

Graph stores

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M2 Data and Knowledge
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Motivation

- Graphs correspond to a natural organization of knowledge
- They generalize
 - Relations
 - Trees (documents)
 - Key-value pairs ...
- Graph stores simplify / facilitate data representation
- They do not simplify query evaluation (and may make it more complex)

Graph database models

- Graph = N (nodes) and E (edges, subset of ExE)
- Directed vs. undirected edges
- Nodes:
 - Unlabeled
 - With a single label (in some cases called *type*)
 - With a set of attribute-value pairs
 - With complex internal structure (persistent objects)
- Graphs may have semantics (RDF, RDFS)

Object-oriented databases

- 1980 – 2000 (approx)
- Idea: capitalize on the flexibility of OO programming languages such as C++ and Java to handle databases of persistent objects
- Object Database Management Group (ODMG): consortium of OODB vendors which produced a standard
 1. Object Model // classes, attributes, methods...
 2. Object Definition Language (ODL)
// persistency roots (persistent collections)
 3. Object Query Language (OQL)
// navigation from one object to its attribute
// method invocation
// structured query language
 4. C++ and Java Bindings

Sample OQL queries

- **select** a.number
from a in ATM_MACHINE.accounts_list
where a.balance > 0
- **select** max(**select** c.age **from** p.children c) // nested queries
from Persons p
where p.name = "Paul"
- **select** p.oldest_child.address.street
from Persons p
where p.lives_in("Paris") // method invocation
- **select** ((Student)p).grade // explicit type test
from Persons p
where "course of study" in p.activities // set attribute

Where are OODBs now?

- Object-oriented extensions are present in all major (relational) databases → Object-Relational Database Management Systems (ORDBMS)
 - Mostly relational
 - Modest but useful object extensions
- E.g. complex types in Postgres:
 - **create type** `inventory_item` **as** (name text, supplier_id integer, price numeric);
 - **create table** `on_hand` (item `inventory_item`, count integer);
 - **insert into** `on_hand` **values** (`ROW('fuzzy dice', 42, 1.99), 1000`);

Working with composite type in the Postgres ORDBMS

```

create type inventory_item as ( name text, supplier_id integer,
price numeric );
create table on_hand (item inventory_item, count integer );
select (on_hand.item).name // ( ) specific to composite type
from on_hand
where (on_hand.item).price > 9.99;

create type complex as (r double precision, i double precision );
insert into mytab (complex_col) values ((1.1,2.2));
update mytab set complex_col = row(1.1,2.2) where ...;

```

This would have required a join in a classical RDBMS!

The first (graph) semistructured data model: OEM [PGW95]

OEM: Object Exchange Model, introduced as a global data model for mediator systems

E.g. GAV scenario where several product databases are integrated under a unique global schema

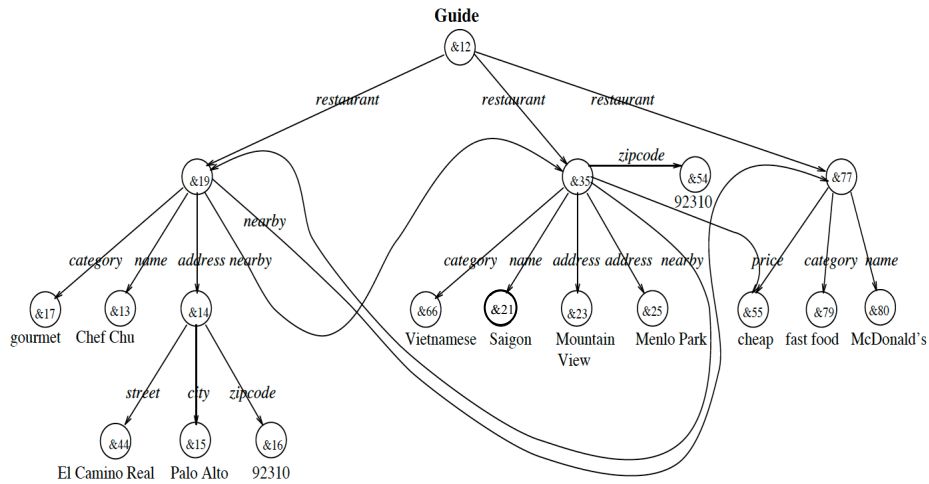
- Some have one **price**, some have several (e.g. price reductions)
- Some have a **description**, some have a **technical_description**, some have **description.text**, **description.price**...
- Some have a **photo**, some don't

OEM: Labeled, directed, unordered **graph of objects**

Every **object** has a unique **identity**
 Every **edge** has a **direction** and a **label**
 Atomic object = **value** (simple atomic type)
 No (a priori) schema

Semistructured data: the data has internal structure (as opposed to a BLOB) but the structure is not regular, some parts may be more structured than others

A restaurant OEM database



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9

Restaurant database, serialized

Guide &12

```

restaurant &19
  category &17 "gourmet"
  name &13 "Chef Chu"
  address &14
    street &44 "El Camino Real"
    city &15 "Palo Alto"
    zipcode &16 92310
  nearby_eating_place &35
  nearby_eating_place &77
    
```

```

restaurant &35
  category &66 "Vietnamese"
  name &21 "Saigon"
  address &23 "Mountain View"
  address &25 "Menlo Park"
  nearby_eating_place &19
  zipcode &54 "92310"
  price &55 "cheap"
restaurant &77
  category &79 "fast food"
  name &80 "McDonald's"
  price &55
    
```

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10

Querying OEM data with LOREL [AQH+97]

Semistructured database principle: no query should fail; query evaluation should adapt gracefully

select Guide.restaurant.address

where Guide.restaurant.address.zipcode=92310

Guide is a *persistence root* (name starts with a capital)

Empty results if expected labels are not found

Tries to convert zipcode to an integer; also accepts strings

select Guide.restaurant.name,
Guide.restaurant.(address?).zipcode

where Guide.restaurant.% grep "cheap"

Address is optional; "cheap" can occur anywhere in the restaurant object

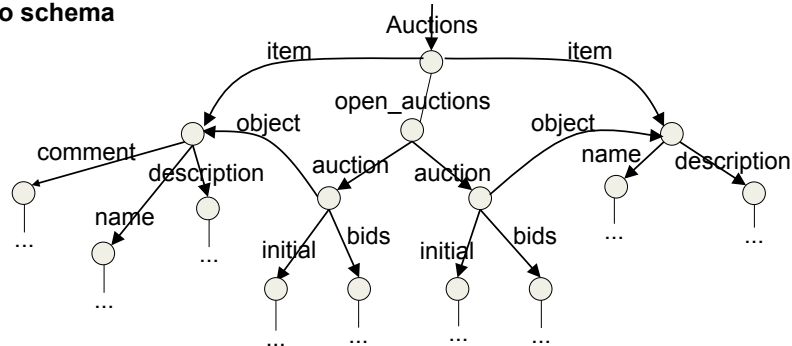
The first (graph) semistructured data model: OEM [PGW95]

Semistructured data: the data has internal structure (opposed to e.g. unstructured *text* or *blob* – *Binary Large Object*) but the structure is not regular

Some items have comments/bids, others do not

One description may be just text, another one have complex structure

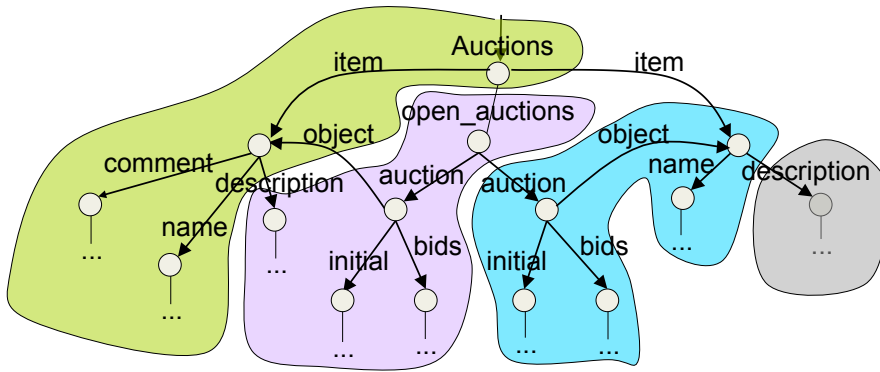
No schema



Storing OEM objects in LORE [MAG+97]

Objects clustered in pages in depth-first order, including simple value leaves

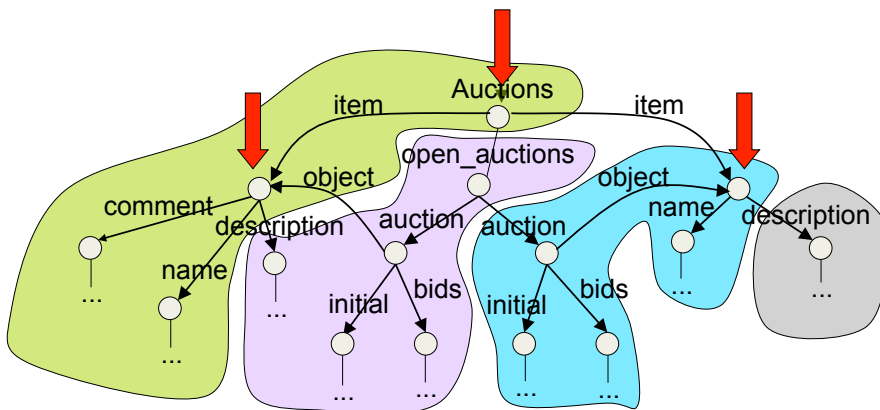
Basic physical operator: **Scan(obj, path)**



Navigation in a persistent graph

Navigation-based scan implementation (aka tuple-at-a-time, pointer-chasing)

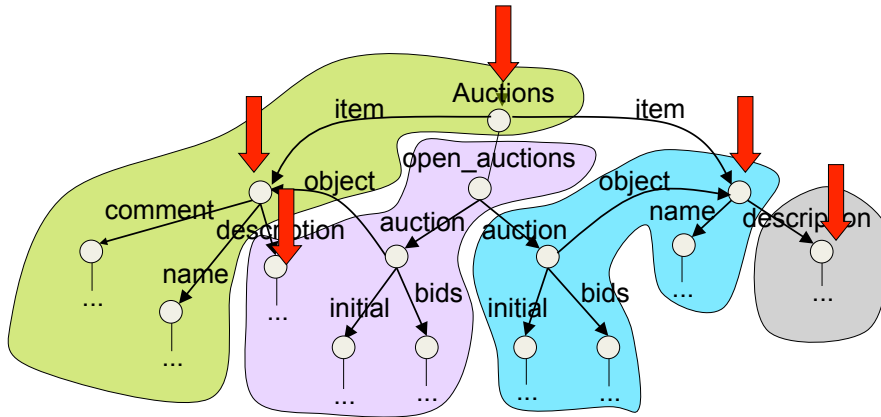
Scan(Auctions, "item"): 2 pages accessed



Navigation in a persistent graph

Scan(Auctions, "item.description"): 4 pages accessed

Scan(Auctions, "open_auctions.auction.object"): 4 pages accessed



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15

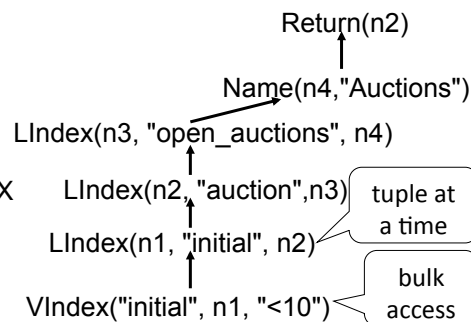
Indexing objects in a graph [MW97, MWA+98, MW99]

VIndex(I, o, pred): all objects *o* with an incoming *I*-edge, satisfying **pred**

LIndex(o, I, p): all parents of *o* via an *I*-edge
– "Reverse pointers"

BIndex(x, I, y): all edges labeled *I*

```
select X
from Auction.open_auctions.auction X
where X.initial < 10
```



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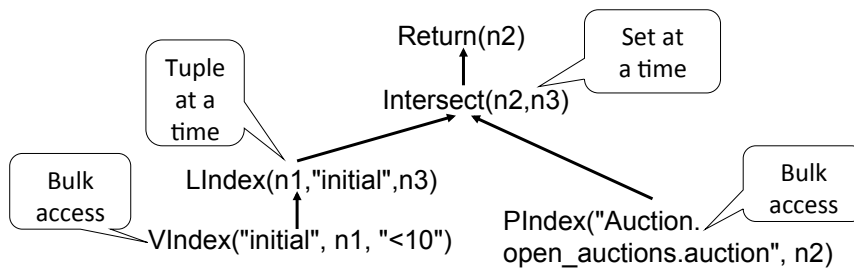
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16

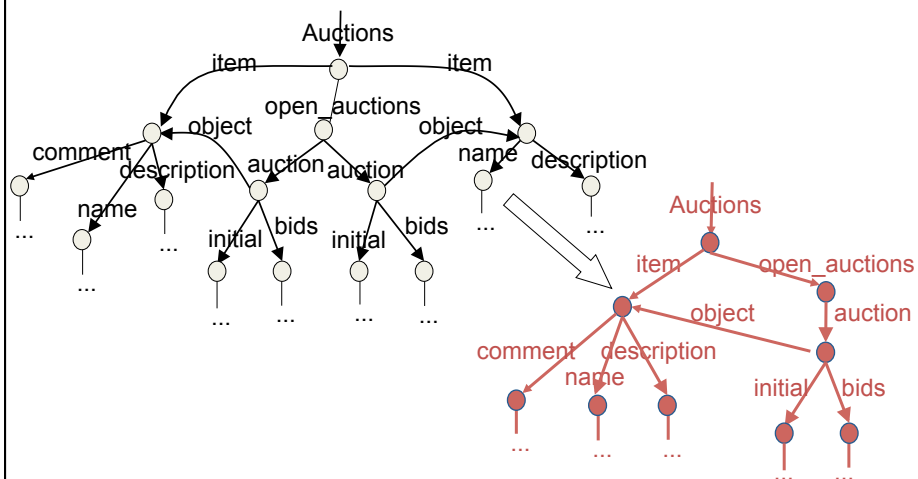
Indexing objects in a graph [MW97]

$PIndex(p, o)$: all objects o reachable by the path p

```
select X
from Auction.open_auctions.auction.initial X
where X.initial < 10
```



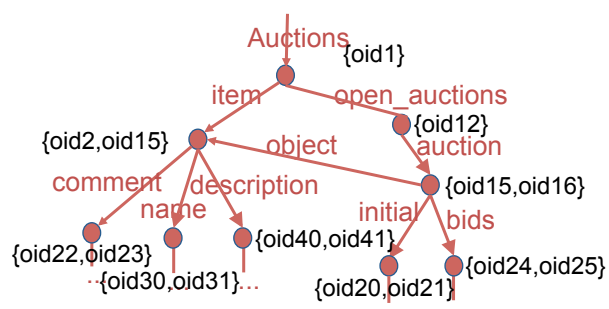
The idea behind path indexes: DataGuides [GW97]



The idea behind path indexes: DataGuides [GW97]

Graph-shaped summaries of graph data

- "A-posteriori schema"
- Groups all nodes *reachable by the same paths*



More on graph indexing

Graph indexing:

1. Partition nodes into **equivalence classes**
2. Store the **extent** of each equivalence class, use it as "pre-cooked" answer to some queries

Equivalence notions:

1. Reachable by some common paths: *DataGuide* [MW97]
2. Reachable by exactly the same paths: *1-index* [MS99] or, equivalently, indistinguishable by any forward path expression
3. Indistinguishable by any (forward and backward) path expression: *F&B Index* [KBN+02]
4. Indistinguishable by the (forward and backward) path expressions in the set Q: *covering index* [KBN+02]
5. Indistinguishable by any path expression of length $< k$: *A(k) index* [KSB+02]

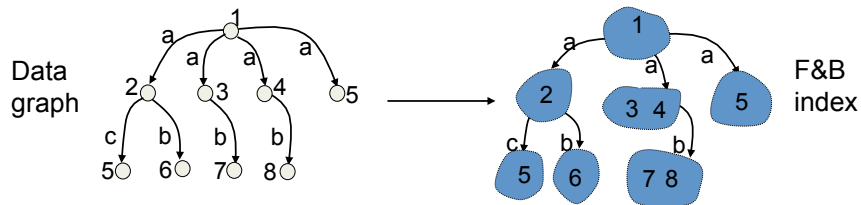
F&B index

Group together nodes reachable by exactly the same paths

Path language:

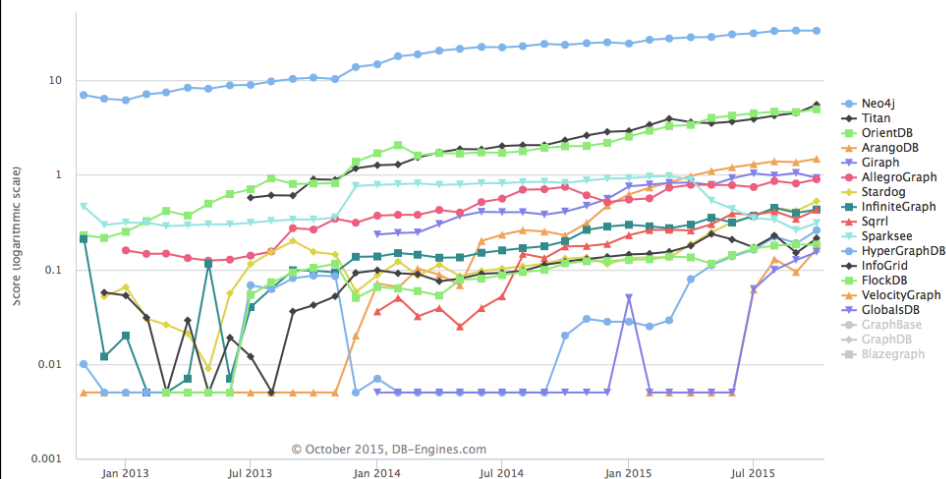
- Navigate along one edge in both directions
- Navigate along any number of edges, in both directions

$n1 \sim n2$: for any path expression p , either $n1$ and $n2$ are in the answer of p , or neither are in the answer of p .



Current graph stores

DB-Engines Ranking of Graph DBMS



Neo4J basics

Data model: labeled, directed graphs

Data manipulation language (CRUD): **Cypher**, used to describe data and patterns to be matched

Node descriptions in Cypher:

```
( ) // empty anonymous node
(matrix) // node whose identifier is matrix.
(:Movie) // node of type Movie
(matrix:Movie) // node whose ID is matrix and type Movie
(matrix:Movie {title: "The Matrix"})
// node with an attribute
(matrix:Movie {title: "The Matrix", released: 1997})
// node with two attributes
```

Identifiers can be used to refer to this node in another place in the **same** statement

Strings vs. integers

Identifiers are not stored in the database (they are related to "variables")

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Neo4J basics

Relationship descriptions in Cypher

-- (undirected) vs. --> or <-- (directed)

Sample relationship descriptions:

```
-->
-[role]-> // relationship ID
-[:ACTED_IN]-> // relationship type
-[role:ACTED_IN]->
-[role:ACTED_IN {roles: ["Neo"]}]-> // relationship with
attributes
```

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24

Data manipulation with Cypher

Patterns combine node and relationship descriptors:

```
(keanu:Person:Actor {name: "Keanu Reeves"})
-[role:ACTED_IN {roles: ["Neo"]} ]-> (matrix:Movie {title: "The
Matrix"})
```

Data **creation**:

```
CREATE (a:Person { name:"Tom Hanks", born:1956 })
      -[r:ACTED_IN { roles: ["Forrest"]}]->
      (m:Movie { title:"Forrest Gump",released:1994 })
CREATE (d:Person { name:"Robert Zemeckis", born:1951 })
      -[:DIRECTED]->(m)
```

Data manipulation with Cypher

Querying data: **MATCH** pattern **RETURN** matched variables

```
MATCH (p:Person { name:"Tom Hanks" })
      -[r:ACTED_IN]->(m:Movie)
RETURN m.title, r.roles
```

Successive match-create-return steps can be used to update the data:

```
MATCH (p:Person { name:"Tom Hanks" })
CREATE (m:Movie { title:"Cloud Atlas",released:2012 })
CREATE (p)-[r:ACTED_IN { roles: ['Zachry']}]->(m)
RETURN p,r,m
```

Data manipulation with Cypher

Inserting data only if it didn't exist:

```
MERGE (m:Movie { title:"Cloud Atlas" })
      // create or check the existence of movie node m
ON CREATE SET m.released = 2012
      // if we had to create it, set the release year

RETURN m
```

Insert relationship only if it did not exist:

```
MATCH (m:Movie { title:"Cloud Atlas" })
MATCH (p:Person { name:"Tom Hanks" })
MERGE (p)-[r:ACTED_IN]->(m)
ON CREATE SET r.roles =['Zachry']

RETURN p,r,m
```

Returning results with Cypher

```
MATCH (a { name: "A" })-[r]->(b)
RETURN *
```

a	b	r
Node[0]{name:"A",happy:"Yes!",age:55}	Node[1]{name:"B"}	:BLOCKS[1]{}
Node[0]{name:"A",happy:"Yes!",age:55}	Node[1]{name:"B"}	:KNOWS[0]{}

```
MATCH (n)
RETURN n.age // returns null if no age
```

```
MATCH (a { name: "A" })
RETURN a.age > 30, "I'm a literal", (a)-->()
```

Edge
creation
(ability to
return new
graphs)

Other Cypher operations

- Booleans:
MATCH (n)
WHERE n.name = 'Peter' XOR (n.age < 30 AND n.name = "Tobias") OR NOT (n.name = "Tobias" OR n.name="Peter")
RETURN n
- Optional matching
- Returned data can be: ordered, truncated, aggregated
- Unwind: unfolds a collection into a set
UNWIND[1,2,3] AS x **RETURN** x // three results
- Indexes: **CREATE INDEX ON** :Person(name)
- **EXPLAIN** to get the query plan
- **PROFILE** to measure the effort

Richer path specification in SPARQL

- RDF: W3C standard for semantic Web data (graphs)
 - Nodes are labeled with URIs or constants
 - Edges are labeled with URIs
- SPARQL: query language for RDF data
- SPARQL 1.1 provides rich property path descriptions (think regular expressions):
<http://www.w3.org/TR/sparql11-query/#propertypaths>

```
{ :book1 dc:title|rdfs:label ?displayString }
{ ?x foaf:mbox <mailto:alice@example> .
  ?x foaf:knows/foaf:name ?name . }
{ ?x foaf:knows/^foaf:knows ?y . FILTER(?x != ?y) }
{ ?ancestor (ex:motherOf|ex:fatherOf)+ ?me }
```

Graph stores: summary

- Graph databases repeatedly "attempted" but not fully "solved" yet
- Very convenient data model, natural representation
- Typically no strict schema
- No standard query language
- Semantic graphs are a particular case (RDF and SPARQL *are* standards)
- Most powerful tools around: *distributed* graph stores (Pregel, Spark Graphix)
 - Extra dimension: graph partitioning
 - Less effort on query language; in progress

References

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