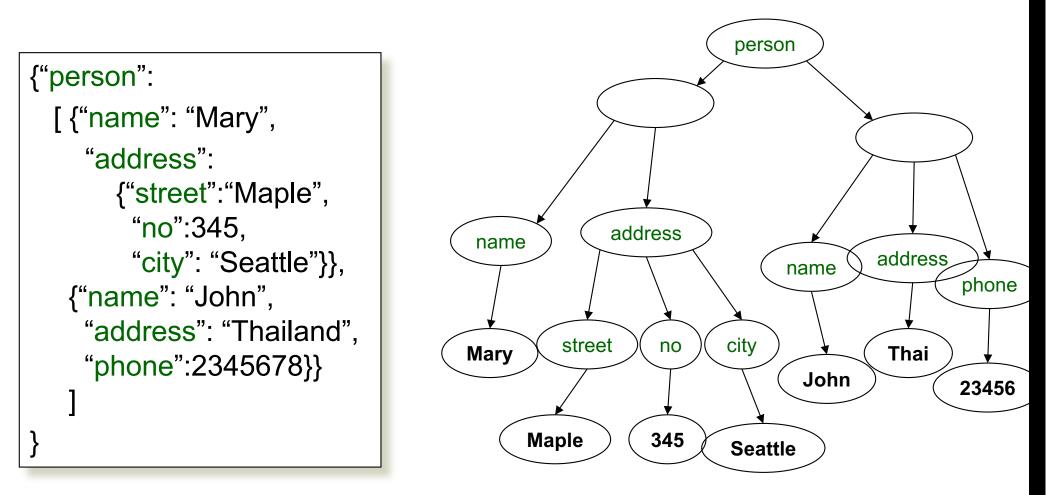
Number

String = double-quoted

**Boolean = true or false** 

null empty

### **JSON SEMANTICS: A TREE !**



### **JSON DATA**

### JSon is self-describing

#### Schema elements become part of the data

- Relational schema: person(name,phone)
- In Json "person", "name", "phone" are part of the data, and are repeated many times

### **Consequence: JSon is much more flexible (but wasteful)**

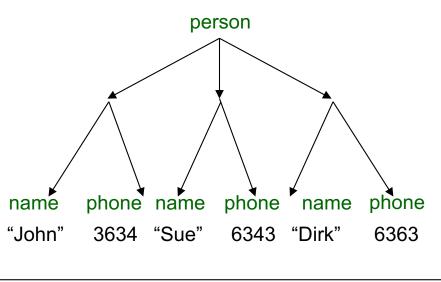
### JSon is also semistructured data

• (more soon...)

# MAPPING RELATIONAL DATA TO JSON

#### Person

name	phone
John	3634
Sue	6343
Dirk	6363



#### {"person":

[{"name": "John", "phone":3634}, {"name": "Sue", "phone":6343}, {"name": "Dirk", "phone":6383}

### MAPPING RELATIONAL DATA TO .

May inline foreign keys

#### Person

name	phone
John	3634
Sue	6343

#### Orders

personName	date	product
John	2002	Gizmo
John	2004	Gadget
Sue	2002	Gadget

```
{"Person":
   [{"name": "John",
    "phone":3646,
    "Orders":[{"date":2002,
              "product":"Gizmo"},
              {"date":2004,
               "product":"Gadget"}
    {"name": "Sue",
     "phone":6343,
      "Orders":[{"date":2002,
                "product":"Gadget"}
```

# JSON=SEMI-STRUCTURED DATA (1/3)

### **Missing attributes:**

{"person":
 [{"name":"John", "phone":1234},
 {"name":"Joe"}]
}

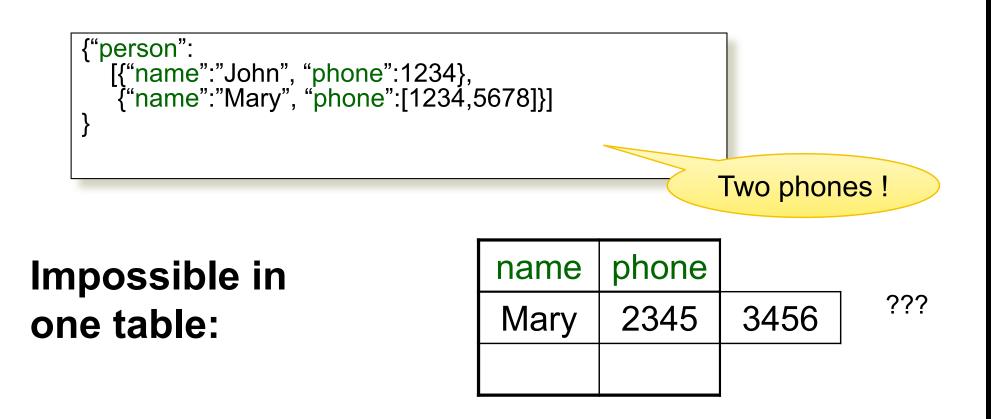
no phone !

# Could represent in a table with nulls

name	phone
John	1234
Joe	-

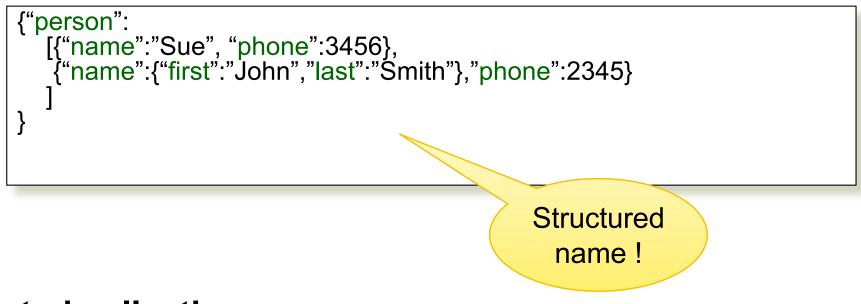
# **JSON=SEMI-STRUCTURED DATA (2/3)**

### **Repeated attributes**



# JSON=SEMI-STRUCTURED DATA (3/3)

### Attributes with different types in different objects



### Nested collections

### **Heterogeneous collections**

• Downside: you need to think about these cases in queries!

# QUERY LANGUAGES FOR SS DATA

### XML: XPath, XQuery (textbook)

- Supported inside many RDBMS (SQL Server, DB2, Oracle)
- Several standalone XPath/XQuery engines
- XPath widely used in the browser: CSS, JQuery

### JSon:

- CouchBase: N1QL, may be replaced by AQL (better designed)
- Asterix: SQL++ (based on SQL)
- MongoDB: has a pattern-based language
- JSONiq <u>http://www.jsoniq.org/</u>

# **ASTERIXDB AND SQL++**

### AsterixDB

- NoSQL database system
- Developed at UC Irvine
- Now an Apache project
- Own query language: AsterixQL or AQL, based on XQuery

### SQL++

SQL-like syntax for AsterixQL

# ASTERIX DATA MODEL (ADM)

### **Objects:**

- {"Name": "Alice", "age": 40}
- Fields must be distinct: {"Name": "Alice", "age": 40, "age": 50}

### Arrays:

- [1, 3, "Fred", 2, 9]
- Note: can be heterogeneous

### Multisets:

• {{1, 3, "Fred", 2, 9}}





Try these queries:

SELECT x.age FROM [{'name': 'Alice', 'age': ['30', '50']}] x;

SELECT x.age FROM {{ {'name': 'Alice', 'age': ['30', '50']} }} x;

-- error SELECT x.age FROM {'name': 'Alice', 'age': ['30', '50']} x;



Boolean, integer, float (various precisions), geometry (point, line, ...), date, time, etc

UUID = universally unique identifier Use it as a system-generated unique key

# NULL V.S. MISSING

{"age": null} = the value NULL (like in SQL)
{"age": missing} = { } = really missing

SELECT x.b FROM [{'a':1, 'b':2}, {'a':3}] x;

```
{ "b": { "int64": 2 } } 
{ }
```

SELECT x.b FROM [{'a':1, 'b':2}, {'a':3, 'b':missing }] x;

{ "b": { "int64": 2 } } { }

### **SQL++ OVERVIEW**

### Data Definition Language (DDL): create a

- Dataverse
- Type
- Dataset
- Index

### Data Manipulation Language (DML): select-from-where



A Dataverse is a Database

CREATE DATAVERSE lec344 CREATE DATAVERSE lec344 IF NOT EXISTS

DROP DATAVERSE lec344

DROP DATAVERSE lec344 IF EXISTS

USE lec344

### TYPE

Defines the schema of a collection It lists all <u>required</u> fields Fields followed by ? are <u>optional</u> CLOSED type = no other fields allowed OPEN type = other fields allowed

### **Semi-structured** data

- type defines the structured part
- rest is unstructured

### **CLOSED TYPES**

```
USE lec344;
DROP TYPE PersonType IF EXISTS;
CREATE TYPE PersonType AS CLOSED {
Name : string,
age: int,
email: string?
}
```

{"Name": "Alice", "age": 30, "email": "a@alice.com"}

```
{"Name": "Bob", "age": 40}
```

-- not OK: {"Name": "Carol", "phone": "123456789"}



```
USE lec344;
DROP TYPE PersonType IF EXISTS;
CREATE TYPE PersonType AS OPEN {
Name : string,
age: int,
email: string?
}
```

{"Name": "Alice", "age": 30, "email": "a@alice.com"}

```
{"Name": "Bob", "age": 40}
```

-- Now it's OK: {"Name": "Carol", "phone": "123456789"}

# TYPES WITH NESTED COLLECTIONS

```
USE lec344;
DROP TYPE PersonType IF EXISTS;
CREATE TYPE PersonType AS CLOSED {
Name : string,
phone: [string]
}
```

```
{"Name": "Carol", "phone": ["1234"]}
{"Name": "David", "phone": ["2345", "6789"]}
{"Name": "Eric", "phone": []}
```



#### **Dataset = relation**

#### Must have a type

Can be a trivial OPEN type

#### Must have a key

• Can also be a trivial one

# DATASET WITH EXISTING KEY

USE lec344; DROP TYPE PersonType IF EXISTS; CREATE TYPE PersonType AS CLOSED { Name : string, email: string? }

{"Name": "Alice"} {"Name": "Bob"}

USE lec344; DROP DATASET Person IF EXISTS; CREATE DATASET Person(PersonType) PRIMARY KEY Name;

# DATASET WITH AUTO GENERATED KEY

```
USE lec344;
DROP TYPE PersonType IF EXISTS;
CREATE TYPE PersonType AS CLOSED {
myKey: uuid,
Name : string,
email: string?
}
```

USE lec344; DROP DATASET Person IF EXISTS; CREATE DATASET Person(PersonType) PRIMARY KEY myKey AUTOGENERATED; {"Name": "Alice"} {"Name": "Bob"}

Note: no myKey since it will be autogenerated

## **DISCUSSION OF NFNF**

NFNF = Non First Normal Form

One or more attributes contain a collection

One extreme: a single row with a huge, nested collection

Better: multiple rows, reduced number of nested collections



mondial.adm is totally semistructured:
{"mondial": {"country": [...], "continent":[...], ..., "desert":[...]}}

country	continent	organization	sea	 mountain	desert
[{"name":"Albania",}, {"name":"Greece",}, ]					

country.adm, sea.adm, mountain.adm are more structured

Country:

-car_code	name	 ethnicgroups	religions	 city
AL	Albania	 []	[]	 []
GR	Greece	 []	[]	 []



#### Can declare an index on an attribute of a top-most collection

- used to improve query performance
  - allows DBMS to perform certain lookups efficiently
- (more next week...)

### Available:

- BTREE: good for equality and range queries
   E.g. name="Greece"; 20 < age and age < 40</li>
- RTREE: good for 2-dimensional range queries
   E.g. 20 < x and x < 40 and 10 < y and y < 50</li>
- KEYWORD: good for substring search

### **INDEXES**

# Cannot index inside a nested collection

USE lec344; CREATE INDEX countryID ON country(`-car\_code`) TYPE BTREE;

AL BG GR... NZ

USE lec344; CREATE iNDEX cityname ON country(city.name) TYPE BTREE;

Country:

-car_code	name	 ethnicgroups	religions		city
AL	Albania	 []	[]	•••	[]
GR	Greece	 []	[]		[]
BG	Belgium				