Query Languages for Graph Databases

Peter T. Wood

School of Computer Science and Information Systems Birkbeck, University of London ptw@dcs.bbk.ac.uk



Third Alberto Mendelzon International Workshop on Foundations of Data Management

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background

Graphs

Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Motivation

Graphs are widely used for representing data

- transportation and other networks
- geographical information
- semistructured data
- (hyper)document structure
- semantic associations in criminal investigations
- bibliographic citation analysis
- pathways in biological processes
- knowledge representation (e.g. semantic web)
- program analysis
- workflow systems
- data provenance

▶ ...

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Example

A graph of cities and flight durations:



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Example

A graph of cities and flight durations:



Nodes

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Example

A graph of cities and flight durations:



Edges

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Types of Edges



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Types of Edges



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Types of labels

Node labels

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Types of labels



Edge labels



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Undirected A B D C

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking





Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking





Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Acyclic graphs

Tree B D C

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Acyclic graphs



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Acyclic graphs

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs

Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Formal graph definition

For our purposes:

- database comprises a single labelled (multi-)graph G
- (finite) set of nodes N with identifiers drawn from an infinite vocabulary V
- (finite) set of (directed) edges E
- ► incidence function \(\phi\) : \(E\) \(\mathcal{N}\) × \(N\) (allows multi-edges)
- edge labelling function $\lambda : \boldsymbol{E} \mapsto \boldsymbol{\Sigma}$
- Σ is a finite alphabet

So $G = (N, E, V, \Sigma, \phi, \lambda)$

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background

Graphs Personal interest

Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interest

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Many years ago ...

- PhD on "Queries on Graphs" (1988)
- supervised by Alberto Mendelzon

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interest

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Many years ago ...

- PhD on "Queries on Graphs" (1988)
- supervised by Alberto Mendelzon

More recently

- querying RDF (allowing for query relaxation and ranking)
- ranking approximate answers to semantic web queries
- investigating operators for finding/manipulating paths
- ... with Pablo Barcelo and Carlos Hurtado (Chile) and Alex Poulovassilis (Birkbeck)

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interest Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Transportation and other networks

- airline, train, bus ... networks
- communication networks
- planning networks—single source and sink, acyclic

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Transportation and other networks

- airline, train, bus ... networks
- communication networks
- planning networks—single source and sink, acyclic

Typical queries:

- reachability: can I get from a to b?
- shortest path: find the quickest/shortest route from a to b
- reliability/capacity of paths
- critical path

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Knowledge representation

- semantic networks
- conceptual graphs
- RDF/S, OWL
- ontologies
- taxonomies
- • •

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

Knowledge representation

- semantic networks
- conceptual graphs
- RDF/S, OWL
- ontologies
- taxonomies

▶ ...

Typical queries:

- instance and subclass relationships
- finding connections between entities

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Program/workflow analysis

- nodes are program points or agents/products
- edges are program or workflow steps
- often single source and sink nodes
- also data provenance applications

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Program/workflow analysis

- nodes are program points or agents/products
- edges are program or workflow steps
- often single source and sink nodes
- also data provenance applications

Typical queries:

- reachability of code
- variables used before defined
- deadlock/livelock
- what agents/processes/products were involved in producing something

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Biological applications

metabolic pathways

> . . .

- gene regulatory networks
- protein interaction networks

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Biological applications

- metabolic pathways
- gene regulatory networks
- protein interaction networks

Typical queries include:

path existence

▶ ...

▶ ...

- subgraph isomorphism
- k-shortest paths
- neighbourhood queries
- approximate matching

(see https://hpcrd.lbl.gov/staff/olken/
graphdm/graphdm.htm)

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

ackground Graphs Personal intere Uses of graphs

Graph models and query languages

G, G^+ and Graphlog

Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog Lore/Lorel

YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

G, G^+ and Graphlog

- from the 1980s [Cruz, Mendelzon and Wood, 1987]
 [Cruz, Mendelzon and Wood, 1988]
 [Consens and Mendelzon, 1989]
- developed at University of Toronto
- data model is a labelled, directed graph
- in G and G⁺, query is a set of pairs of pattern graphs and summary graphs
- pattern graph nodes are labelled with variables or constants
- pattern graph edges are labelled with regular expressions over edge labels and variables
- Graphlog adds edge inversion, negation, distinguished edge and different semantics

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

G, G⁺ Example

- given a graph
 - nodes representing people
 - edges labelled with *m* (for *motherOf*) and *f* (for *fatherOf*)

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

G, G⁺ Example

- given a graph
 - nodes representing people
 - edges labelled with *m* (for *motherOf*) and *f* (for *fatherOf*)
- following query finds parents followed by pairs of people who have a common ancestor



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, **G⁺** and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

verview Sraph pattern matching Path finding Edge and path variables Aggregation Approximate matching and anking

Outline

Background Graphs Personal intere Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and language

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog

YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking
Lore/Lorel

- from the 1990s [Abiteboul et al., 1997]
- developed at Stanford
- Lore: Lightweight Object Repository
- Lorel: Lore query language
- for semistructured data
 - no predefined schema
 - may be heterogeneous
- uses Object Exchange Model (OEM)
- Lore/Lorel can be viewed as extension of ODMG model/OQL

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog

YAGO/NAGA Other models an

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Lore model

- data model is graph with two types of nodes
 - complex objects
 - atomic objects (values) with no outgoing edges
- each node has a unique oid
- each edge is labelled with a string
- graph has a number of named nodes (entry points)
- every node must be reachable from a named node

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog

YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Lorel example

- graph representing a restaurant guide
- find addresses of restaurants with a given zipcode select Guide.restaurant.address where Guide.restaurant.address.zipcode = 92310
- Guide is a named node
- restaurant, address and zipcode are edge labels

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog

YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background Graphs Personal inter Uses of graph

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages *G*, *G*⁺ and Graphlog

ore/Lorel

YAGO/NAGA

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

YAGO/NAGA

- from the 2000s [Weikum et al., 2009]
- developed at Max Planck Institute for Informatics
- YAGO: Yet Another Great Ontology
- NAGA: Not Another Google Answer
- semantic search engine for web derived knowledge
- combines DB and IR
- 26 relationships between entities derived using information extraction
 - ▶ e.g., isA, bornInYear, hasWonPrize, locatedIn, ...

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages G, G⁺ and Graphlog

.ore/Lorel

YAGO/NAGA

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

NAGA model

- data model is directed, weighted multigraph
- nodes represent entities
- edges represent relationships
- weights represent confidence of extracted facts
- query is a connected, directed graph
- each edge labelled with a regular expression over edge labels or a variable or *connect* keyword
- answers are ranked by
 - informativeness
 - confidence
 - compactness

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages **G**, **G⁺** and Graphlog

.ore/Lorel

YAGO/NAGA

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

NAGA examples

- graph representing information on people and films
- in which films did a governor act?
 - X isA governor X actedIn Y Y isA film
- X and Y are node variables
- isA and actedIn are relationships (edge labels)
- what do Albert Einstein and Niels Bohr have in common?

Albert_Einstein connect Niels_Bohr

- Albert_Einstein and Niels_Bohr are node labels
- asks for paths connecting nodes—ranked

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages *G*, *G*⁺ and Graphlog

YAGO/NAGA Other models ar

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background Graphs Personal inter Uses of graph

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Other graph data models and query languages

- Functional Data Model
- Logical Data Model
- O2
- GOOD, GDM
- Strudel and StruQL
- G-BASE, Gram, GraphDB, GRAS
- hypergraphs, hypernode model, hygraphs
- RDF/S and SPARQL

See Survey of Graph Database Models [Angles and Gutierrez, 2008] Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages **G**, **G⁺** and Graphlog

G, G ' and Graphio Lore/Lorel

Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview

Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview

Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query functionality

- graph pattern matching
- path finding
- edge label variables
- negation
- path variables
- aggregation
- approximate matching and ranking
- (disjunction)

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview

Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding Edge and path variables Aggregation Approximate matching and ranking

Example graph

A graph of authors, prizes they have won, and countries where they were born:





Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Example query

Which authors born in South Africa have won both the Nobel Prize in Literature and the Man Booker prize?



X is a variable

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding Edge and path variables Aggregation Approximate matching and ranking

Matching subgraphs

Two matching subgraphs





Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an

Matching subgraphs

Two matching subgraphs





Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and

Matching subgraphs

Two matching subgraphs





Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and

Query answers

Depending on the query language and whether the database is a set of graphs or a single graph, answers mights be the

- set of graphs in which a match is found (e.g. biological applications)
- set of matching subgraphs (NAGA)
- set of variable bindings for each variable (most others)

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation

anking

Forms of query expression

 similar to SQL/OQL (Lorel, RQL):
 select X
 from X.hasWon Y, X.hasWon Z, X.bornIn W
 where Y = Nobel and Z = Booker and W = SouthAfrica

W, X, Y and Z are variables

conjunctive query (similar to NAGA and others):

 $(X) \leftarrow (X, hasWon, Nobel),$ (X, hasWon, Booker),(X, bornIn, SouthAfrica)

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Aggregation Approximate matching and ranking

Query evaluation problem

Given a query expression Q and a graph (database) G, is Q(G) non-empty?

- Combined complexity: both Q and G are part of the input
- Query complexity: input is Q while G is fixed
- Data complexity: input is G while Q is fixed

Often consider data complexity since graphs are assumed to be large while query expressions are assumed to be short

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and

Complexity of query evaluation

For graph pattern matching, the complexity is the same as

- relational conjunctive queries
- subgraph isomorphism

namely

- NP-complete in terms of query and combined complexity
- PTIME in terms of data complexity

Query and combined complexity are in PTIME if the variables in the query satisfy an *acyclicity* condition [Yannakakis, 1981]

But can still be exponential if output all variable bindings

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

More flexible matching



South African if a citizen or born or lives in a place located there

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and Juery languages

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Regular expressions

Regular expression over alphabet Σ of edge labels:

- ϵ (empty string) is a regular expression
- any label in Σ is a regular expression
- if r_1 and r_2 are regular expressions, then so are
 - $(r_1|r_2)$ (alternation)
 - $(r_1 \cdot r_2)$ (concatenation)
- ▶ if r is regular expression, then so is r* (closure)
- may also use a⁻ to mean traversal of edge labelled a in the reverse direction
- r^+ is shorthand for $(r \cdot r^*)$
- *r*? is shorthand for $(r|\epsilon)$
- Σ is shorthand for $(a_1 | \cdots | a_n)$ if $\Sigma = \{a_1, \ldots, a_n\}$

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Regular languages

Language L(r) (set of sequences of labels) denoted by r is given by:

- ϵ denotes $\{\epsilon\}$
- $a \in \Sigma$ denotes $\{a\}$
- $(r_1|r_2)$ denotes $L(r_1) \cup L(r_2)$
- $(r_1 \cdot r_2)$ denotes $L(r_1) \cdot L(r_2)$
- r* denotes L(r)*

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

araph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Paths satisfying regular expressions

Given a graph $G = (N, E, V, \Sigma, \phi, \lambda)$

- ▶ a path p is a sequence of edges $(e_1, e_2, ..., e_n)$ such that, for each $1 \le i \le n$, if $\phi(e_i) = (x, y)$, then $\phi(e_{i+1}) = (y, z)$ for some $x, y, z \in N$
- the path label of p is given by λ(e₁) · λ(e₂) · · · λ(e_n) and is denoted λ(p)
- ▶ path *p* satisfies regular expression *r* if $\lambda(p) \in L(r)$

Regular path query: given *r* and *G*, find all pairs of nodes (x, y) in *G* such there is a path from *x* to *y* which satisfies *r*

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation Approximate matching and ranking



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking



Query Languages for Graph Databases

Complexity of regular path query evaluation

REGULAR PATH PROBLEM

Given graph *G*, pair of nodes x and y and regular expression r, is there a path from x to y satisfying r?

- algorithm:
 - construct a nondeterministic finite automaton (NFA) *M* accepting *L*(*r*)
 - assume *M* has initial state s₀ and final state s_f
 - consider G as an NFA with initial state x and final state y
 - form the "intersection" I of M and G
 - check if there is a path from (s_0, x) to (s_f, y)
- Each step can be done in PTIME, so REGULAR PATH PROBLEM has PTIME combined complexity

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

araph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

NFA *M* for $r = \text{citizenOf} \mid ((\text{bornIn} \mid \text{livesIn}) \cdot \text{locatedIn}^*)$



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Graph G:



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Intersection of G and M



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Intersection of G and M



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Intersection of G and M



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking
Intersection of G and M



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Intersection of G and M



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Intersection of G and M



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Alternatively can translate

citizenOf | ((bornIn | livesIn) · locatedIn*)

to Datalog (as done by Graphlog, e.g.)

- $assoc(X, Y) \leftarrow bornln(X, Y)$
- $assoc(X, Y) \leftarrow livesln(X, Y)$
- $partOf(X, Y) \leftarrow locatedIn(X, Y)$
- $partOf(X, Y) \leftarrow locatedIn(X, Z), partOf(Z, Y)$
- answer(X, Y) \leftarrow citizenOf(X, Y)
- answer(X, Y) \leftarrow assoc(X, Y)
- answer(X, Y) \leftarrow assoc(X, Z), partOf(Z, Y)

Query Languages for Graph Databases

Peter T Wood

G, G⁺ and Graphlog

Path finding

Regular simple path queries

- path p is simple if no node is repeated on p
- REGULAR SIMPLE PATH PROBLEM Given graph G, pair of nodes x and y and regular expression r, is there a simple path from x to y satisfying r?

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Regular simple path queries

- path p is simple if no node is repeated on p
- REGULAR SIMPLE PATH PROBLEM Given graph G, pair of nodes x and y and regular expression r, is there a simple path from x to y satisfying r?
- REGULAR SIMPLE PATH PROBLEM is NP-complete, even for fixed expressions [Mendelzon and Wood, 1989]

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Regular simple path queries

- path p is simple if no node is repeated on p
- REGULAR SIMPLE PATH PROBLEM Given graph G, pair of nodes x and y and regular expression r, is there a simple path from x to y satisfying r?
- REGULAR SIMPLE PATH PROBLEM is NP-complete, even for fixed expressions [Mendelzon and Wood, 1989]
- ► there can be a path from x to y satisfying r but no simple path satisfying r, e.g., r = (c · d)*



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching

Path finding

Edge and path variables Aggregation Approximate matching and ranking

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Edge label variables

what relationship(s) exist between Coetzee and SouthAfrica?

 $X \leftarrow (Coetzee, X, SouthAfrica)$

a "schema-level" query

- answers might be: {bornIn, livesIn, citizenOf}
- find people X and things Y such that X is related Y in the same way as Coetzee is related to Y

 $(X, Y) \leftarrow (Coetzee, Z, Y), (Y, Z^-, X)$

superscript - indicates traversal in reverse direction

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Edge label variables

program analysis example: database transaction

graph:

- nodes represent points in a program
- special nodes start and end
- edges represent operations, e.g., *lock(b)* and *unlock(b)* of some data item b
- is it the case that a transaction tries to lock the same item more than once (not two-phase)?

 $\leftarrow (start, (\Sigma^* \cdot lock(X) \cdot \Sigma^* \cdot lock(X) \cdot \Sigma^*), end)$

- Σ* matches any sequence of edge labels
- sometimes called parameterised regular expressions

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Negation

- program analysis: *def* and *use* of program variables
- to find program points that immediately follow a use of an uninitialized variable

 $Y \leftarrow (start, (\neg def(X))^* \cdot use(X), Y)$

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Negation

- program analysis: *def* and *use* of program variables
- to find program points that immediately follow a use of an uninitialized variable

 $Y \leftarrow (start, (\neg def(X))^* \cdot use(X), Y)$

 to find only the first use of each uninitialized variable along each path

 $\begin{array}{rcl} Y, Z & \leftarrow & (start, ((\neg(def(X) \mid use(X)))^*), Y), \\ & & (Y, use(X), Z) \end{array}$

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Path variables

- may want to know path(s) connecting two nodes:
 - linked data on the web (DBPedia, Freebase)
 - link analysis in criminal networks
 - data provenance
- given regular expression r and variable X, use (r)%X to bind path matching r to X
- edge label variable X is a special case where first occurrence is equivalent to (Σ)%X
- paths connecting Coetzee and Gordimer given by

 $X \leftarrow (Coetzee, ((\Sigma | \Sigma^{-})^{*}) \% X, Gordimer)$

answers: bornIn · bornIn - and hasWon · hasWon-

Lorel uses @, not %; NAGA uses connect keyword

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Path variables

find entities X and Y such that Coetzee is connected to Y in the same way as X is connected to Y

 $(X, Y) \leftarrow (\textit{Coetzee}, (\Sigma^*)\%Z, Y), (X, Z, Y)$

- similar to regular expressions with backreferencing, e.g., in *egrep* (Unix) and in Perl
- membership problem is NP-complete [Aho, 1980]; data complexity is PTIME
- in general, can denote non-context-free languages,
 e.g., {*ww* | *w* ∈ Σ*} as above
- can also do local binding: $((\Sigma \% X) \cdot X)^*$

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding

Edge and path variables

Aggregation Approximate matching and ranking

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

- Overview Graph pattern matching Path finding Edge and path variables Aggregation
- Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation

Approximate matching and ranking

Motivation

To be able to answer traditional graph queries like

- degree of a node
- distance between pairs of nodes
- eccentricity of a node
- diameter, radius and centre of a graph

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation

Approximate matching and ranking

Motivation

To be able to answer traditional graph queries like

- degree of a node
- distance between pairs of nodes
- eccentricity of a node
- diameter, radius and centre of a graph
- and applications like
 - shortest path
 - most reliable path
 - critical path
 - bill of materials

▶ ...

need operators such as count, min, max, sum

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation

Approximate matching and ranking

Aggregation in Graphlog

- aggregate terms are allowed in label of distinguished edge or distinguished node
- following query computes, for each directory D, the total file space used by all contained files and sub-directories, other than those residing on disk1



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

araph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation Approximate matching and

From Graphlog to Datalog



Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation

Approximate matching and ranking

From Graphlog to Datalog

containsPlus(X, Y) containsPlus(X, Y) diskUtil(D, sum(S))

 $\leftarrow contains(X, Y) \\ \leftarrow contains(X, Z), containsPlus(Z, Y) \\ \leftarrow containsPlus(D, F), size(F, S), \\ \neg residesOn(F, disk1)$

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation

Approximate matching and ranking

Graphlog example

- we might also want to summarise values along a path and then aggregate
- following query computes the length of the shortest path between each pair of nodes

shortestPath(min(sum(D))



- D is called a collecting variable
- sum is used to summarise distances along a path
- min is used to aggregate the summarised distances
- query evaluation is in PTIME if summarisation and aggregation operators form a closed semiring [Consens and Mendelzon, 1990]

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables

Aggregation

Approximate matching and ranking

Outline

Background

Graphs Personal interest Uses of graphs

Graph models and query languages

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation

Approximate matching and ranking

Motivation

- users may not be familiar with graph structure/constraints
- may formulate queries which return no answers or too few answers, e.g.
 - expression course · student when correct path is student · course
 - expression restaurant · zipcode when address is required between them
- can perform approximate matching of paths
- rank results in terms of "closeness" to original query

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation

Approximate matching and ranking

Approximate matching

- can modify the user's original query (regular expression r)
- one way is to apply edit operations to L(r)
 - insertions
 - deletions
 - substitutions
 - transpositions
 - invertions
- each operation may have a different cost
- somewhat related to user preferences
 - prepared to substitute train by bus but at cost 2

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation

Approximate matching and ranking

Approximate matching algorithm

- for conjunctive regular path queries
- can use algorithms from approximate string matching
- incrementally build an approximate NFA
- perform incremental joins for conjuncts
- PTIME combined complexity if conjuncts are acyclic and fixed number of head variables
- in general, can transform NFA using a regular transducer
- see [Hurtado, Poulovassilis and Wood, 2009]

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

araph models and

G, *G*⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation

Approximate matching and ranking

Summary

- motivated that graph-based data is widely used and available
- a brief high-level overview of some query languages for graph databases
- focussed on query language functionality
- some discussion of query evaluation algorithms
- some complexity results mentioned

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

araph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

Issues not covered

Many issues not covered

- other languages
- more query evaluation strategies, e.g., using indexes
- graphs with schemas
- query optimisation, e.g., containment

• . . .

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking

- S. Abiteboul, D. Quass, J. McHugh, J. Widom, and J. L. Wiener.
 The LOREL query language for semistructured data. *Int. J. on Digital Libraries*, 1(1):68–88, April 1997.
- A. V. Aho.

Pattern matching in strings.

In R. V. Book, editor, *Formal Language Theory: Perspectives and Open Problems*, pages 325–347. Academic Press, 1980.

- R. Angles and C. Gutierrez. Survey of graph database models. ACM Comput. Surv., 40(1):1–39, 2008.
- M. P. Consens and A. O. Mendelzon. Expressing structural hypertext queries in GraphLog. In *Proc. Second ACM Conf. on Hypertext*, pages 269–292, 1989.

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

- M. P. Consens and A. O. Mendelzon.
 Low complexity aggregation in GraphLog and Datalog.
 In *Proc. 3rd Int. Conf. on Database Theory*, pages 379–394, 1990.
- I. F. Cruz, A. O. Mendelzon, and P. T. Wood.
 A graphical query language supporting recursion.
 In ACM SIGMOD Int. Conf. on Management of Data, pages 323–330, 1987.
- I. F. Cruz, A. O. Mendelzon, and P. T. Wood.
 G⁺: Recursive queries without recursion.
 In *Proc. 2nd Int. Conf. on Expert Database Systems*, pages 355–368, 1988.

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

C. A. Hurtado, A. Poulovassilis, and P. T. Wood. Ranking Approximate Answers to Semantic Web Queries.

In Proc. 6th European Semantic Web Conference, pages 263–277, 2009.

- A. O. Mendelzon and P. T. Wood. Finding regular simple paths in graph databases. In *Proc. 15th Int. Conf. on Very Large Data Bases*, pages 185–193, 1989.
- G. Weikum, G. Kasneci, M. Ramanath, and F. Suchanek.

Database and information-retrieval methods for knowledge discovery.

Commun. ACM, 52(4):56-64, 2009.

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching and ranking

M. Yannakakis.

Algorithms for acyclic database schemes.

In Proc. 7th Int. Conf. on Very Large Data Bases, pages 82–94, 1981.

Query Languages for Graph Databases

Peter T. Wood

Background

Graphs Personal interes Uses of graphs

Graph models and

G, G⁺ and Graphlog Lore/Lorel YAGO/NAGA Other models and languages

Query Functionality

Overview Graph pattern matching Path finding Edge and path variables Aggregation Approximate matching an ranking