

Locality Optimization for traversal-based Queries on Graph Databases

Fabian Klopfer

Databases and Information Systems Group Department of Computer and Information Science University of Konstanz, 30.04.2021

Outline

- Introduction
- Background
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- Locality-optimizing Record Layout 4

- 5 **Evaluation**
- 6 Conclusion

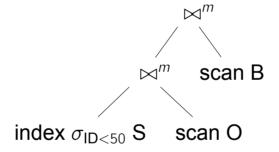
Motivation I

Current state of performance-optimized

- relational databases:
 - Storage order determined by join attribute. → enables localized access.
 - Explicit clustered indices (often B+-Trees). → represents order, speeds up range queries
 - ⇒ Accesses are made as sequential as possible.
- graph databases:
 - Storage order determined by insertion order.
 - Implicit, possibly unclustered index for relationships (doubly-linked list).
 - Lucene-based indexes on properties, labels, ..., unclustered.
 - ⇒ Access is mostly random.

Example I

Show me all boats owned by sailors with an ID less than 50:



Reads are mostly sequential.

⇒ Prefetch & cache hit.

Example II

Nodes are Sailors and Boats, relationships "owns"

$$\sigma_{y.\text{label}} = \text{Boat}$$

$$\uparrow_{x}^{y}$$

$$\mid$$

$$\sigma_{x.ID<50} (\bigcirc x)$$

Scaning and filtering is sequential. Expand is not.

 \Rightarrow Expand causes prefetch & cache misses.

1 Introduction

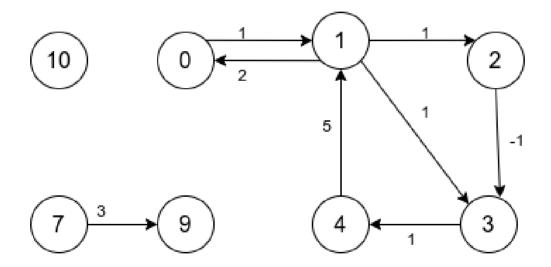
Example III

Especially Expand jumps a lot. Potentially back and forth.

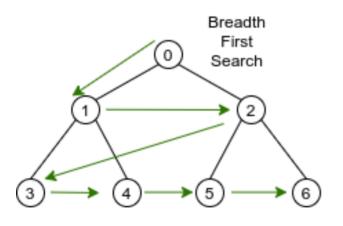
Locality Optimization for traversal-based Queries on Graph Databases

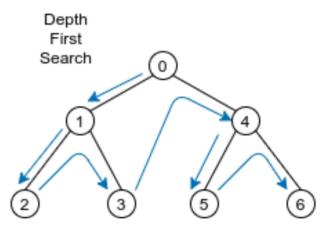
Traversals rely primarily on expand.

Graphs



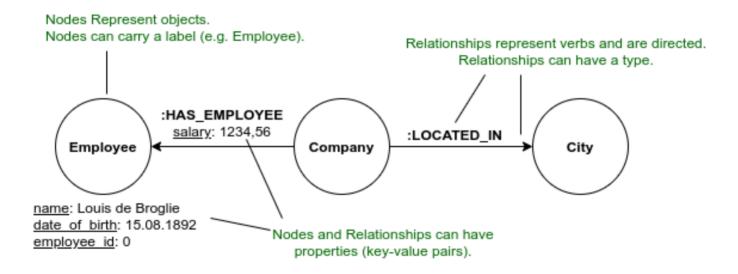
Traversals





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Property Graph Model



Data Structures I

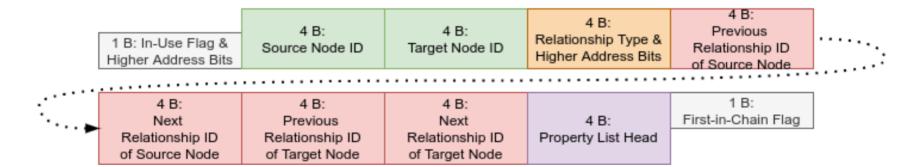
Two essential record structures:

- Node records
- 2. Relationship records

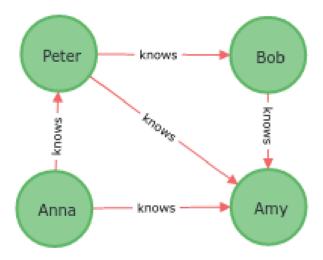
Data Structures II

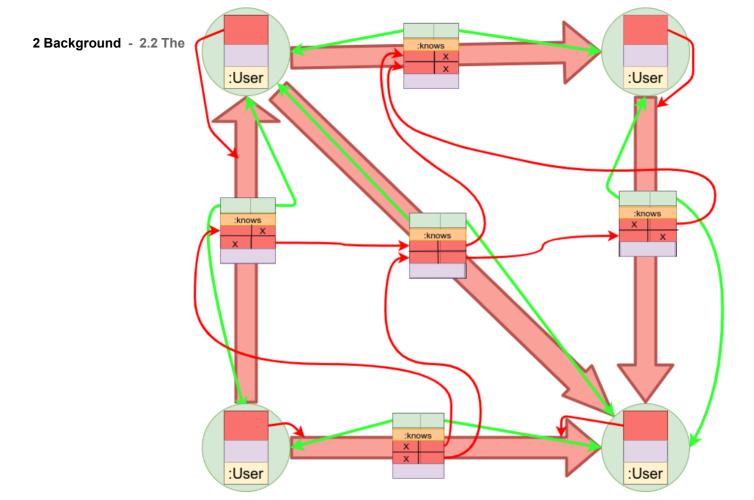


Data Structures III



Data Structures IV





Problem Definition I

Given a graph G, logical block size b, page size p.

Desired is

- 1. A partition of G into blocks of vertex records V_i and E_i relationship records,
- 2. permutations π_v , π_e of the blocks of vertex and edge records V_i , E_i ,
- 3. a reordering of the incidence list pointers

such that spatial locality is as high as possible for traversal-based queries.

Problem Definition II

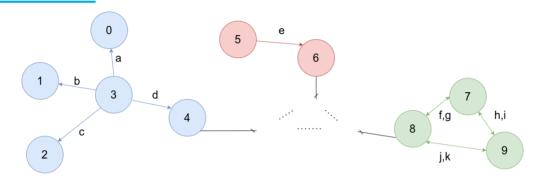
Temporal locality based on blocks.

$$P(X_{t+\Delta} = B|X_t = B)$$

Spatial locality in the same sense:

$$P(X_{t+\Delta} = B \pm \varepsilon | X_t = B)$$

Problem Definition III



node.db	0, 5, 7	1, 4, 9	2, 6, 8	3		
edge.db	a, f	b, g	c, h	d, i	e, j	k

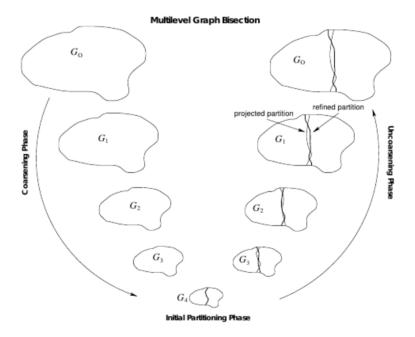
node.db	7,8,9	0, 1, 3	2, 4, 5	6		
edge.db	f, h	g, k	i, j	a, b	c, d	е

3 Problem Definition

Problem Definition IV

TODO insert incidence list chaos and ordered here

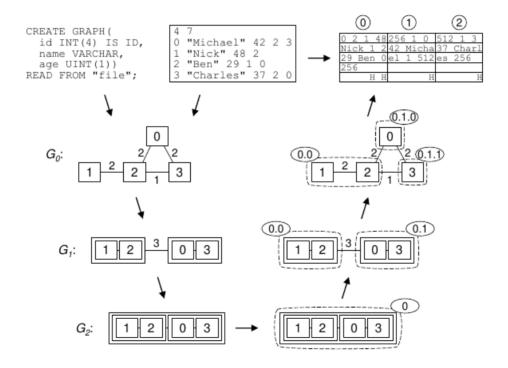
G-Store I



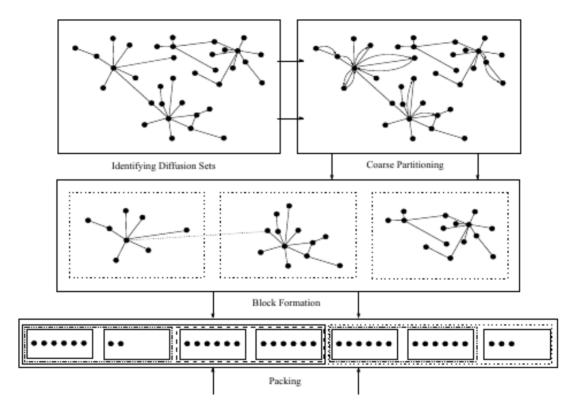
G-Store II

- 1. Coarsening: Heavy-Edge Matching
- 2. Turn-around
- 3. Uncoarsening
 - 3.1 Project
 - 3.2 Reorder
 - 3.3 Refine

G-Store III



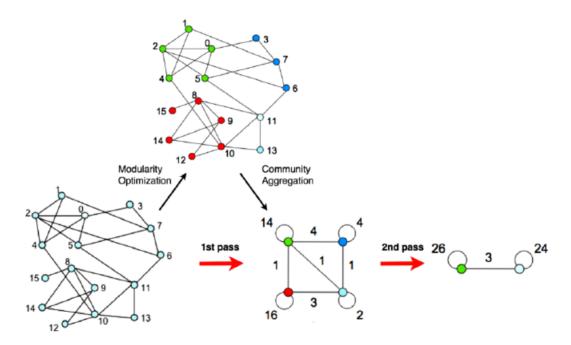
ICBL I



ICBL II

- Feature extraction: Do *t* random walks of length /.
- C Coarse clustering: Adapted K-Means.
- B Block Formation: Agglomerative hierarchical clustering.
- L Layout Blocks: Sort blocks and subgraphs

Louvain Method I



Louvain Method II

- 1. Initialize all nodes in singleton community.
- Merge community into a neighboring community where modularity gain is maximal, until modularity gain is below threshold.
- 3. Construct new graph from aggregated communities and go to 1.

$$Q = \frac{1}{2m} \sum_{u,v \in V} \left(w_{(u,v)} - \frac{w_u w_v}{2m} \right) \cdot \delta(c_u, c_v)$$

Incidence List Rearrangement

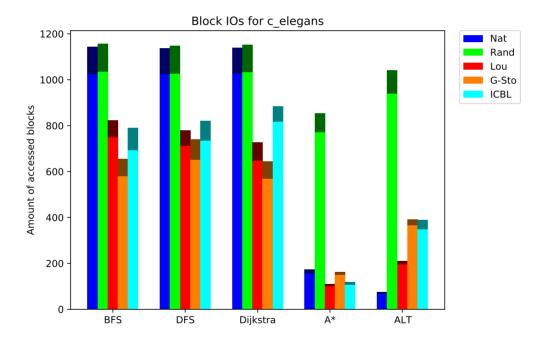
- Traverse incidence list and store IDs.
- Sort IDs.
- Assign first relationship pointer to lowest ID.
- Assign next pointer of new first relationship to second ID.
- 5. Assign next pointer of relationship i to i + 1. ID and prev pointer to i 1. ID.
- 6. Assign next pointer of last relation to first and prev pointer of first to last.

5 Evaluation

Setup

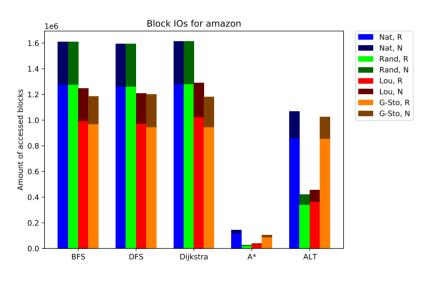
- Queries: BFS, DFS, Dijkstra, A*, ALT.
- Datasets: [131, 1'134'890] nodes, [764, 2'987'624] edges, average degree [2.6, 25.5]
- Domains include biological neural net, E-Mails, Co-authors, Frequent item sets, Comments.

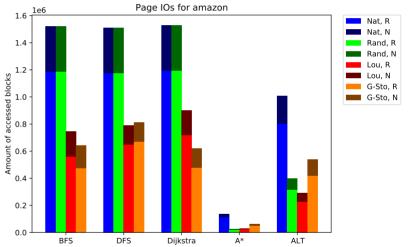
Results I



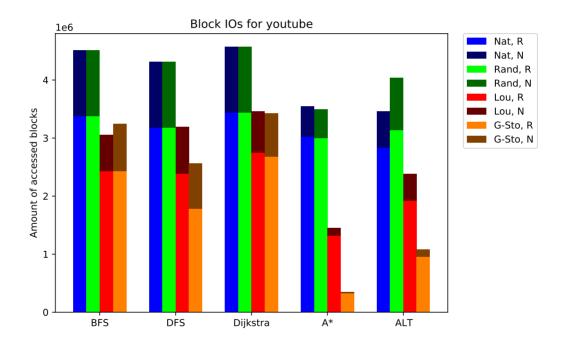
5 Evaluation

Results II

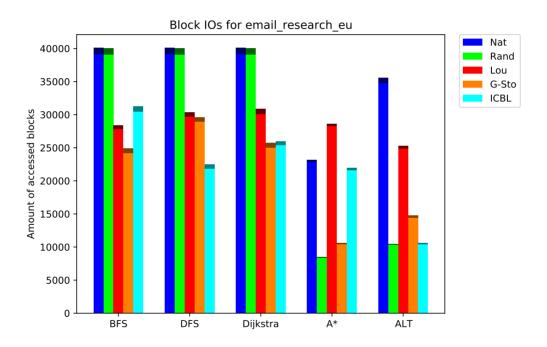




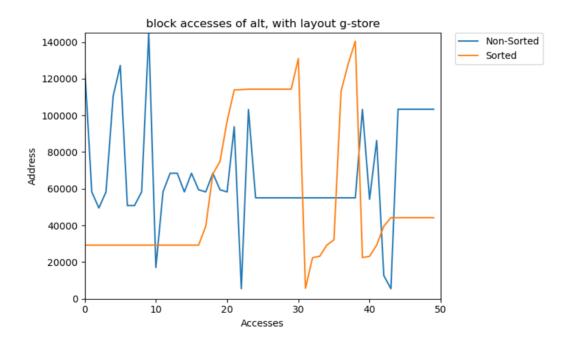
Results III



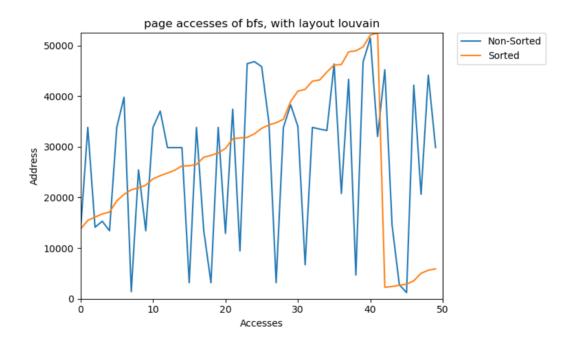
Results IV



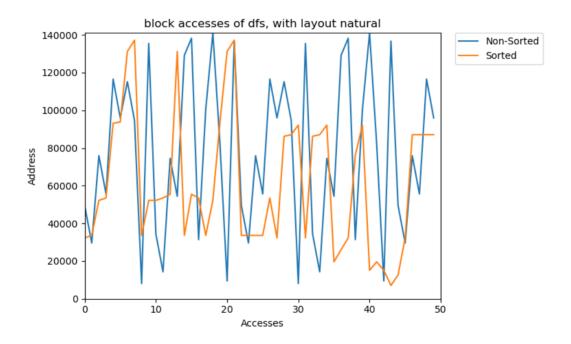
Results V



Results VI



Results VII



Summary I

- Static rearrangement methods decrease number of block accesses.
 - ⇒ increase locality
- Sorting the incidence lists leads to more sequential access sequences.
- Ordering the blocks is crucial for spatial locality.

Future Work I

- Leiden instead of Louvain
- RCM-based rearrangement
- Dynamic Rearrangement Query-based
- Disk-based implementation