

DuckPGQ:

Efficient Property Graph Queries in an analytical RDBMS

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Outline

1. the why and what of SQL/PGQ

2. competent graph database systems architecture

3. graph query processing in DuckDB



Graph data management





tables often represent graphs



Graph data management





tables often represent graphs







SELECT count(*) **FROM** person WHERE name LIKE 'E%'

7

7

relational operators



Storing graphs in SQL

```
CREATE TABLE city (
                                                                    :person
                                                                   name: Bob
  id bigint PRIMARY KEY,
  name varchar
                                                          :person
                                                                             :person
);
                                                         name: Chloe
                                                                            name: Jack
CREATE TABLE person (
                                                       follows
                                                                    :person
                                                                                          :city
  id bigint PRIMARY KEY,
                                                                  name: Emily
                                                                                      name: Utrecht
  name varchar,
                                                                              livesIn
  livesIn bigint,
  CONSTRAINT c FOREIGN KEY (livesIn) REFERENCES city (id)
);
CREATE TABLE follows (
  p1id bigint,
  p2id bigint,
  CONSTRAINT p1 FOREIGN KEY (p1id) REFERENCES person (id),
  CONSTRAINT p2 FOREIGN KEY (p2id) REFERENCES person (id)
);
```





SQL:1999 query

```
WITH RECURSIVE paths(startNode, endNode, path) AS (
  SELECT plid AS startNode, p2id AS endNode, ARRAY[p1id, p2id] AS path
    FROM follows JOIN person p1 ON p1.id = follows.p1id WHERE p1.name = 'Bob'
  UNION ALL (
    WITH paths AS (TABLE paths)
       SELECT paths.startNode AS startNode, p2id AS endNode, array_append(path, p2id) AS path
       FROM paths JOIN follows ON paths.endNode = follows.p1id
       WHERE NOT EXISTS (SELECT true FROM paths previous_paths
                         JOIN person p2 ON p2.id = follows.p2id
                          WHERE p2.name = 'Bob' OR follows.p2id = previous_paths.endNode)))
SELECT count(p2.id) AS cp2
FROM person p1
JOIN paths ON paths.startNode = p1.id
JOIN person p2 ON p2.id = paths.endNode
JOIN city ON city.id = p2.livesIn AND city.name = 'Utrecht'
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```



Graph query languages







nGQL

SPARQL

Gremlin

Oracle Labs PGX



GSQL



Cypher

PGQL





The (sorry) State of Graph Database Systems

Peter Boncz CWI

comparing graph with relational database systems + provide pointers to related literature



The Sorry State of Graph Database Systems

"The six blunders of graph database systems" (see keynote)

- time may be running out for native property graph database systems
 - Some success in certain use cases: Data Integration, Data cleaning & Enrichment, Fraud Detection, Recommendation, Historical Analysis, Root-Cause Analysis,...
 - still a niche solution and maturity+usability problems remain
- especially if SQL/PGQ becomes a (moderate) success
 - Relational systems will be able to handle their use cases
 - Only Data Integration, Data cleaning & Enrichment would be left (RDF/SPARQL territory)

SQL/PGQ (Property Graph Queries)





SQL/PGQ

- Extension in the upcoming SQL:2023 standard, 2b released in June
- Property Graphs as views over existing tables
 - edge,vertex=table, property (value) =column (value), label=table-name
- Read-only operations for property graph queries
 - Path-finding + Pattern matching in Cypher-like syntax, producing a "Graph-Table" in FROM



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Tabular schema

CREATE TABLE city (id bigint **PRIMARY KEY**, name varchar); **CREATE TABLE** person (id bigint **PRIMARY KEY**, name varchar, livesIn bigint, **CONSTRAINT** c FOREIGN KEY ...); **CREATE TABLE** follows (p1id bigint, p2id bigint, CONSTRAINT p1 FOREIGN KEY ... CONSTRAINT p2 FOREIGN KEY ...);

SQL/PGQ graph tables

```
CREATE PROPERTY GRAPH socialNetwork
VERTEX TABLES (
    city,
    person
)
EDGE TABLES (
    livesIn SOURCE person DESTINATION city,
    follows SOURCE person DESTINATION person
);
```





SQL/PGQ query

"count the number of people Bob (in)directly follows who live in the city Utrecht"

```
SELECT count(gt.id)
FROM
```

```
GRAPH_TABLE (socialNetwork,
```

MATCH (p1:person WHERE p1.name='Bob')-[:follows]->*(p2:person)
 -[:livesIn]->(c:city WHERE c.name='Utrecht')

```
COLUMNS (p2.id)
```

) gt

DuckPGQ module for DuckDB





DuckDB

- open-source in-process SQL OLAP DBMS
- Created by Mark Raasveldt

& Hannes Mühleisen (keynote Wednesday)

- very popular in data science notebooks, but suitable for many analytics applications
- "Modern": Vectorized execution engine, Morsel-driven parallelism, ...
- Allows extension modules:
 - scalar user-defined functions (UDF), parser extensions
 - data sources (scans), table-returning functions

DuckDB crossed 1M/month PyPI downloads by 2023!!







Current DuckPGQ pipeline



Base tables Graph view



Current DuckPGQ pipeline





Path finding: Compressed Sparse Row (CSR)

- **On-the-fly** creation (no update handling needed)
- Using **scalar UDFs** (parallel, very fast)
- Index in the vertex array corresponds to the ROWID of the vertex
- Vertex array contains offsets for the edge arrays



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10

11

12



- Batched variant developed by Manuel Then
 - Works like regular BFS, but starts from multiple nodes
- Share the memory access
 - Major bottleneck
 - Can make use of SIMD instructions (SSE/AVX)

VLDB'14 The More the Merrier: Efficient Multi-Source Graph Traversal

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ABSTRACT

Graph analytics on social networks, Web data, and communication networks has been widely used in a plethon of breadth-first search (BFS) graph traversal, which is not only time-consuming for large datasets but also involves much redundant computation when executed multiple times from different start vertices. In this paper, we propose Multi-Source BFS (MAS-BFS), an algorithm that is designed to have influence on others and, as a consequence, are of great importance to spread information, e.g., for marketing purposes [20].

In a wide range of graph analytics algorithms, including shortest path computation [13], graph centrality calculation [9, 27], and k-hop neighborhood detection [12], breadthfirst search (BFS)-based graph traversal is an elementary building block used to systematically traverse a graph, i.e., to visit all reachable vertices and edges of the graph from a





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Last Slide



Conclusion

- Why should you read our DuckPGQ paper?
 - **Learn SQL/PGQ** in less than 1 page (or become ldbcouncil.org member & read 200+ pages of spec)
 - Read our **12 golden rules** of competent graph systems design (just 1 page of reading)
 - See how DuckDB extensibility can be leveraged for a modular **implementation of SQL/PGQ**

(..and we also present some benchmark results..)

- DuckPGQ availability? Not yet.. WIP & ETA in 2023
- Many avenues for future data systems research :
 - Factorized query execution, Vectorized WCOJs & their query optimization
 - Path-finding and query optimization, better path-finding parallelism