Introduction to Data Management CSE 344

Unit 3: NoSQL, JSON, Semistructured Data (3 lectures*)

*Slides may change: refresh each lecture

Introduction to Data Management CSE 344

Lecture 11: NoSQL

Announcements

• HW3 (Azure) due on Friday

• HW4 (datalog) due next Friday

• Midterm next Friday (May 3rd)

Class Overview

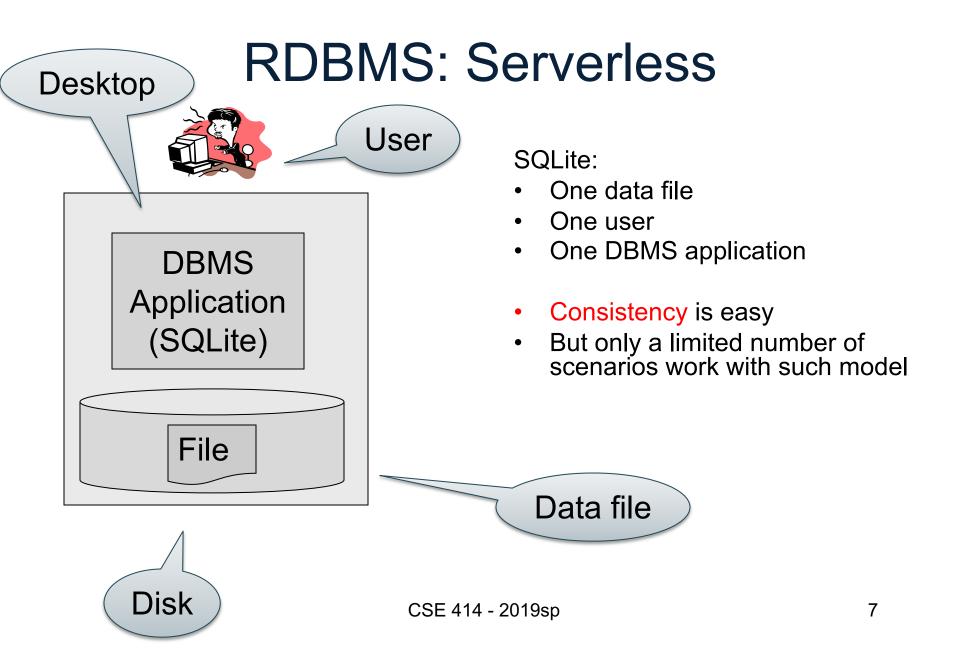
- Unit 1: Intro
- Unit 2: Relational Data Models and Query Languages
- Unit 3: Non-relational data
 - NoSQL
 - JSON
 - SQL++
- Unit 4: RDMBS internals and query optimization
- Unit 5: Parallel query processing
- Unit 6: DBMS usability, conceptual design
- Unit 7: Transactions
- Unit 8: Advanced topics (time permitting)

Two Classes of Database Applications

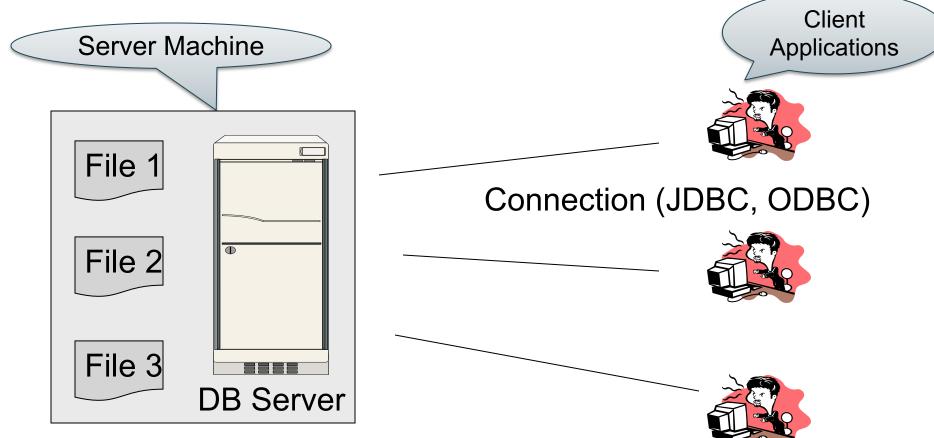
- OLTP (Online Transaction Processing)
 - Queries are simple lookups: 0 or 1 join
 E.g., find customer by ID and their orders
 - Many updates. E.g., insert order, update payment
 - Consistency is critical: transactions (more later)
- OLAP (Online Analytical Processing)
 - aka "Decision Support"
 - Queries have many joins, and group-by's
 E.g., sum revenues by store, product, clerk, date
 - No updates

RDBMS Architectures

- Serverless
- 2 tier: client/server
- 3 tier: client/app-server/db-server

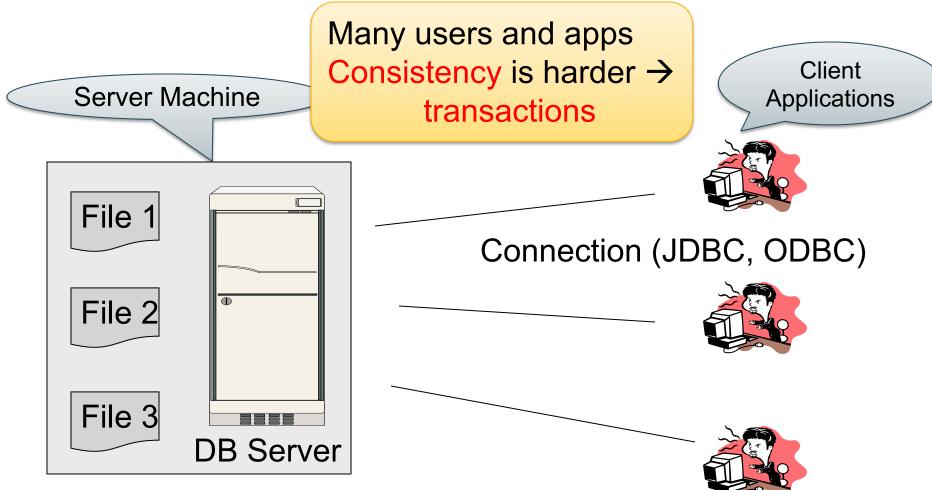


RDBMS: Client-Server



- One server running the database
- Many clients, connecting via the ODBC or JDBC (Java Database Connectivity) protocol

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Client-Server

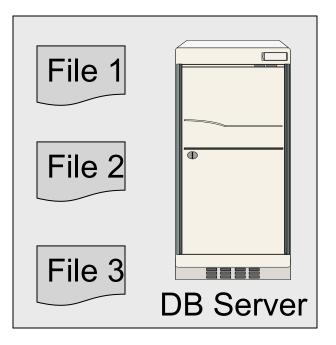
- One server that runs the DBMS (or RDBMS):
 - Your own desktop, or
 - Some beefy system, or
 - A cloud service (SQL Azure)

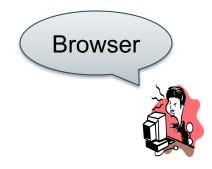
Client-Server

- One *server* that runs the DBMS (or RDBMS):
 - Your own desktop, or
 - Some beefy system, or
 - A cloud service (SQL Azure)
- Many *clients* run apps and connect to DBMS
 - Microsoft's Management Studio (for SQL Server), or
 - psql (for postgres)
 - Some Java program (HW8) or some C++ program

Client-Server

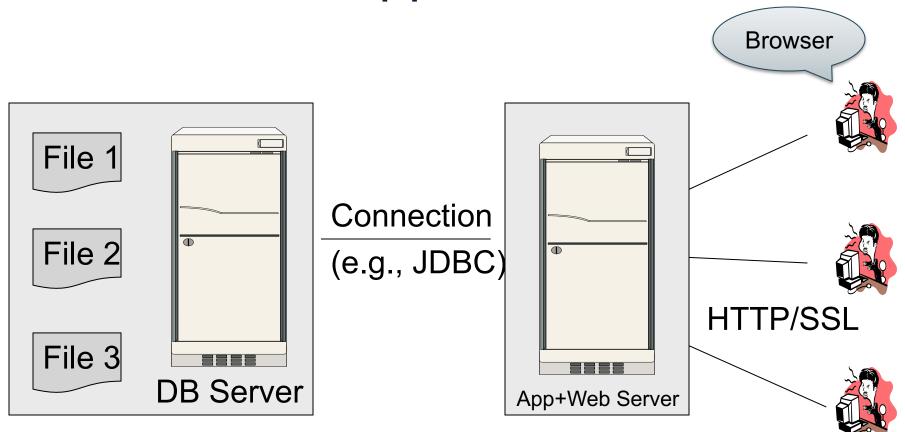
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 - Some Java program (HW8) or some C++ program
- Clients "talk" to server using JDBC/ODBC protocol

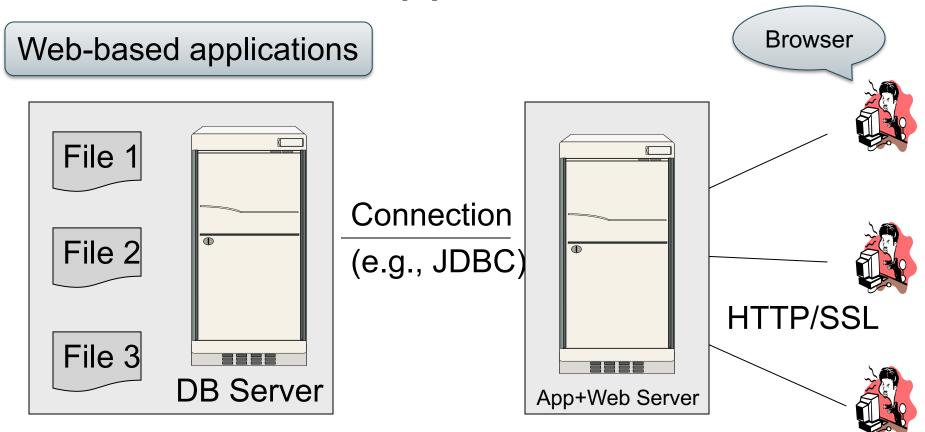


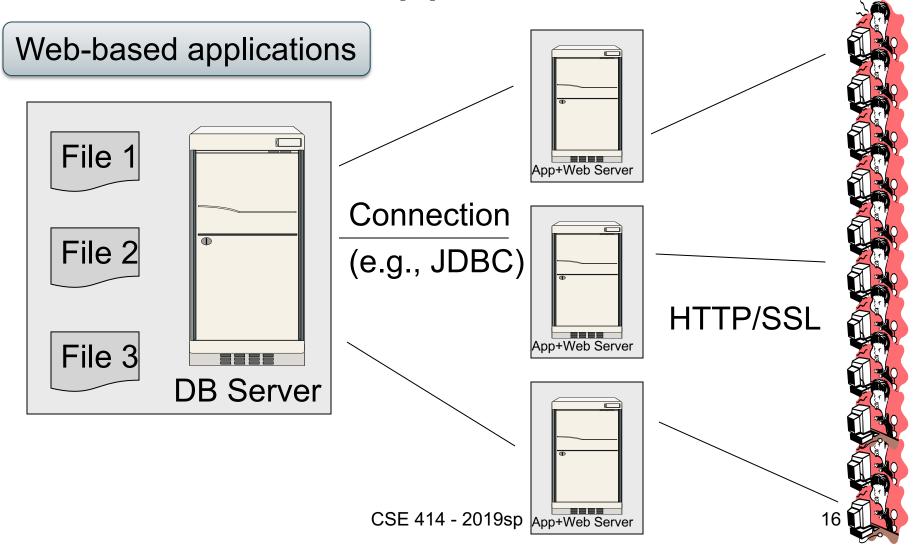


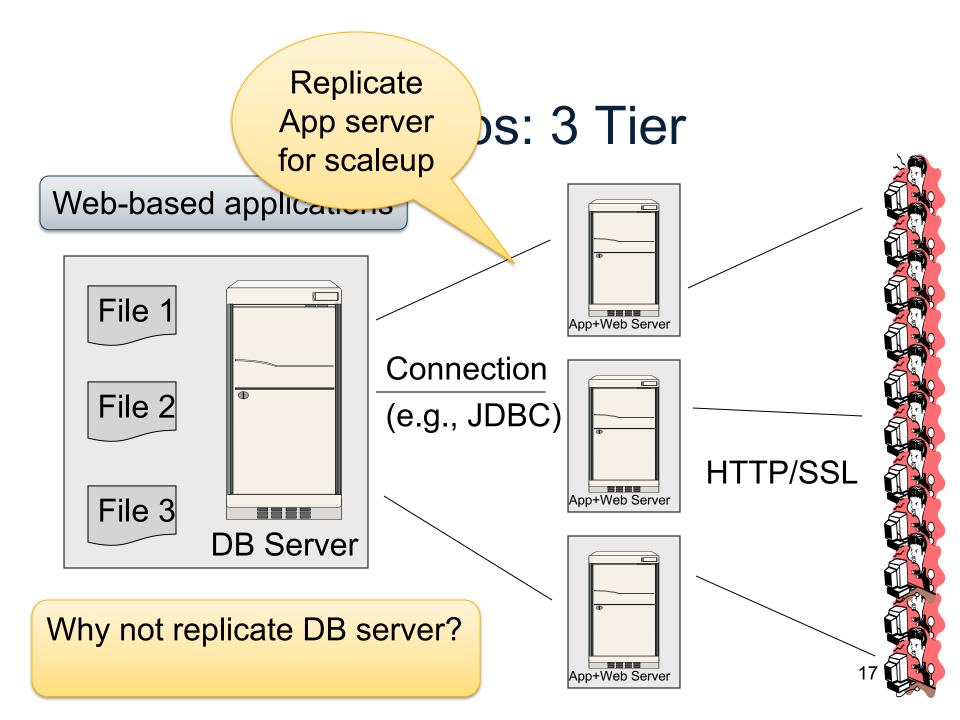


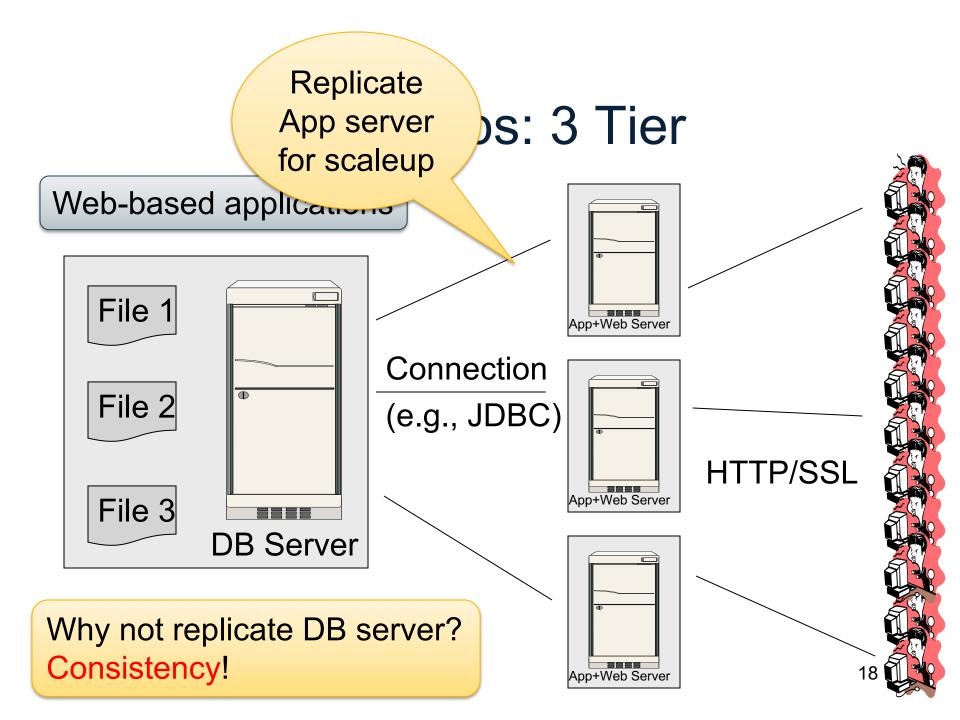












NoSQL Motivation

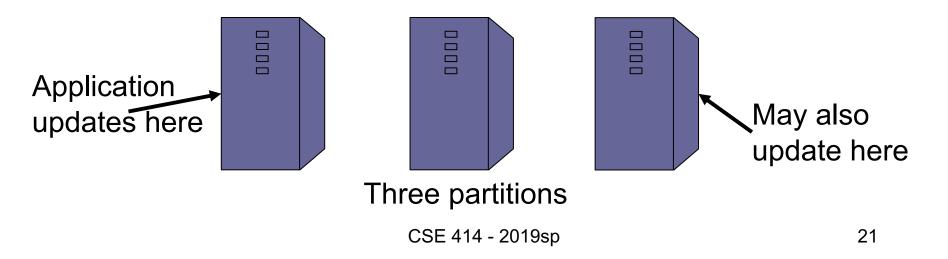
- Originally motivated by Web 2.0 applications
 - E.g. Facebook, Amazon, Instagram, etc
 - Startups need to scaleup from 10 to 10⁷ quickly
- Needed: very large scale OLTP workloads
- Give up on consistency, give up OLAP
- NoSQL: reduce functionality
 - Simpler data model
 - Very restricted updates

Replicating the Database

- Two basic approaches:
 - Scale up through partitioning "sharding"
 - Scale up through replication
- Consistency is much harder to enforce

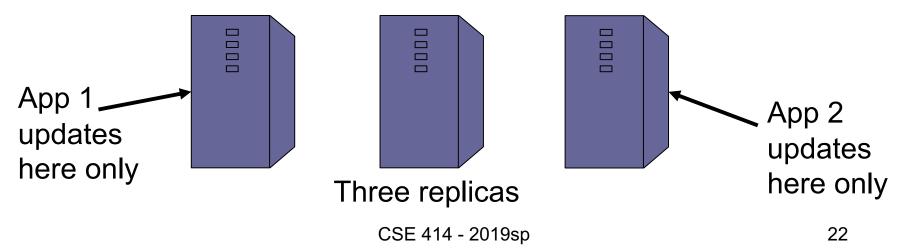
Scale Through Partitioning

- Partition the database across many machines in a cluster
 - Database now fits in main memory
 - Queries spread across these machines
- Can increase throughput
- Easy for writes but reads become expensive!



Scale Through Replication

- Create multiple copies of each database partition
- Spread queries across these replicas
- Can increase throughput and lower latency
- Can also improve fault-tolerance
- Easy for reads but writes become expensive!



Relational Model \rightarrow NoSQL

- Relational DB: difficult to replicate/partition.Eg
 Supplier(sno,...), Part(pno,...), Supply(sno,pno)
 - Partition: we may be forced to join across servers
 - Replication: local copy has inconsistent versions
 - Consistency is hard in both cases (why?)
- NoSQL: simplified data model
 - Given up on functionality
 - Application must now handle joins and consistency

Data Models

Taxonomy based on data models:

- Key-value stores
 - e.g., Project Voldemort, Memcached
 - Document stores
 - e.g., SimpleDB, CouchDB, MongoDB
 - Extensible Record Stores

– e.g., HBase, Cassandra, PNUTS

- **Data model**: (key,value) pairs
 - Key = string/integer, unique for the entire data
 - Value = can be anything (very complex object)

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Operations

- get(key), put(key,value)
- Operations on value not supported

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- get(key), put(key,value)
- Operations on value not supported
- Distribution / Partitioning w/ hash function
 - No replication: key k is stored at server h(k)
 - 3-way replication: key k stored at h1(k),h2(k),h3(k)

- **Data model**: (key,value) pairs
 - Key = string/integer, unique for the entire data
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Operations

- get(key), put(key,value)
- Operations on value not supported
- Distribution / Partitioning w/ hash function
 - No replication: key k is stored at server h(k)
 - 3-way replication: key k stored at h1(k),h2(k),h3(k)

How does get(k) work? How does put(k,v) work?

Example

• How would you represent the Flights data as key, value pairs?

Example

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- Option 1: key=fid, value=entire flight record

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- Option 2: key=date, value=all flights that day

Example

- How would you represent the Flights data as key, value pairs?
- Option 1: key=fid, value=entire flight record
- Option 2: key=date, value=all flights that day
- Option 3: key=(origin,dest), value=all flights between

Key-Value Stores Internals

- Partitioning:
 - Use a hash function h
 - Store every (key,value) pair on server h(key)
- Replication:
 - Store each key on (say) three servers
 - On update, propagate change to the other servers; eventual consistency
 - Issue: when an app reads one replica, it may be stale
- Usually: combine partitioning+replication

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 - e.g., HBase, Cassandra, PNUTS

Motivation

 In Key, Value stores, the Value is often a very complex object

- Key = '2010/7/1', Value = [all flights that date]

- Better: value to be structured data
 - JSON or Protobuf or XML
 - Called a "document" but it's just data

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Extensible Record Stores

- Based on Google's BigTable
- HBase is an open source implementation of BigTable
- Data model:
 - Variant 1: key = rowID, value = record
 - Variant 2: key = (rowID, columnID), value = field
- Will not discuss in class

Introduction to Data Management CSE 344

Lecture 12: JSON, Semistructured Data, SQL++

Where We Are

- So far we have studied the *relational data model*
 - Data is stored in tables(=relations)
 - Queries are expressions in SQL, relational algebra, or Datalog
- Today: Semistructured data model
 Popular formats today: XML, JSON, protobuf

JSON - Overview

- JavaScript Object Notation = lightweight textbased open standard designed for humanreadable data interchange. Interfaces in C, C++, Java, Python, Perl, etc.
- The filename extension is .json.

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

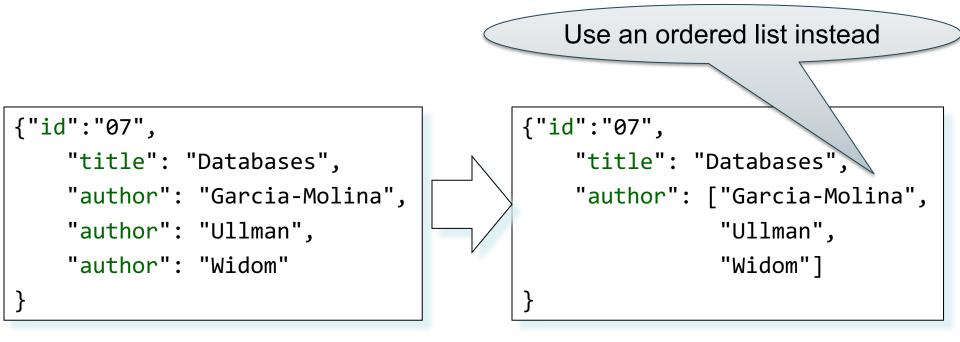
- Relational data model
 - Rigid flat structure (tables)
 - Schema must be fixed in advanced
 - Binary representation: good for performance, bad for exchange
 - Query language based on Relational Calculus
- Semistructured data model / JSON
 - Flexible, nested structure (trees)
 - Does not require predefined schema ("self-describing")
 - Text representation: good for exchange, bad for performance
 - Most common use: Language API; query languages emerging

JSON Types

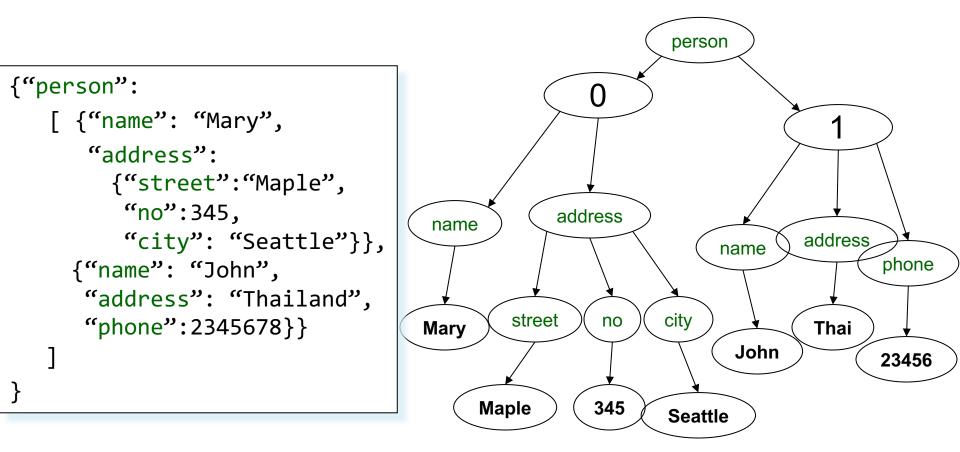
- Primitive: number, string, Boolean, null
- Object: collection of name-value pairs:
 - {"name1": value1, "name2": value2, ...}
 - "name" is also called a "key"
- Array: ordered list of values:
 [obj1, obj2, obj3, ...]

Avoid Using Duplicate Keys

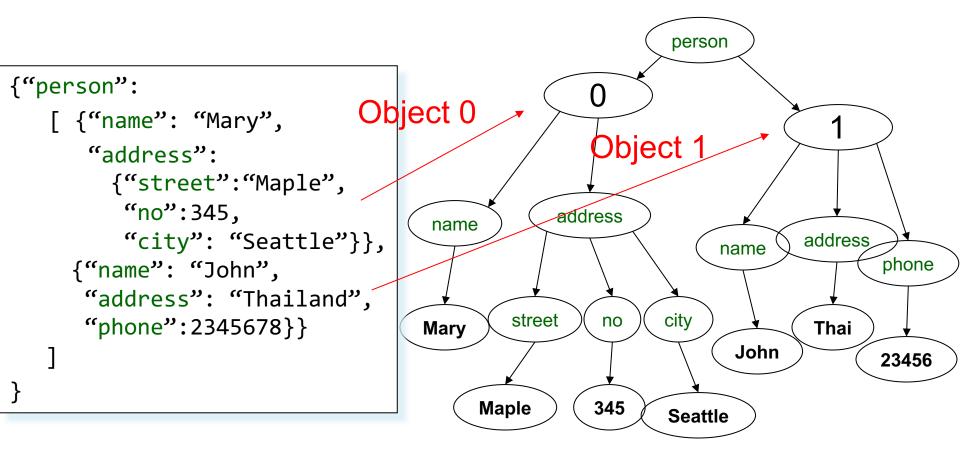
The standard allows them, but many implementations don't



JSON Semantics: a Tree !



JSON Semantics: a Tree !

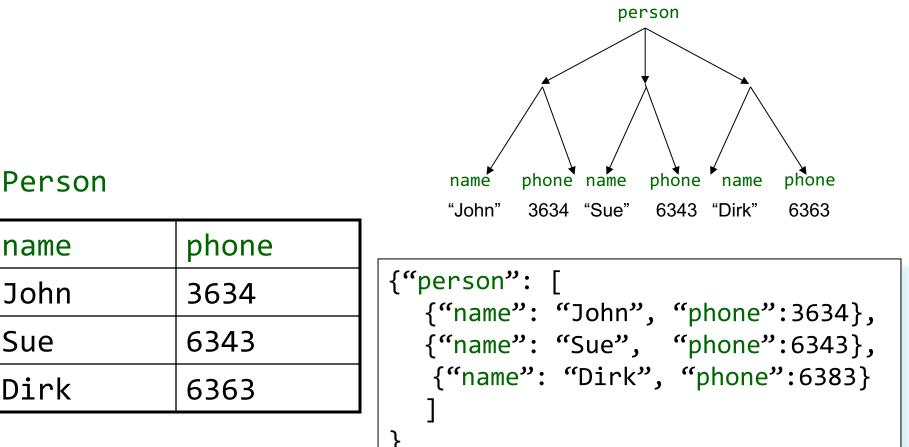


Recall: arrays are ordered in JSON!

Intro to Semi-structured Data

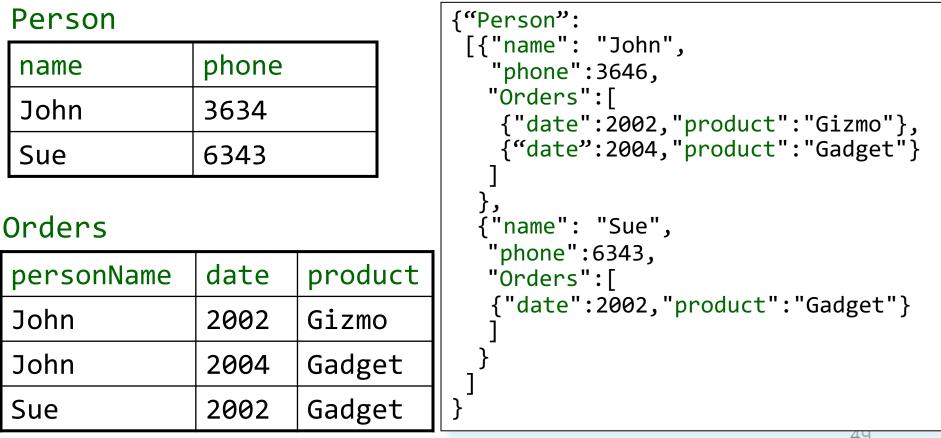
- JSON is self-describing
- Schema elements become part of the data
 - Relational schema: person(name, phone)
 - In JSON "person", "name", "phone" are part of the data, and are repeated many times
- ⇒ JSON is more flexible
 - Schema can change per tuple

Mapping Relational Data to JSON



Mapping Relational Data to JSON

May inline multiple relations based on foreign keys



Mapping Relational Data to JSON

Many-many relationships are more difficult to represent

Person

name	phone	
John	3634	
Sue	6343	

Product

prodName	price
Gizmo	19.99
Phone	29.99
Gadget	9.99

Orders

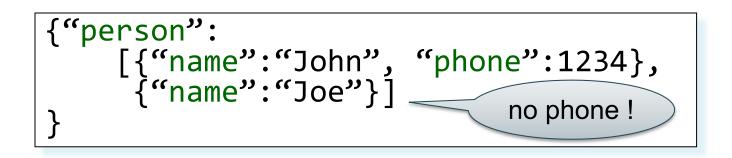
personName	date	product
John	2002	Gizmo
John	2004	Gadget
Sue	2002	Gadget

Options for the JSON file:

- 3 flat relations: Person,Orders,Product
- Person→Orders→Products products are duplicated
- Product→Orders→Person persons are duplicated

Semi-structured data

• Missing attributes:

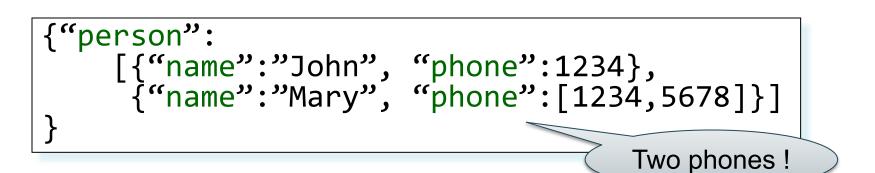


• Could represent in a table with nulls

name	phone	
John	1234	
Joe	NULL	

Semi-structured data

Repeated attributes

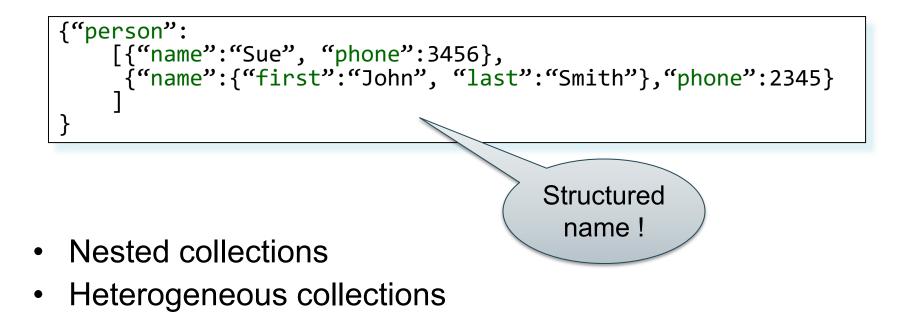


Impossible in one table:

name	phone		
Mary	2345	3456	0.0

Semi-structured data

• Attributes with different types in different objects



These are difficult to represent in the relational model

Discussion: Why Semi-Structured Data?

- Semi-structured data works well

 as data exchange formats
 i.e., exchanging data between different apps
 Examples: XML, JSON, Protobuf (protocol buffers)
- Increasingly, systems use them as a data model for databases:
 - -- SQL Server supports for XML-valued relations
 - -- CouchBase, MongoDB, Snowflake: JSON
 - -- Dremel (BigQuery): Protobuf

Query Languages for Semi-Structured Data

XML: XPath, XQuery (see textbook)

- Supported inside many RDBMS (SQL Server, DB2, Oracle)
- Several standalone XPath/XQuery engines

Protobuf:

- Dremel (~ SQL): google internal
- BigQuery (~ SQL): google external

JSON:

- CouchBase: N1QL
- AsterixDB: SQL++ (~ SQL)
- MongoDB: JSONiq: <u>http://www.jsoniq.org/</u>

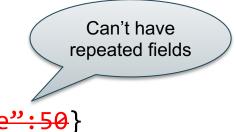


- AsterixDB
 - NoSQL database system
 - Developed at UC Irvine
 - Now an Apache project, being incorporated into CouchDB (another NoSQL DB)
- Uses JSON as data model
- Query language: SQL++
 SQL-like syntax for JSON data

ADM Derived Types

- Based on the JSON standard
- Objects:
 - {"Name": "Alice", "age": 40}
 - Fields must be distinct: {"Name": "Alice", "age": 40, "age": 50}
- Ordered arrays:
 - [1, 3, "Fred", 2, 9]
 - Can contain values of different types
- Multisets (aka bags):
 - {{1, 3, "Fred", 1, 9}}
 - Mostly internal use only but can be used as inputs
 - All multisets are converted into ordered arrays (in arbitrary order) when returned at the end





Basic Queries

What do these queries return?

SELECT x.name
FROM [{"name": "Alice", "phone": [300, 150]}] AS x;

Answer: {"name": "Alice"}

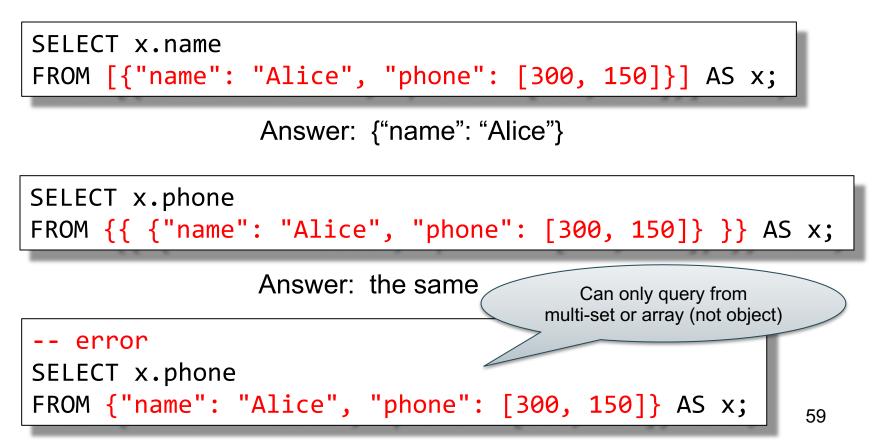
SELECT x.phone
FROM [{"name": "Alice", "phone": [300, 150]}] AS x;

Answer: {"phone": [300, 150]}

SELECT x.name, x.phone
FROM [{"name": "Alice", "phone": [300, 150]}] AS x;
Answer: {"name": "Alice", "phone": [300, 150]}⁵⁸

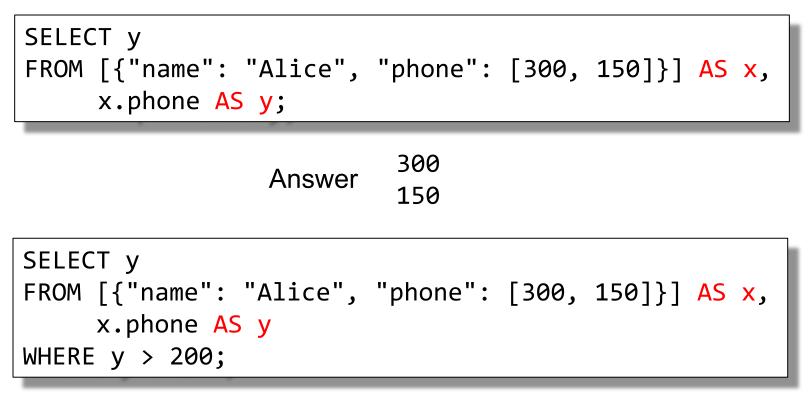
Query FROM Array / Multiset

What do these queries return?



Query Nested Collections

What do these queries return?



Query Semi-structured Data

What do these queries return?

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

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

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

Datatypes

- Boolean, integer, float (various precisions), geometry (point, line, ...), date, time, etc
- UUID = universally unique identifier
 Use it as a system-generated unique key
- Values:
 - NULL means null
 - MISSING means it's not there (see next)

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}] AS x;

Answer {"b": 2} {"b": null }

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

Finally, a language that we can use!

SELECT x.age
FROM Person AS x
WHERE x.age > 21
GROUP BY x.gender
HAVING x.salary > 10000
ORDER BY x.name;

is exactly the same as

```
FROM Person AS x
WHERE x.age > 21
GROUP BY x.gender
HAVING x.salary > 10000
SELECT x.age
ORDER BY x.name;
```

FWGHOS lives!!

Introduction to Data Management CSE 344

Lecture 13: SQL++

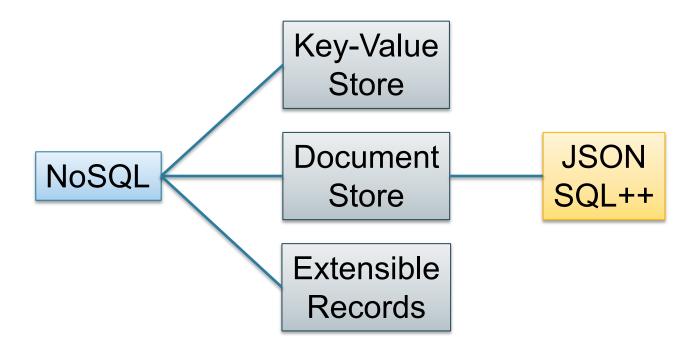
Announcements

- HW3 is due tonight!
- Midterm next Friday

 Cover material up to date
- HW4 due next Friday

Review – Big Picture

NoSQL -> Document Store -> JSON



SQL++ Overview

- Data Definition Language: create a
 - Туре
 - Dataset (like a relation)
 - Dataverse (a collection of datasets)
 - Index: for speeding up query execution
- Data Manipulation Language: SELECT-FROM-WHERE

Dataverse

A Dataverse is a Database (i.e., collection of tables)

CREATE DATAVERSE myDB CREATE DATAVERSE myDB IF NOT EXISTS

DROP DATAVERSE myDB DROP DATAVERSE myDB IF EXISTS

USE myDB

Туре

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

Closed Types

```
USE myDB;
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"}
```

Open Types

```
USE myDB;
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", "age": 20, "phone": "123456789"} 72
```

Types with Nested Collections

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

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

Types within Types

USE myDB; DROP TYPE PersonType IF EXIS CREATE TYPE PersonType AS CI	
<pre>Name : string, contact: [ContactType] }</pre>	<pre>USE myDB; DROP TYPE ContactType IF EXISTS; CREATE TYPE ContactType AS CLOSED { Method : string, Address: string }</pre>

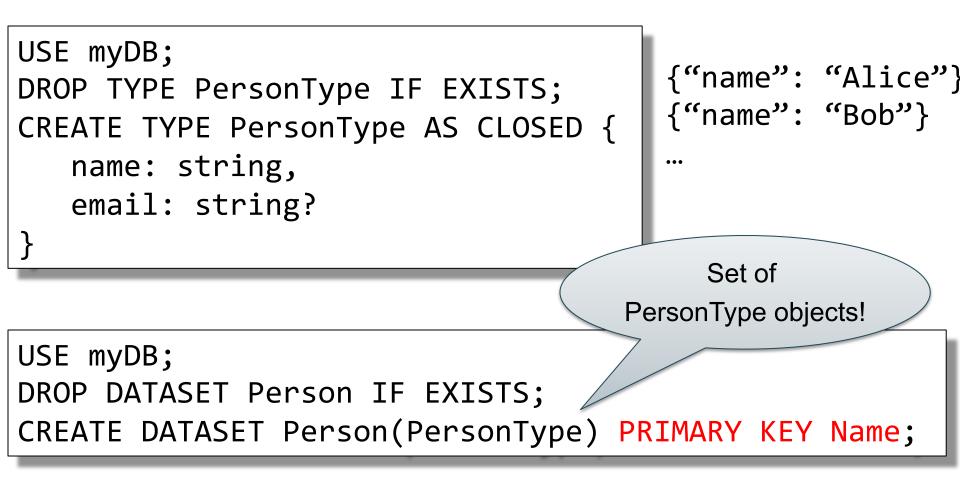
```
{"Name": "Carol", "contact": [
    {"Method": "phone", "Address": "1234"},
    {"Method": "email", "Address": "carol@uw.edu"}
]}
```

Datasets

Dataset = relation/table

- Must have a type
 - Can be a trivial OPEN type
- Must have a key
 - Can also be a trivial one

Dataset with Existing Key



Dataset with Auto Generated Key

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

```
{"name": "Alice"}
{"name": "Bob"}
```

...

```
Note: no myKey
inserted as it is
autogenerated
```

USE myDB; DROP DATASET Person IF EXISTS; CREATE DATASET Person(PersonType) PRIMARY KEY myKey AUTOGENERATED;

JSON is no longer 1NF

- NFNF = Non First Normal Form
- One or more attributes contain a collection
- One extreme: a single row with a huge, nested collection (HW5 mondial.adm)
- Better: multiple rows, reduced number of nested collections

Example from HW5

mondial.adm is totally semi-structured:

{"mondial": {"country": [...], "continent":[...], ..., "desert":[...]}}

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

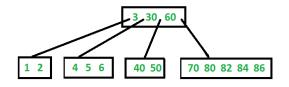
country.adm, sea.adm, mountain.adm are more structured

Country:

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

- A way to access our data (efficiently)
- Can declare an index on an **top-level type attribute**, i.e. the type used by the dataset
- Will discuss how they work later in the quarter (used to speed up queries)

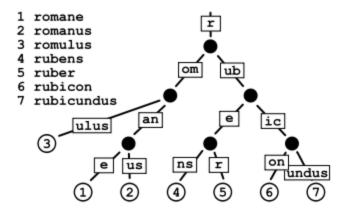
BTREE: good for equality and range queries E.g., name="Greece"; 20 < age and age < 40



RTREE: good for 2-dimensional range queries E.g., 20 < x and x < 40 and 10 < y and y < 50



KEYWORD: good for substring search if your dataset contains strings

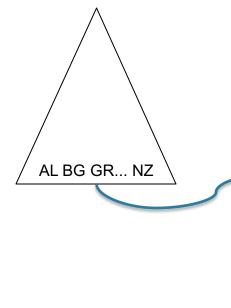


Cannot index inside a nested collection

USE myDB; CREATE INDEX countryID ON country(`-car_code`) TYPE BTREE;

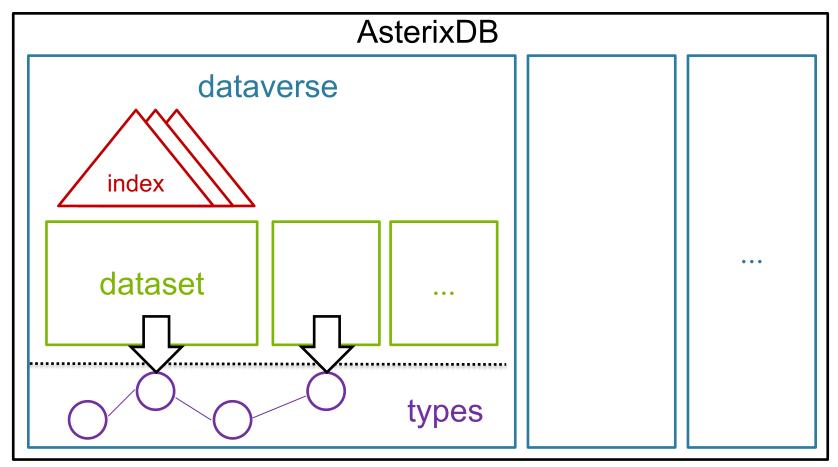
Country:

USE myDB; CREATE NDEX cityname ON country(city.name) TYPE BIREE;



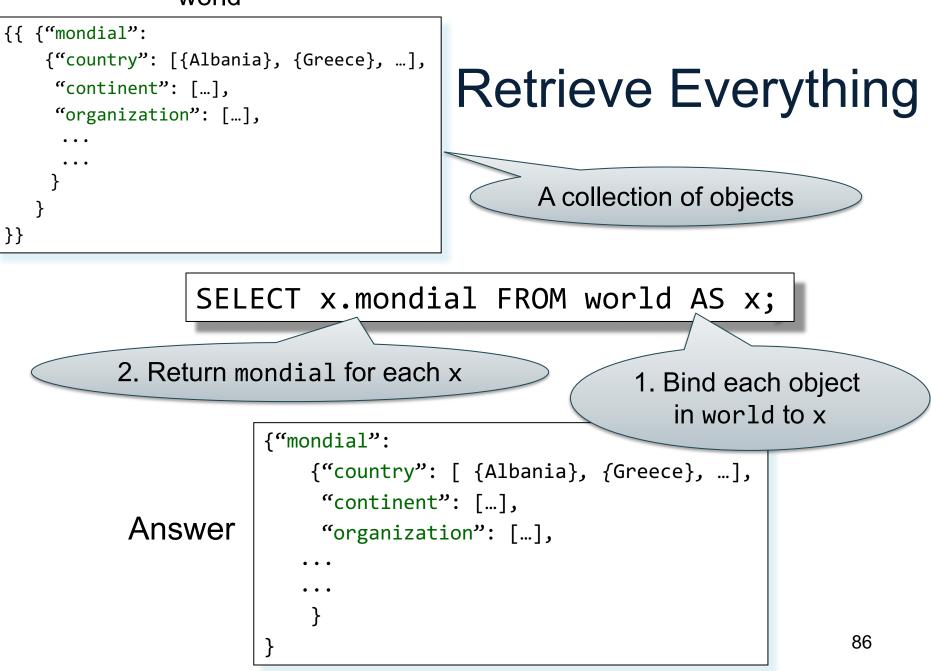
	-	 		
-car_code	name	 ethnicgroups	religions	 city
AL	Albania	 []	[]	 []
GR	Greece	 []	[]	 []
BG	Belgium			

AsterixDB Data Model Recap



SQL++ Overview

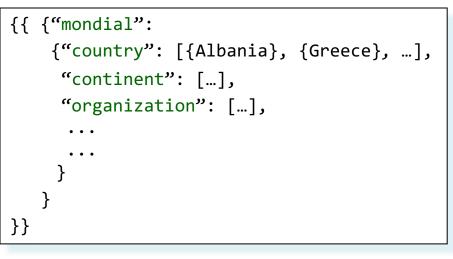
SELECT	•		
FROM			
WHERE			
GROUP BY	•	•	•
HAVING	•		
ORDER BY	•	•	•



```
{{ {"mondial":
    {"country": [{Albania}, {Greece}, ...],
        "continent": [...],
        "organization": [...],
        ...
        }
    }
}
```

Retrieve Everything

SELECT x.mondial AS ans FROM world AS x;

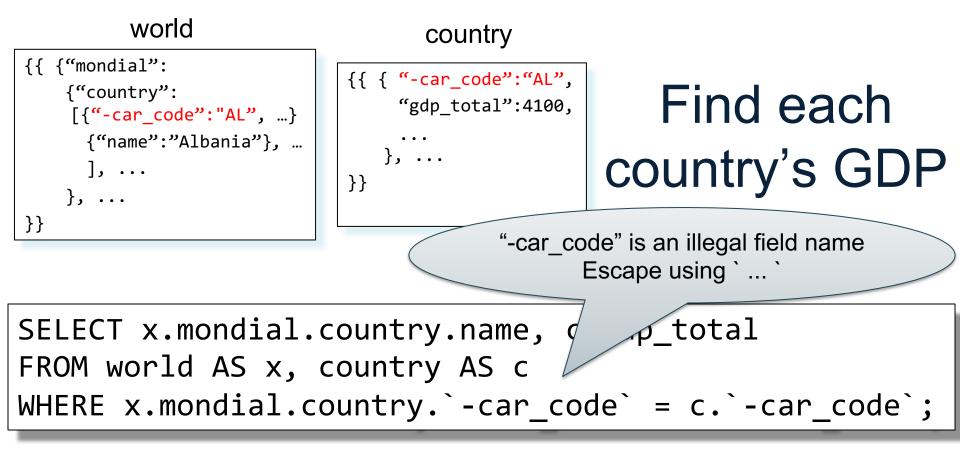


Retrieve countries

SELECT x.mondial.country FROM world AS x;

Answer

{"country": [{Albania}, {Greece}, ...]}

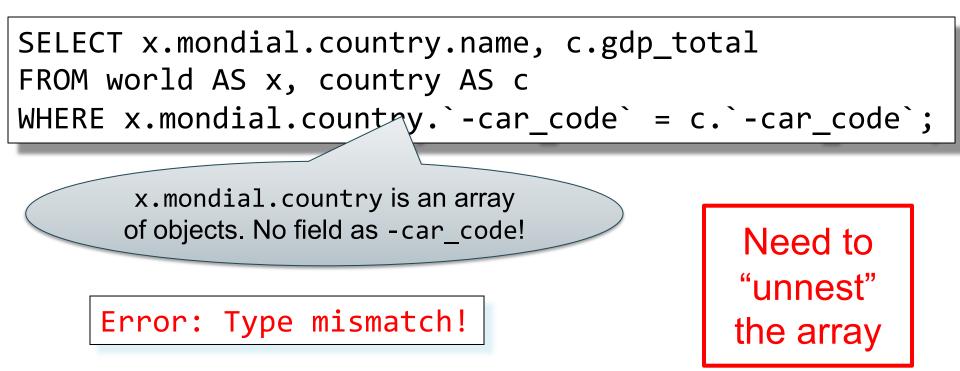




country

{{ { "-car_code":"AL", "gdp_total":4100, }, ... }}

Find each country's GDP



Unnesting collections

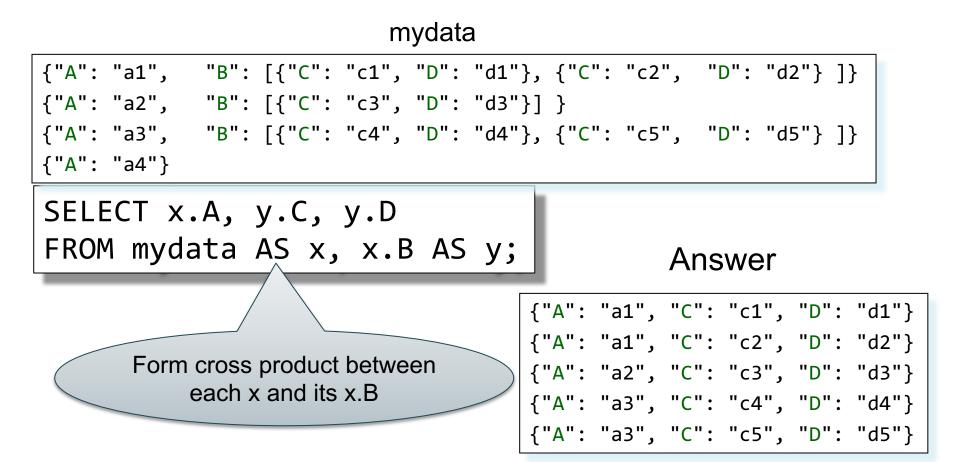
mydata

{"A": "a1", "B": [{"C": "c1", "D": "d1"}, {"C": "c2", "D": "d2"}]}
{"A": "a2", "B": [{"C": "c3", "D": "d3"}] }
{"A": "a3", "B": [{"C": "c4", "D": "d4"}, {"C": "c5", "D": "d5"}]}
{"A": "a4"}

SELECT x.A, y.C, y.D FROM mydata AS x, x.B AS y;

> Iterate over each x and bind each object in x.B to y

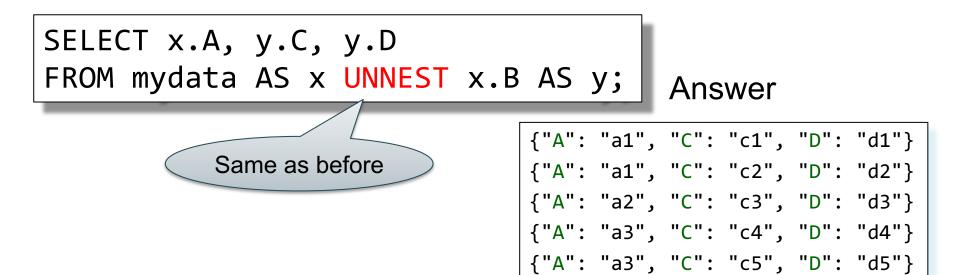
Unnesting collections

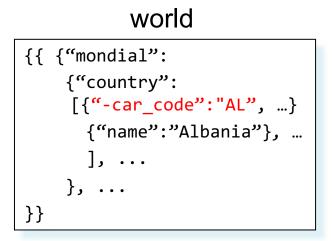


Unnesting collections

mydata

{"A": "a1",	"B": [{"C": "c1", "D": "d1"}, {"C": "c2", "D": "d2"}]}	
{"A": "a2",	"B": [{"C": "c3", "D": "d3"}] }	
{"A": "a3",	"B": [{"C": "c4", "D": "d4"}, {"C": "c5", "D": "d5"}]}	





country

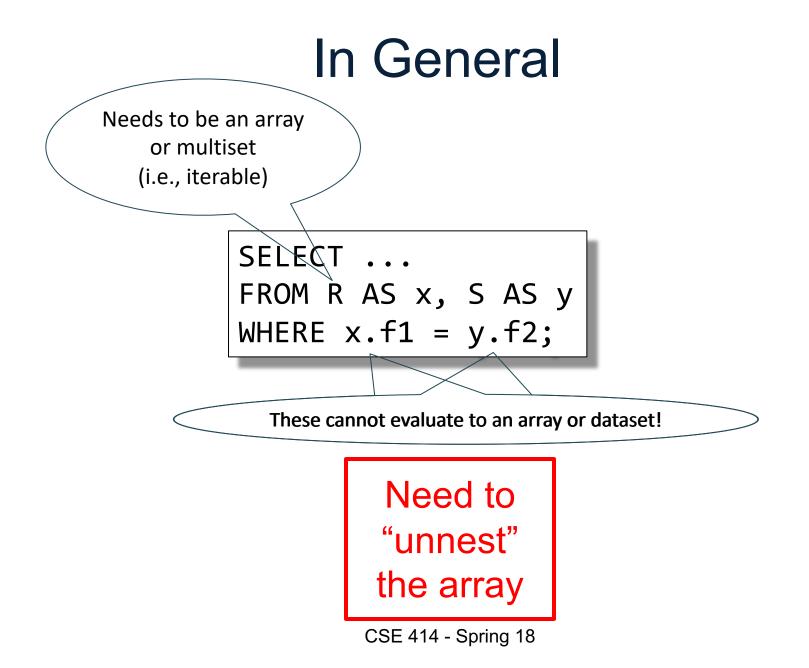
{{	<pre>{ "-car_code":"AL",</pre>
}}	 },

Find each country's GDP

SELECT y.name, c.gdp_total
FROM world AS x, x.mondial.country AS y, country AS c
WHERE y.`-car_code` = c.`-car_code`;

Answer

{ "name": "Albania", "gdp_total": "4100" }
{ "name": "Greece", "gdp_total": "101700" }
....



```
{{ {"mondial":
    {"country": [{Albania}, {Greece}, ...],
        "continent": [...],
        "organization": [...],
        ...
        ...
      }
    }
}
```

Return province and city names

(each country may have many provinces and cities)

96

SELECT z.name AS province_name, u.name AS city_name
FROM world x, x.mondial.country y, y.province z, z.city u
WHERE y.name="Greece";

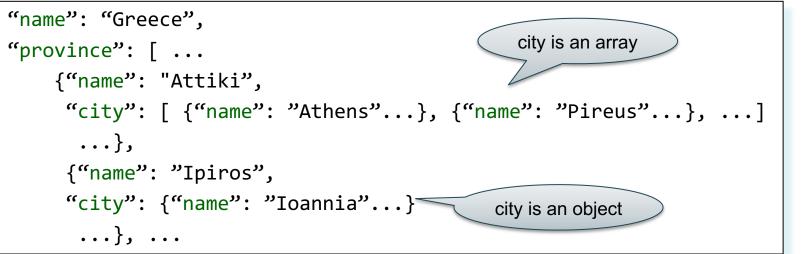
```
{{ {"mondial":
    {"country": [{Albania}, {Greece}, ...],
        "continent": [...],
        "organization": [...],
        ...
        ...
        }
    }
}
```

Return province and city names

97

SELECT z.name AS province_name, u.name AS city_name
FROM world x, x.mondial.country y, y.province z, z.city u
WHERE y.name="Greece" AND IS_ARRAY(z.city);

The problem:



```
{{ {"mondial":
    {"country": [{Albania}, {Greece}, ...],
        "continent": [...],
        "organization": [...],
        ...
        }
    }
}
```

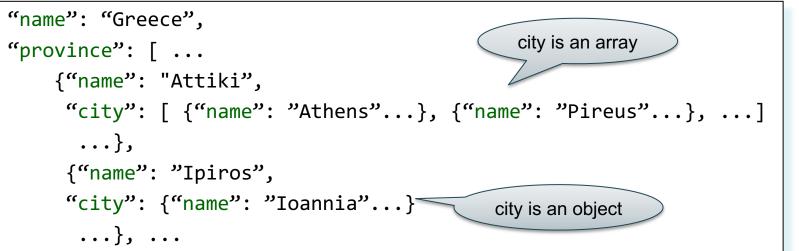
Return province and city names

Note: get name directly from z

98

SELECT z.name AS province_name, z.city AS city_name
FROM world x, x.mondial.country y, y.province z
WHERE y.name="Greece" AND NOT IS_ARRAY(z.city);

The problem:



```
{{ {"mondial":
    {"country": [{Albania}, {Greece}, ...],
        "continent": [...],
        "organization": [...],
        ...
        ...
        }
    }
}
```

Return province and city names

```
SELECT z.name AS province_name, u.name AS city_name
FROM world x, x.mondial.country AS y, y.province AS z,
   (CASE WHEN IS_ARRAY(z.city) THEN z.city
   ELSE [z.city] END) AS u
WHERE y.name="Greece";
Get both!
```

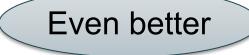
```
{{ {"mondial":
    {"country": [{Albania}, {Greece}, ...],
        "continent": [...],
        "organization": [...],
        ...
        ...
        }
    }
}
```

Return province and city names

SELECT z.name AS province_name, u.name AS city_name FROM world x, x.mondial.country y, y.province z,

(CASE WHEN z.city IS missing THEN []
WHEN IS_ARRAY(z.city) THEN z.city
ELSE [z.city] END) AS u

WHERE y.name="Greece";



Useful Functions

- is_array
- is_boolean
- is_number
- is_object
- is_string
- is_null
- is_missing
- is_unknown = is_null or is_missing

Useful Paradigms

- Unnesting
- Nesting
- Grouping and aggregate
- Joins
- Splitting
- SQL++ ⇒ SQL
 - Semistructured ⇒ Relational

Basic Unnesting

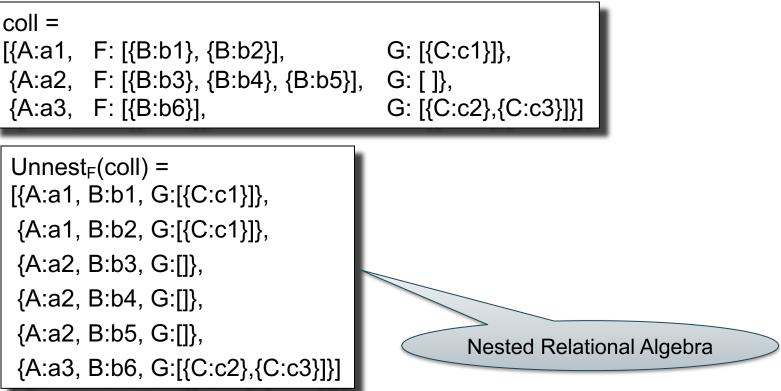
- An array: [a, b, c]
- A nested array: arr = [[a, b], [], [b, c, d]]
- Unnest(arr) = [a, b, b, c, d]



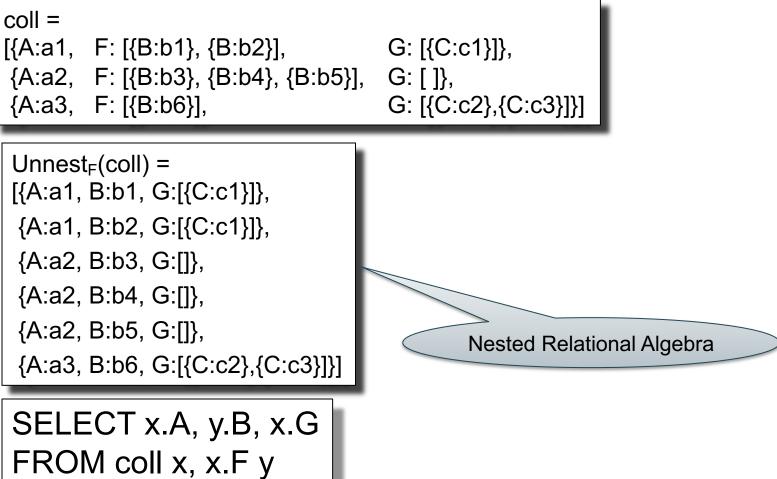
A nested collection

coll =		
[{A:a1,	F: [{B:b1}, {B:b2}],	G: [{C:c1}]},
{A:a2,	F: [{B:b3}, {B:b4}, {B:b5}],	G: []},
{A:a3,	F: [{B:b6}],	G: [{C:c2},{C:c3}]}]

A nested collection

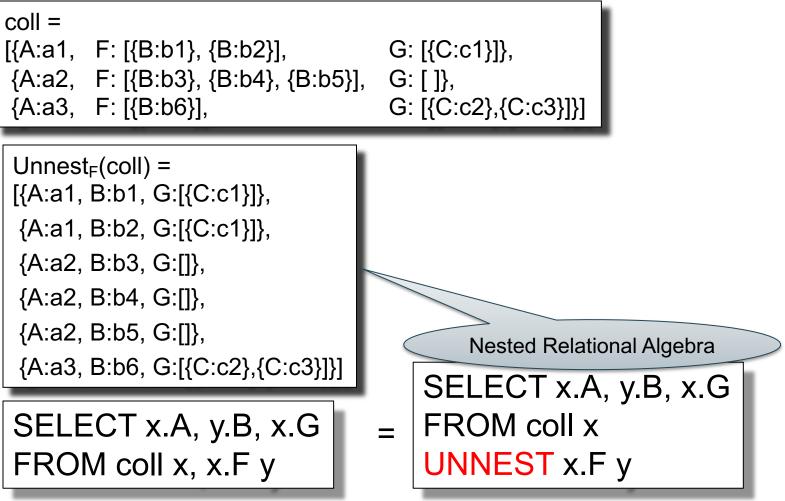


A nested collection



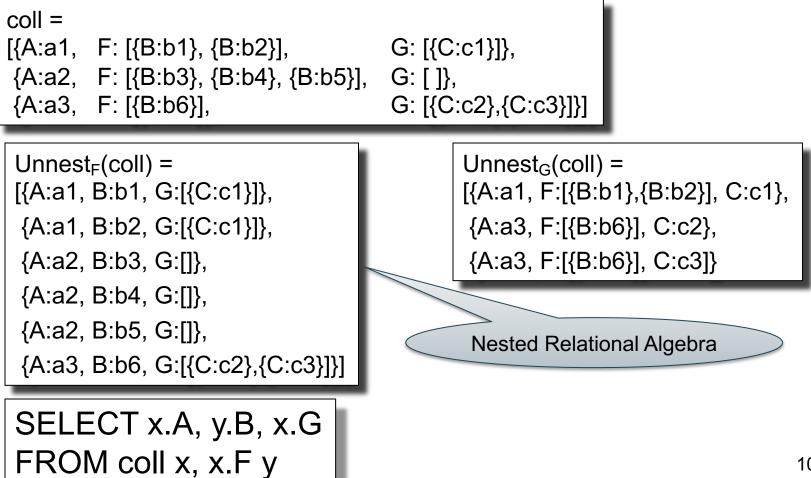
106

A nested collection



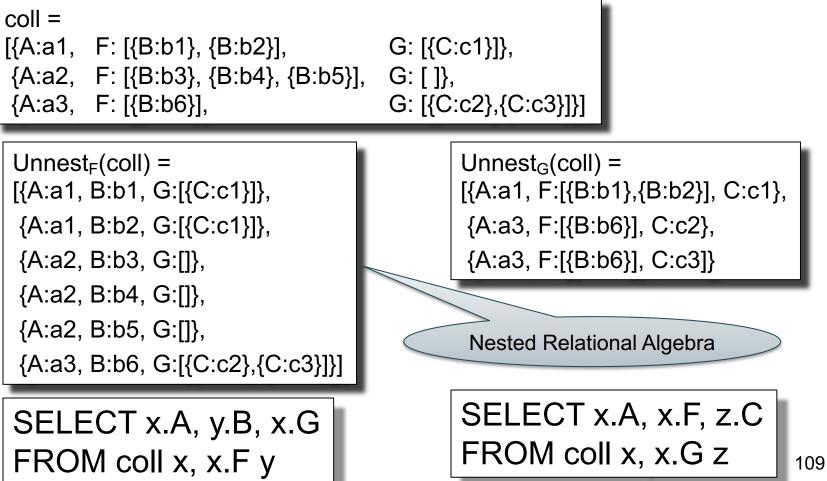
107

A nested collection

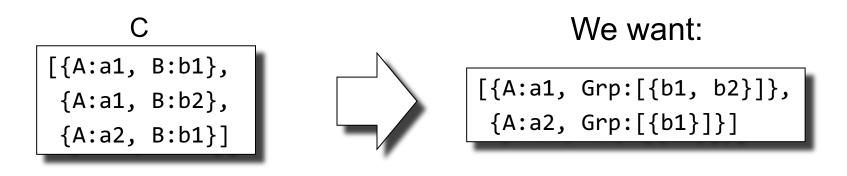


Unnesting Specific Field

A nested collection



Nesting



SELECT DISTINCT x.A, (SELECT y.B FROM C AS y WHERE x.A = y.A) AS Grp FROM C AS x

Using LET syntax:

SELECT DISTINCT x.A, g AS Grp FROM C AS x

LET g = (SELECT y.B FROM C AS y WHERE x.A = y.A)

A flat collection

coll =

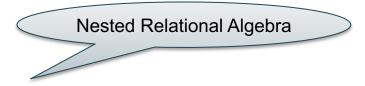
[{A:a1, B:b1}, {A:a1, B:b2}, {A:a2, B:b1}]

A flat collection

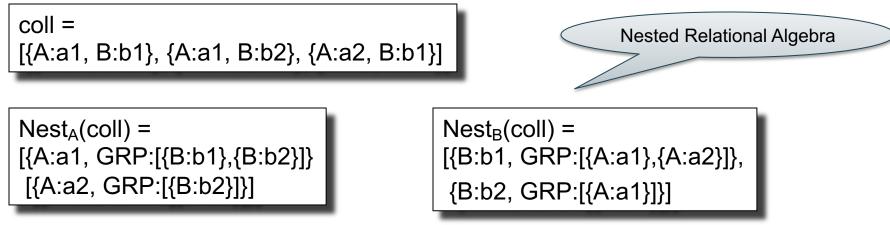
coll =

[{A:a1, B:b1}, {A:a1, B:b2}, {A:a2, B:b1}]

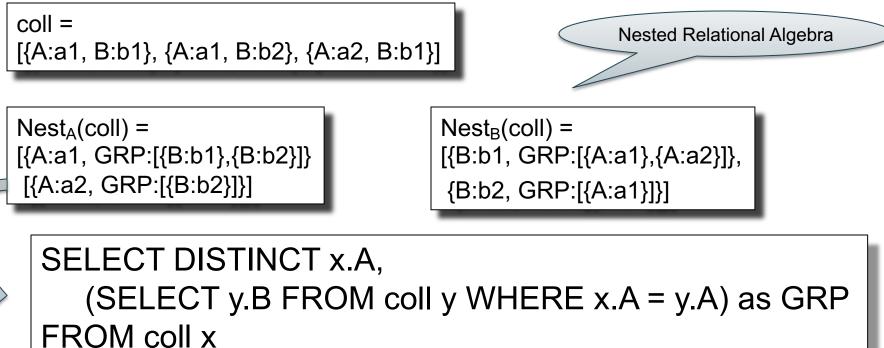
Nest_A(coll) = [{A:a1, GRP:[{B:b1},{B:b2}]} [{A:a2, GRP:[{B:b2}]}]



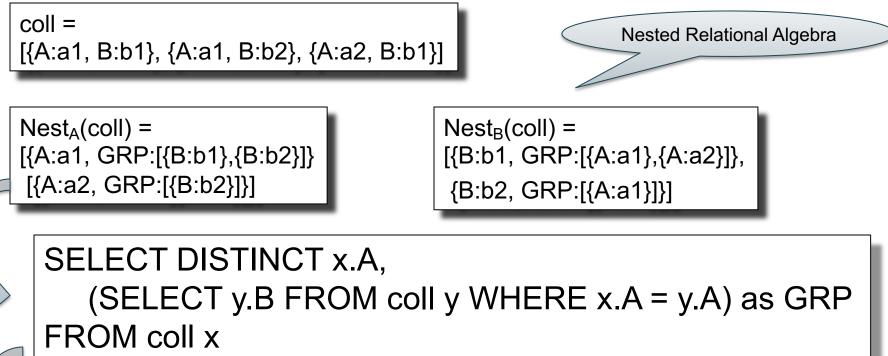
A flat collection



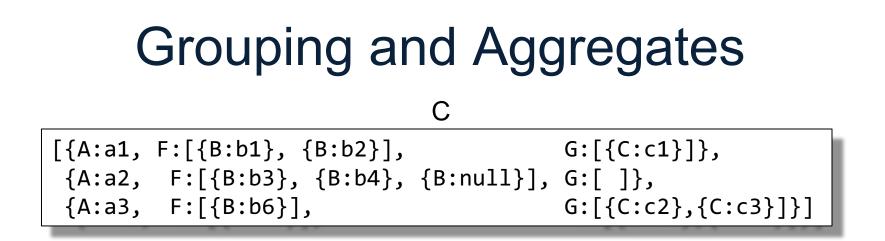
A flat collection



A flat collection



SELECT DISTINCT x.A, g as GRP FROM coll x LET g = (SELECT y.B FROM coll y WHERE x.A = y.A)



Count the number of elements in the F array for each A

SELECT x.A, strict_count(x.F) AS cnt
FROM C AS x

These are NOT equivalent! (why?)

Grouping and Aggregates

Function	NULL	MISSING	Empty Collection
STRICT_COUNT	counted	counted	0
STRICT_SUM	returns NULL	returns NULL	returns NULL
STRICT_MAX	returns NULL	returns NULL	returns NULL
STRICT_MIN	returns NULL	returns NULL	returns NULL
STRICT_AVG	returns NULL	returns NULL	returns NULL
ARRAY_COUNT	not counted	not counted	0
ARRAY_SUM	ignores NULL	ignores NULL	returns NULL
ARRAY_MAX	ignores NULL	ignores NULL	returns NULL
ARRAY_MIN	ignores NULL	ignores NULL	returns NULL
ARRAY_AVG	ignores NULL	ignores NULL	returns NULL

Joins

Two flat collection

coll1 = [{A:a1, B:b1}, {A:a1, B:b2}, {A:a2, B:b1}]
coll2 = [{B:b1, C:c1}, {B:b1, C:c2}, {B:b3, C:c3}]

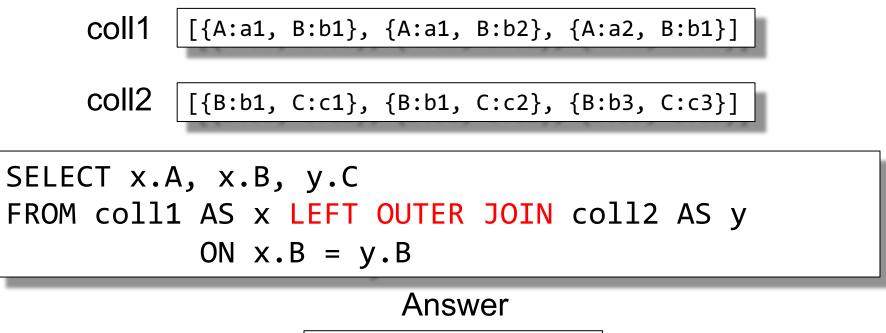
Answer

SELECT x.A, x.B, y.C FROM coll1 AS x, coll2 AS y WHERE x.B = y.B

SELECT x.A, x.B, y.C FROM coll1 AS x JOIN coll2 AS y ON x.B = y.B

Outer Joins

Two flat collection



[{A:a1,	B:b1,	C:c1},
{A:a1,	B:b1,	C:c2},
{A:a2,	B:b1,	C:c1},
{A:a2,	B:b1,	C:c2},
{A:a1,	B:b2}]

Ordering

coll1

[{A:a1, B:b1}, {A:a1, B:b2}, {A:a2, B:b1}]

SELECT x.A, x.B FROM coll AS x ORDER BY x.A

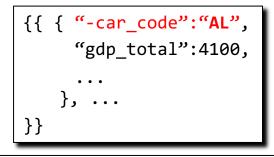
Data type matters!

"90" > "8000" but 90 < 8000 !

Splitting

- Recall: a many-to-one relation should have one foreign key, from "many" to "one"
- Sometimes people represent it in the opposite direction, from "one" to "many":
 - The reference is a string of keys separated by space
 - Need to use split(string, separator) to split it into a collection of foreign keys

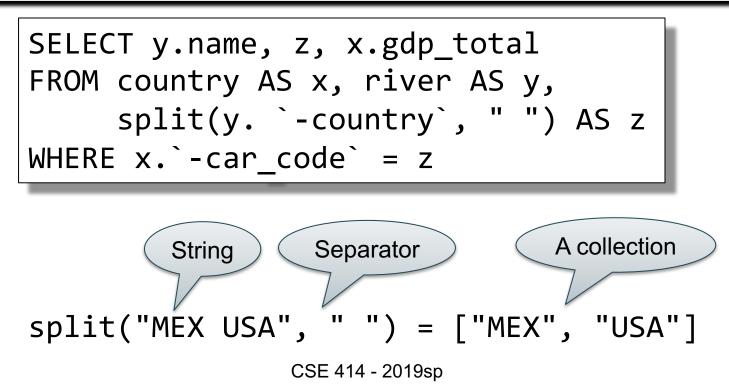
country



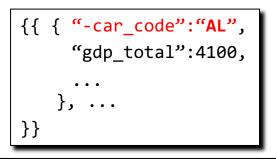
Splitting

river

[{"name": "Donau", "-country": "SRB A D H HR SK BG AL MD UA"},
 {"name": "Colorado", "-country": "MEX USA"},
 ...]



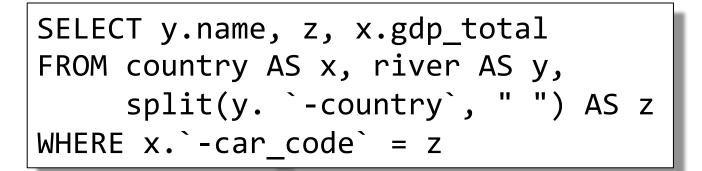
country



Splitting

river

[{"name": "Donau", "-country": "SRB A D H HR SK BG AL MD UA"},
 {"name": "Colorado", "-country": "MEX USA"},
 ...]



[{"name": "Donau", "gdp_total": 4100, "z": "AL"},
 ...]

Behind the Scenes

i.e., "How to execute SQL++ queries internally?"

Query Processing on NFNF data:

- Option 1: give up on query plans, use standard java/python-like execution
- Option 2: represent the data as a collection of flat tables, convert SQL++ to a standard relational query plan

Is it possible to (1) store nested data in flat relational form and (2) run standard relational queries over it?

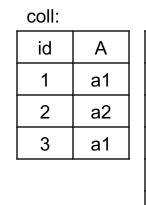
A nested collection

coll = [{A:a1, F:[{B:b1},{B:b2}], G:[{C:b1}]]}, {A:a2, F:[{B:b3},{B:b4},{B:b5}], G:[]}, {A:a1, F:[{B:b6}], G:[{C:b2},{C:b3}]]}]

A nested collection

coll = [{A:a1, F:[{B:b1},{B:b2}], G:[{C:b1}]], {A:a2, F:[{B:b3},{B:b4},{B:b5}], G:[]}, {A:a1, F:[{B:b6}], G:[{C:b2},{C:b3}]]}]

Relational representation



F	
parent	В
1	b1
1	b2
2	b3
2	b4
2	b5
3	b6

G	
parent	С
1	b1
3	b2
3	b3

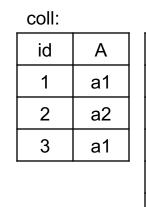
A nested collection

coll = [{A:a1, F:[{B:b1},{B:b2}], G:[{C:b1}]]}, {A:a2, F:[{B:b3},{B:b4},{B:b5}], G:[]}, {A:a1, F:[{B:b6}], G:[{C:b2},{C:b3}]]}]

SQL++

SELECT x.A, y.B FROM coll x, x.F y WHERE x.A = "a1"

Relational representation



F		_
parent	В	
1	b1	
1	b2	
2	b3	
2	b4	
2	b5	
3	b6	
		-

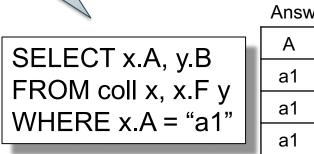
0	
parent	С
1	b1
3	b2
3	b3

G

A nested collection

coll = [{A:a1, F:[{B:b1},{B:b2}], G:[{C:b1}]]}, {A:a2, F:[{B:b3},{B:b4},{B:b5}], G:[]}, {A:a1, F:[{B:b6}], G:[{C:b2},{C:b3}]]]

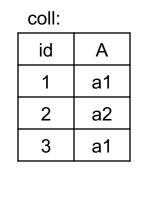
SQL++



|--|

А	В	
a1	b1	
a1	b2	
a1	b6	

Relational representation



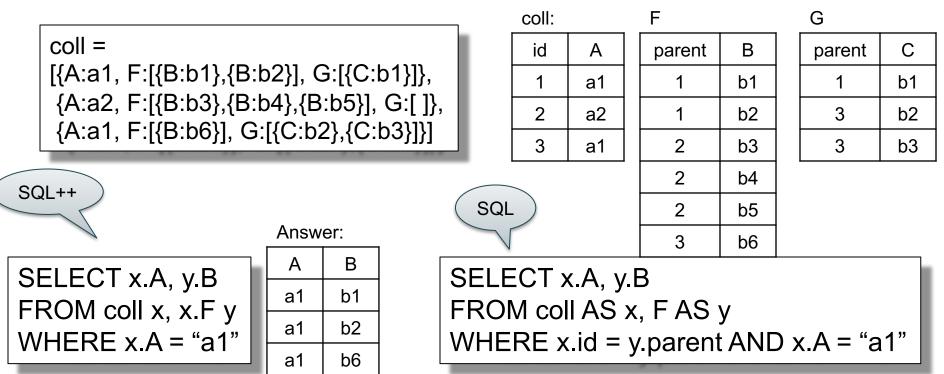
F	_
parent	В
1	b1
1	b2
2	b3
2	b4
2	b5
3	b6

0	
parent	С
1	b1
3	b2
3	b3

G

A nested collection

Relational representation



A nested collection

Relational representation

В

b1

b2

b3

b4

b5

b6

F

parent

1

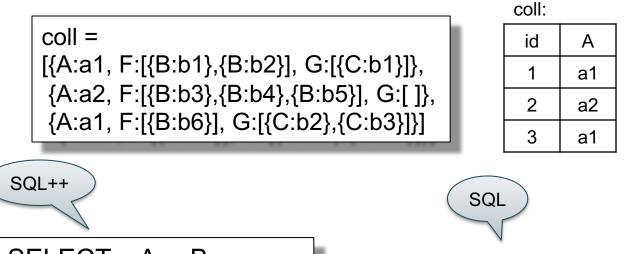
1

2

2

2

3

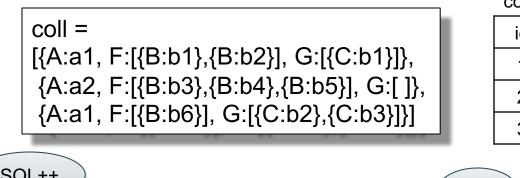


G	
parent	С
1	b1
3	b2
3	b3

SELECT x.A, y.B FROM coll x, x.F y, x.G z WHERE y.B = z.C

A nested collection

Relational representation



_coll:	
id	А
1	a1
2	a2
3	a1

F	
parent	В
1	b1
1	b2
2	b3
2	b4
2	b5
3	b6

0	
parent	С
1	b1
3	b2
3	b3

G

SQL++

SQL

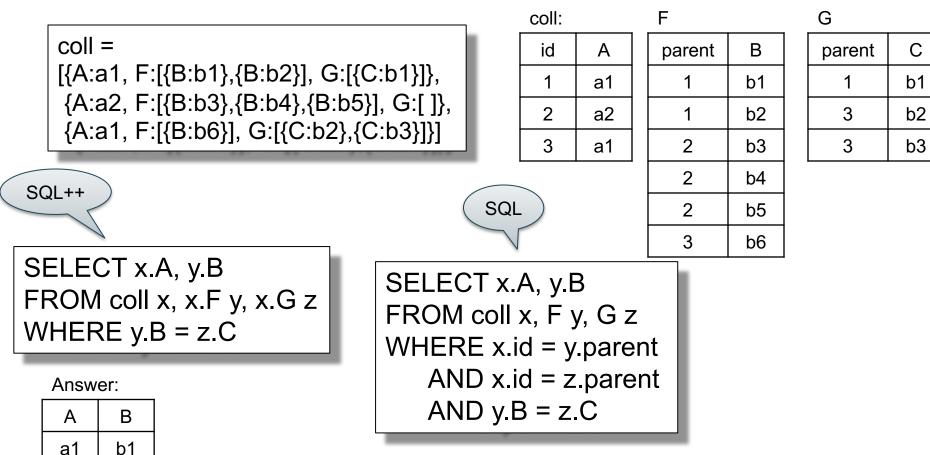
SELECT x.A, y.B FROM coll x, x.F y, x.G z WHERE y.B = z.C

Answer:

А	В
a1	b1

A nested collection

Relational representation



Semistructured Data Model

- Several file formats: JSON, protobuf, XML
- Data model = Tree
- Differ in how they handle structure:
 - Open or closed
 - Ordered or unordered
- Query language take NFNF into account
 - Various "extra" constructs introduced as a result
 - Nesting & Unnesting, strict aggregates, splitting

Conclusion

Semi-structured data: best for *data exchange*

"General" guidelines:

- For quick, ad-hoc data analysis, query it directly in the native format (Json/SQL++)
- Modern, advanced query processors like AsterixDB can process semi-structured data as efficiently as RDBMS
- For long term data analysis: spend the time and effort to normalize it, then store in a RDBMS