Graph Databases

Guillaume Raschia — Nantes Université Last update: October 17, 2023

Graph DB Landscape

Contents

Graph DB Landscape

Graph Data Models

Graph Queries

"Old School" Graph Database Style

Model

1

2

A graph G(V, E) is a binary relation R(src, dst)



Queries

• Relational Algebra (**procedural** language): $\{\sigma, \pi, \bowtie, \rho, \cup, -\}$

• 3-hops: $R \underset{\text{dst}1=\text{src}2}{\bowtie} R \underset{\text{dst}2=\text{src}3}{\bowtie} R$

Limitations

- Join 🛛 is the key operator: costly!
- Reachability: $\bigcup_k R \Join_1 R \Join_2 \ldots \Join_k R$: Recursion and fixpoint

Graph DB

s∩eo4j

The #1 Database for Connected Data

Graph Data Models

Graph DB (cont'd)



Requirements for Graph Databases

The 3D graph data model [Angles et al., ACM CS 2008]

- 1. Data structure
 - data and schema as (distinct) graphs
 - standard abstractions: is-a, is-part-of, is-associated-to
- 2. Update and query language
 - graph transformations
 - primitives on paths, neighborhoods, subgraphs, graph patterns, connectivity and graph statistics (diameter, centrality, etc.)
 - multi-relational graph algorithms
- 3. Integrity constraints

5

• schema-instance consistency, identity, referential integrity

Requirements for Graph Databases (cont'd)

Definition (Graph database (tentative of))

Any storage system that provides index-free adjacency

- Each vertex has direct references to its adjacent vertices
 act as a mini-index
- $\cdot \mathcal{O}(1)$ to move from a vertex to its neighbors
- $\mathcal{O}(\log n)$ b.t.w. of an index in non-graph db's

Graph Partitioning

Usually for a particular (set of) operation(s)

• The shortest path, finding frequent patterns, BFS, spanning tree search, ...

Graph partitioning is counter-intuitive and graph DB are mainly centralized

1-Dimensional Partitioning

- Partitioning of the adjacency matrix
- Focus on Breadth-First Search

Graph DB and NoSQL

Very large graphs such like TAO Social Graph at Facebook: 5 Billions+ nodes!

Bronson et al. (2013). TAO: Facebook's Distributed Data Store for the Social Graph. USENIX ATC.

Audrey Cheng et al. (2021). RAMP-TAO: Layering Atomic Transactions on Facebook's Online TAO Data Store. PVLDB 14(12): 3014-3027.

Abstract: Facebook's graph store TAO, like many other distributed data stores, traditionally prioritizes availability, efficiency, and scalability over storing consistency or isolation guarantees to serve its large, read-dominant workloads. As product developers build diverse applications on top of this system, they increasingly seek transactional semantics. However, providing advanced features for select applications while preserving the system's overall reliability and performance is a continual challenge. In this paper, we first characterize developer desires for transactions that have emerged over the years and describe the current failure-atomic (i.e., write) transactions offered by TAO. We then explore how to introduce an intuitive read transaction API. We highlight the need for atomic visibility guarantees in this API with a measurement study on potential anomalies that occur without stronger isolation for reads. Our analysis shows that 1 in 1,500 batched reads reflects partial transactional updates, which complicate the developer experience and lead to unexpected results. In response to our findings, we present the RAMP-TAO protocol, a variation based on the Read Atomic Multi-Parition (RAMP) protocols that can be feasibly deployed in production with minimal overhead while ensuring atomic visibility for a read-optimized workload at scale.

1D Partitioning

	1	2	3	4	5	6	7	8	9	10	11	12
1	0	0	0	0	0	0	0	0	0	1	1	0
2	0	0	1	0	0	0	0	1	0	0	0	0
3	0	1	0	0	0	0	1	1	0	0	0	0
4	0	0	0	0	1	0	0	0	0	0	0	1
5	0	0	0	1	0	0	0	0	0	0	0	1
6	0	0	0	0	0	0	1	0	0	0	1	0
7	0	0	1	0	0	1	0	1	0	1	1	0
8	0	1	1	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0	1	1
10	1	0	0	0	0	0	1	0	0	0	1	0
11	1	0	0	0	0	1	1	0	1	1	0	0
12	0	0	0	1	1	0	0	0	1	0	0	0

- Matrix rows are randomly assigned to the nodes (processors)
- Each vertex and the edges emanating from it are owned by one processor

8

1D Partitioning (cont'd)

BSF with 1D partitioning

- 1. Each processor has a set of frontier vertices F
- 2. The neighbors of the vertices in F form a set of neighbouring vertices N
 - $\cdot\,$ Some owned by the current processor, some by others
- 3. Messages are sent to all "neighbouring processors", etc.
- 1D partitioning yields to high messaging
- Then, 2D-partitioning of adjacency matrix
 - lower messaging but still very demanding...

Efficient sharding of a graph is a hard problem

Labeled Property Graph Example

Node labels are "groups" whereas edge labels denote compulsory relationships



Source: R. Bouhali & A. Laurent (2015). Exploiting RDF Open Data Using NoSQL Graph Databases. AIAI pp.177-190.

Dominant and Alternative Models of Graph DB's

Data model: Labeled Property Graph

A (attributed) multi-relational labeled digraph G(V, E) where

- V is a finite set of nodes (id's)
- · $E \subseteq V \times \Sigma \times V$ are directed edges –**relationship**
- + Σ is a finite alphabet of **labels** assigned to both nodes and edges
- Optional attributes **properties** may apply to both vertices and edges as a set of key/value pairs

Extensions to

- hypernode: nested graphs
- hypergraph: with hyperedges, i.e., sets of nodes
- multigraph: multiple single-relational edges

12

11

The Web of Data

Resource Description Framework (RDF) Graph, aka. Knowledge Graph



Source: C. Sayers & A.H. Karp (2004). Computing the digest of an RDF graph.

- Data Model for the Semantic Web
- RDF statement: (subject, predicate, object)
- Triple store is a "flavor" of graph db

The Web of Data (cont'd)

RDF Data Model: Edge-Labeled Digraph

Vertex set V is split into URIs (U), literals (L), and blank/anonymous nodes (B), such that:

 $E \subseteq ((U \times B) \times U \times (U \times B \times L))$

Extension to named graphs as ng = (n, g) with $n \in U$ and $g \in G$, the family of RDF graphs

15

16

Labeled Property Graph vs. Edge-Labeled Digraph



Another Graph DB Example



Graph Queries

Conjunctive Queries

Implements subgraph matching

Definition (Graph CQ's)

$$Q(\vec{z}) \leftarrow \bigwedge_{1 \le i \le m} (x_i, a_i, y_i)$$

where x_i , y_i are node variables or constants, each z_j is x_i , y_i or any constant, and $a_i \in \Sigma$

Example

 $Q(x) \leftarrow (x, hasWon, Nobel), (x, hasWon, Booker), (x, bornIn, SouthAfrica)$

Extended CRPQ's

Paths may

- $\cdot\,$ occur in the output by means of free path variables
- \cdot be compared within a regular relation

Examples

Retrieve every path between nodes r and s that go through node e:

$$Q(\pi_1, \pi_2) \leftarrow (r, \pi_1, e), (e, \pi_2, s)$$

Retrieve all pairs (x, y) connected by paths following pattern $a^n b^n$:

$$Q(x,y) \leftarrow (x,\pi_1,z), (z,\pi_2,y), a^*(\pi_1), b^*(\pi_2), (\bigcup_{a,b\in\Sigma} (a,b))^*(\pi_1,\pi_2)$$

20

Conjunctive Regular Path Queries

Implements **reachability** by path expressions

Definition (Graph CRPQ's)

$$Q(\vec{z}) \leftarrow \bigwedge_{1 \le i \le m} (x_i, r_i, y_i)$$

Extend CQ's to r_i as a **regular expression** over Σ

Example

$$Q(x) \leftarrow (x, hasWon, Booker), (x, r, Australia)$$

 $r := citizenOf | ((bornIn | livesIn).locatedIn^*)$

Extended CRPQ's (cont'd)

Definition (Graph ECRPQ's)

$$Q(\vec{z},\vec{\chi}) \leftarrow \bigwedge_{1 \le i \le m} (x_i, \pi_i, y_i), \bigwedge_{1 \le j \le p} R_j^{k_j}(\vec{\omega_j})$$

where χ_ℓ , π_i , ω_{jt} are path variables, and $R_j^{k_j}$ is a regular expression that defines a regular relation over Σ

Extension to approximate matching and ranking

- Define an **edit distance** $d_e(x, y)$ on Σ^*
- Operate with a regular expression over triples of the form (a, k, b), $a, b \in \Sigma \cup \{\epsilon\}$, k the cost of substitution

19

Aggregation and Arithmetic Predicates

count sum min max + - * / for computing:

- degree, eccentricity of a node
- distance between two nodes
- diameter of the graph
- etc.

Example

Length of the *shortest path* between each pair of nodes

22

Querying Knowledge Graphs

SPARQL

Graph pattern-based SQL-like query language for RDF triple stores

PREFIX foaf: <http://xmlns.com/foaf/0.1/>
SELECT ?name ?email
WHERE {
 ?person a foaf:Person.
 ?person foaf:name ?name.
 ?person foaf:mbox ?email.

}

The Jungle of Languages for Property Graphs

No(t yet any) Standard

Alternatives

- Native API in generic programming language
- depends on PL and Graph DB
- Procedural—algebraic-based—language
- Declarative—logic-based—language

depends on Graph DB

depends on Graph DB

24

Neo4J Native Java API

Node person = matching.getSingle(); Iterable<Relationship> relations = startNode.getRelationships(Direction.OUTGOING, RelTypes.FRIEND); for (Relationship rel : relations) { Node friend = rel.getEndNode(); String name = friend.getProperty("name");

}

Source et planches suivantes: F. Holzschuher and R. Peinl, EDBT/ICDT 2013

23

Gremlin



t = new Table();

x = []; g.idx("persons")[[id:id_param]].out("FRIEND_OF").fill(x); g.idx("persons")[[id:id_param]].out("FRIEND_OF"). out("FRIEND_OF").dedup().except(x).id.as("ID"). back(1).displayName.as("name"). table(t,["ID","name"]){it}{it}.iterate();

26

SQL

Neo4j Cypher

Cypher Query Language (CQL)

START person=node:people(id = {id})
MATCH person-[:FRIEND_OF]->friend-[:FRIEND_OF]->friend_of_friend
WHERE not (friend_of_friend<-[:FRIEND_OF]-person)
RETURN friend_of_friend, COUNT(*)
ORDER BY COUNT(*) DESC</pre>

Neo Technology released CQL to open source: see OpenCypher Project

Performance Comparison

FROM person persondb0 WHERE persondb0.oid IN (

);

SELECT persondb0.ID, persondb0.display name

FROM person persondb1, friend frienddb2

WHERE persondb1.oid=frienddb2.person_id AND

SELECT frienddb2.friend_id

(persondb1.person id IN (?))



27

Cypher Use Case

User story

As an employee,

I want to know who in the company has similar skills to me

So that we can exchange knowledge

Query

Which people, who work for the same company as me, have similar skills to me?

30

Query as a Graph Pattern



Path Patterns in Cypher

Running the Query

+-----

2 rows

| name | score | skills

| "Bill" | 1 | ["Neo4j"]

| "Lucy" | 2 | ["Java","Neo4j"] |

----+

Examples



(x)--(y) (m:MOVIE)-->(a:ACTOR) (:MOVIE)-->(a { name: "Ivan Trojan" }) ()<-[r:PLAY]-() (m)-[:PLAY { role: "Ivan" }]->() (:ACTOR { name: "Ivan Trojan" })-[:KNOW *2]->(:ACTOR) ()-[:KNOW *5..]->(f)





Towards a Standard Declarative Language for Graph DB's

Graph Query Language (GQL) is a Global ISO/IEC Standards Project alongside SQL¹ • gqlstandards.org Last Update: October 4th, 2023 • A. Deutsch et al. (2022). Graph Pattern Matching in GQL and SQL/PGQ. SIGMOD Alastair Green, Query Languages Standards & Research Lead, Neo4j Sep 16, 2019 · 4 mins read The votes are in. This past June, national standards bodies around the world - belonging to ISO/IEC's Joint Technical Committee 1 (responsible for IT standards) – began voting on the GQL project proposal. Graph Query Language (GQL) is a new language being developed and maintained by the same international working group that also maintains the SQL standard. Now, it's official: Earlier this week, the ballot closed and the proposal passed, with seven countries putting experts forward to work on the four-year standard query language project. ¹The very 1st since SQL itself! 34

GQL (cont'd)



GQL

