Universität Konstanz

Locality Optimization for traversal-based Queries on Graph Databases

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University of Konstanz, 30.04.2021

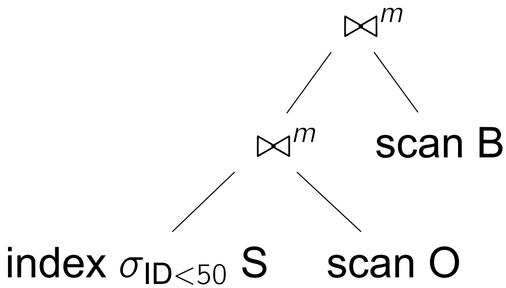
Motivation

Current state of performance-optimized

- relational databases: accesses are made as sequential as possible.
- graph databases: access is often random.

Motivation – Example I

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



Reads are mostly sequential.

$\Rightarrow$ Prefetch & cache hit.

Motivation – Example II

Nodes are Sailors and Boats, relationships “owns”

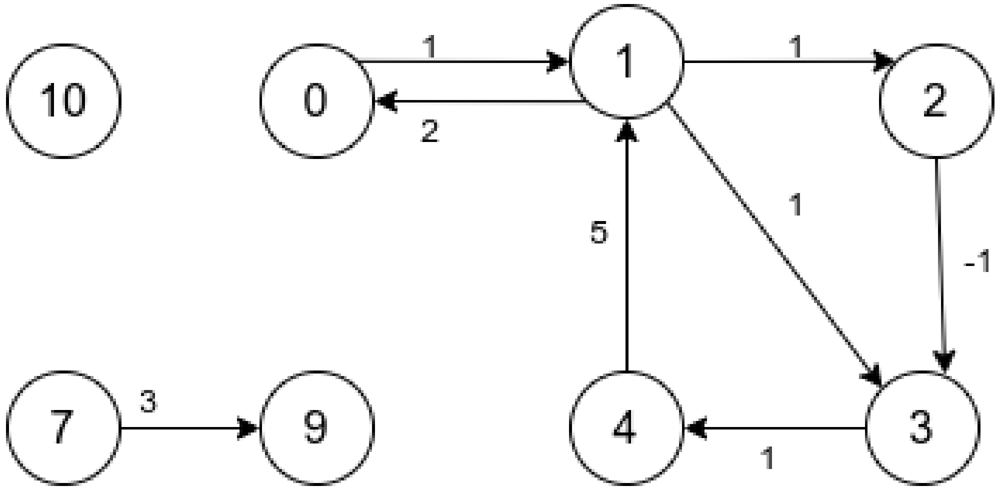
$$\begin{array}{c} \sigma_{y.\text{label = 'Boat'}} \\ | \\ \uparrow^{y}_{x} \\ | \\ \sigma_{x.ID<50 \wedge x.\text{label = 'Sailor'}}\ (\bigcirc\ x) \end{array}$$

Scaning and filtering is sequential. `Expand` is not.

$\Rightarrow$ `Expand` causes prefetch & cache misses.

Traversals rely primarily on expand.

Graphs



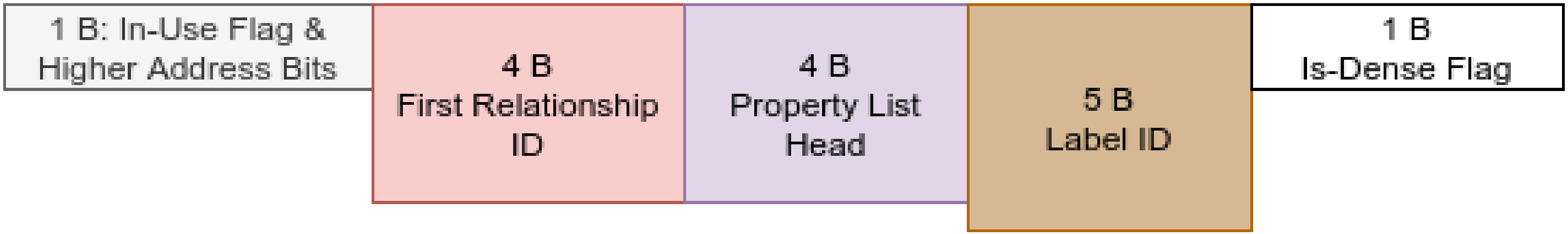
Data Structures I

Two essential record structures:

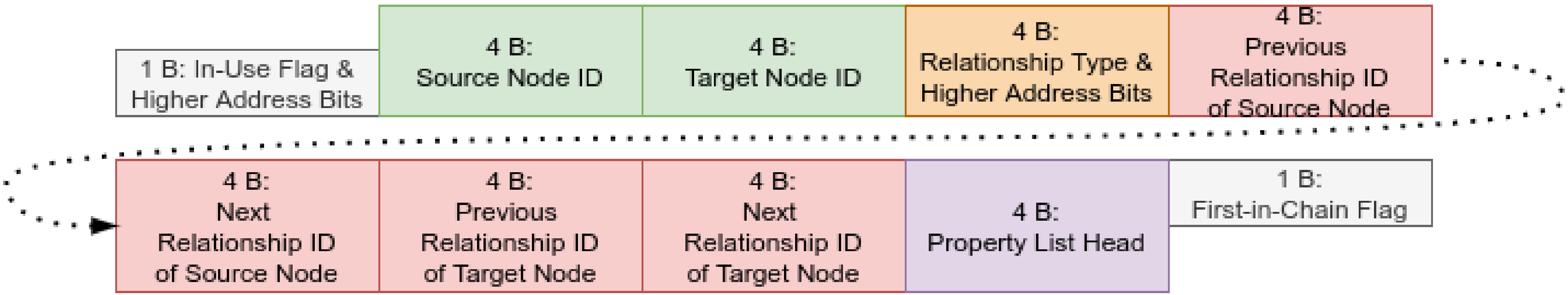
1. Node records
2. Relationship records

Inspired by Neo4J

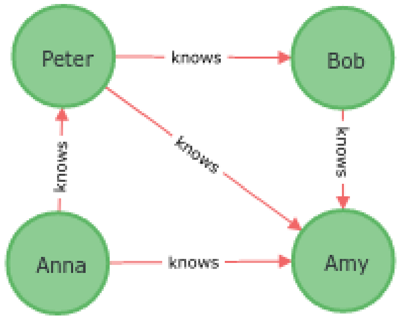
Data Structures II

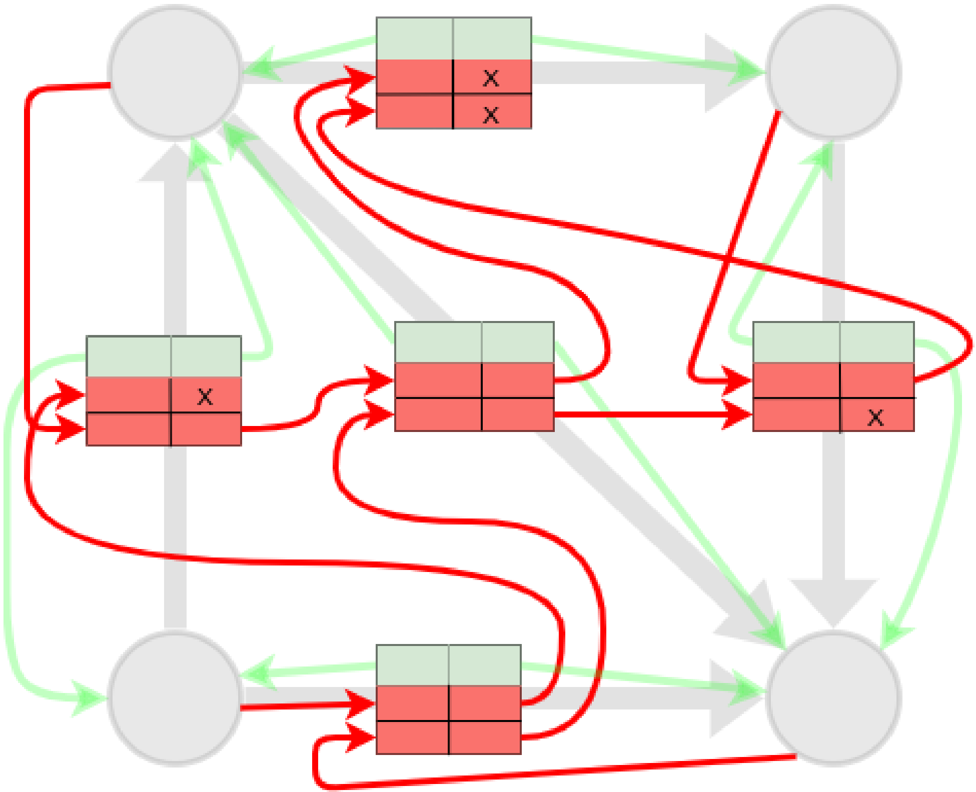


Data Structures III

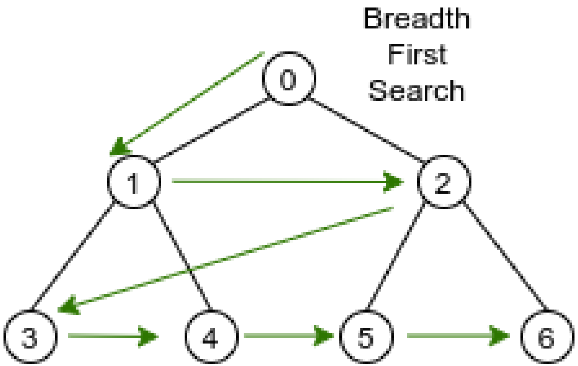


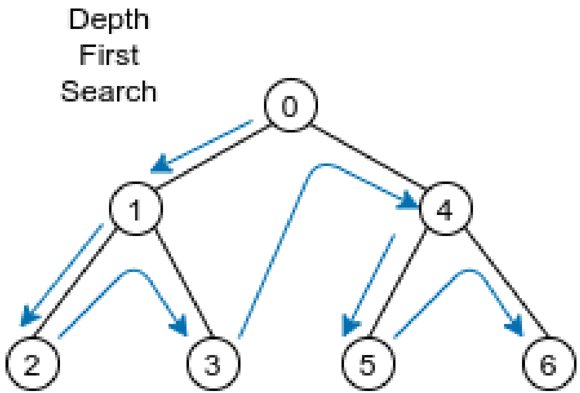
Data Structures – Example

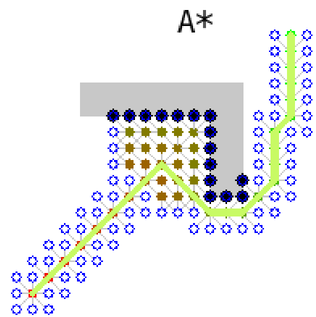



X
X
X
X

Traversals







Problem Definition I

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

Desired is

1. A partition of G into blocks of size b,
2. permutations π_v, π_e of the blocks,

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

Problem Definition II

Temporal locality based on blocks.

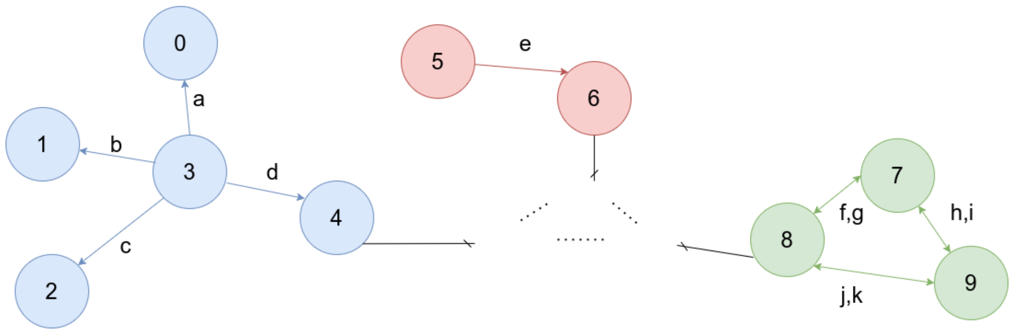
$$P(X_{t+\Delta} = B | X_t = B)$$

Spatial locality in the same sense:

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

with $\varepsilon = \lceil \frac{p}{b} \rceil$.

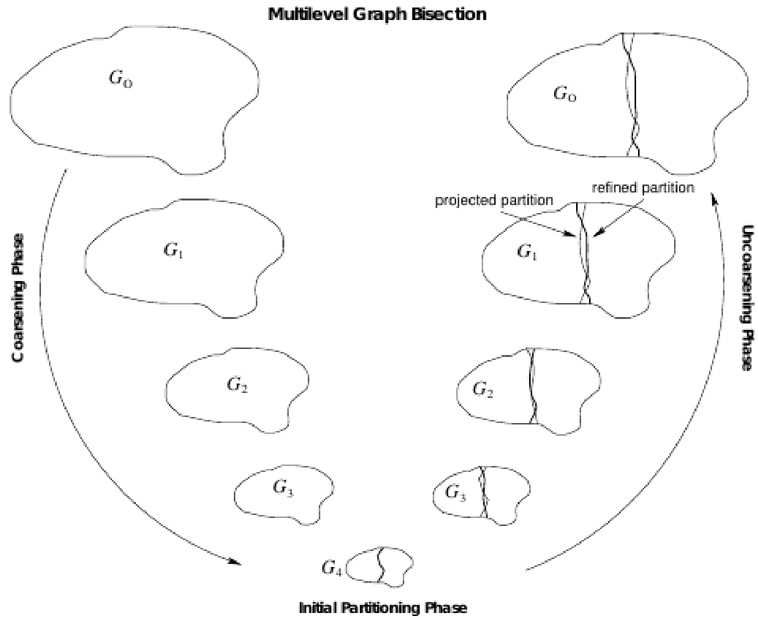
Problem Definition III



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	e

G-Store I

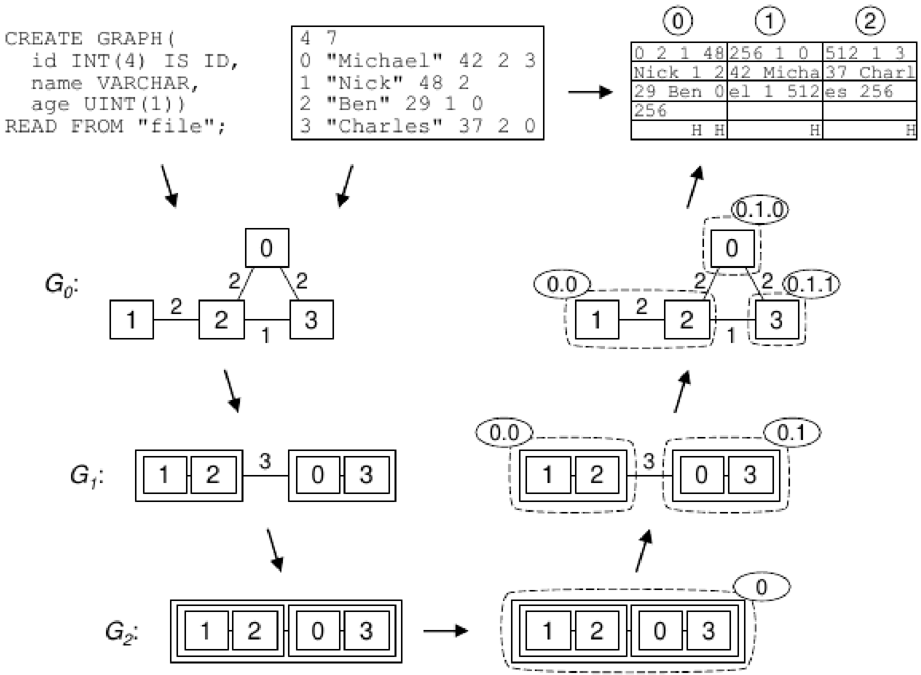

Multilevel Graph Bisection
G_0
G_1
G_2
G_3
G_4
Coarsening Phase
Uncoarsening Phase
projected partition
refined partition
Initial Partitioning Phase

G-Store II

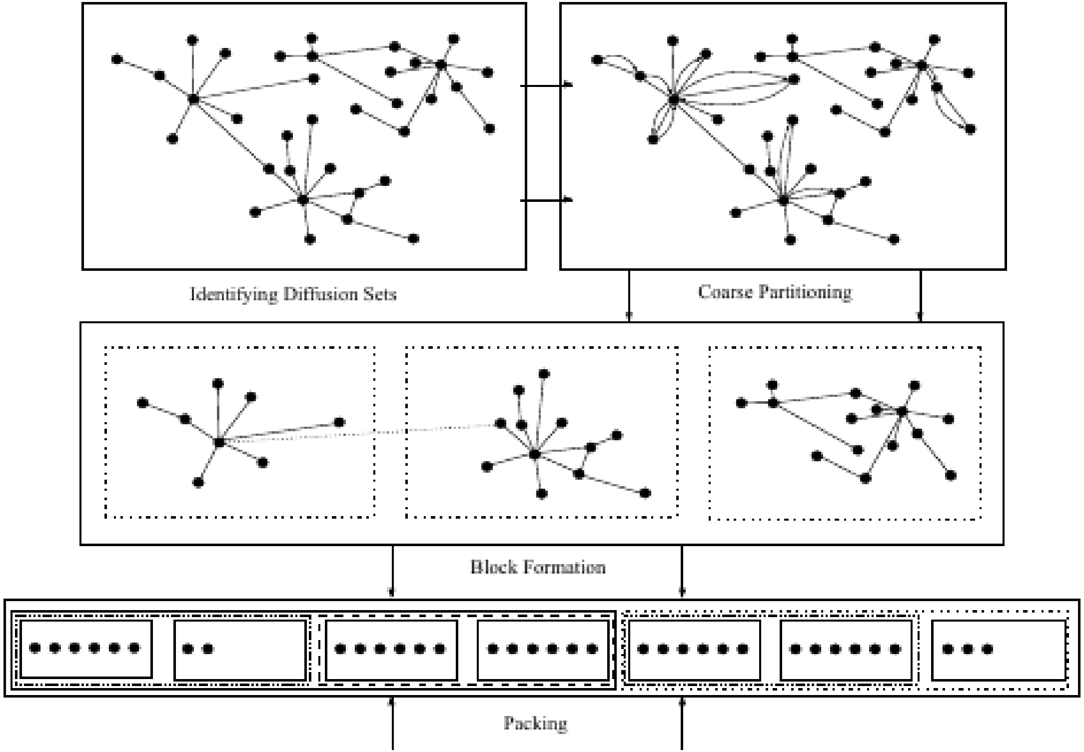
1. Coarsening: Heavy-Edge Matching
2. Turn-around
3. Uncoarsening
   3.1 Project
   3.2 Reorder
   3.3 Refine

$$\min \sum_{(u,v)\in E} |\phi(u) - \phi(v)|$$

G-Store III


CREATE GRAPH(
id INT(4) IS ID,
name VARCHAR,
age UINT(1))
READ FROM "file";
4 7
0 "Michael" 42 2 3
1 "Nick" 48 2
2 "Ben" 29 1 0
3 "Charles" 37 2 0
0
1
2
0 2 1 48
256 1 0
512 1 3
Nick 1 2
42 Micha
37 Charl
29 Ben 0
el 1 512
es 256
256
H H
H
H
G0:
G1:
G2:
0.0
0.1
0.1.0
0.1.1

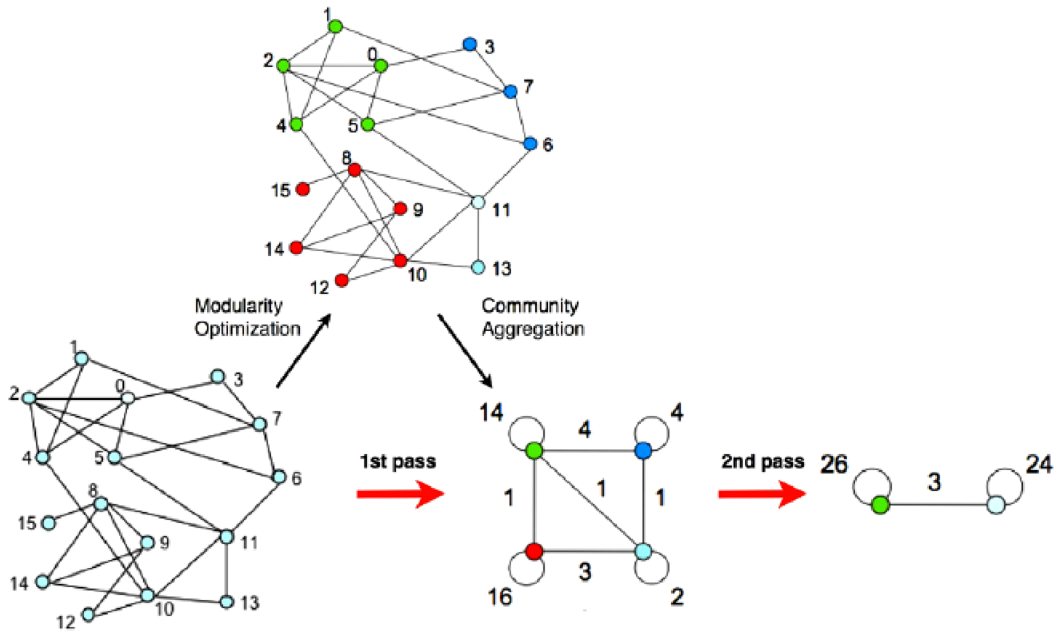
ICBL I


Identifying Diffusion Sets
Coarse Partitioning
Block Formation
Packing

ICBL II

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

Louvain Method I

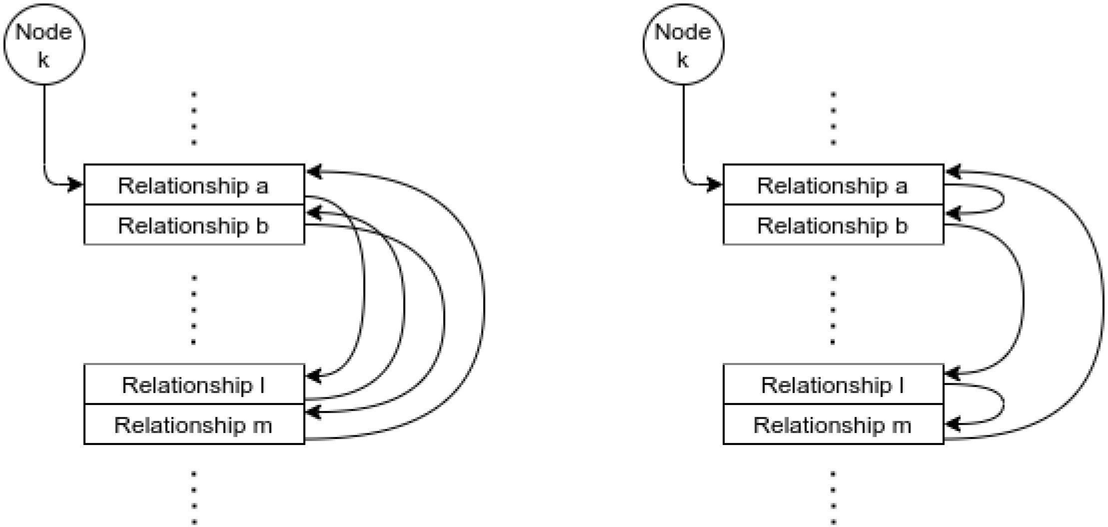


Louvain Method II

1. Initialize all nodes in singleton community.
2. 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.

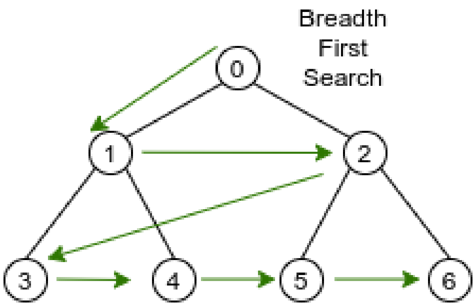
$$\frac{1}{2m}\sum_{u,v\in V}\left(w_{(u,v)}-\frac{w_u w_v}{2m}\right)\cdot\delta(c_u,c_v)$$

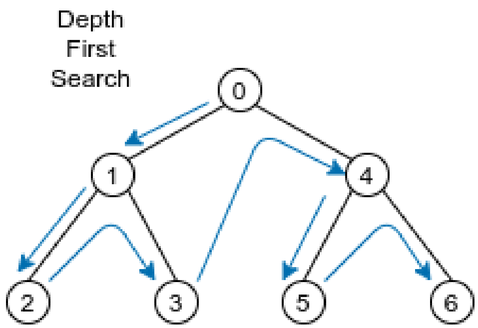
Incidence List Rearrangement

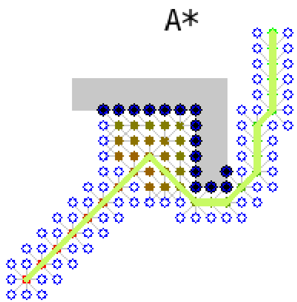


Setup I

- Queries: BFS, DFS, Dijkstra, A*, ALT.





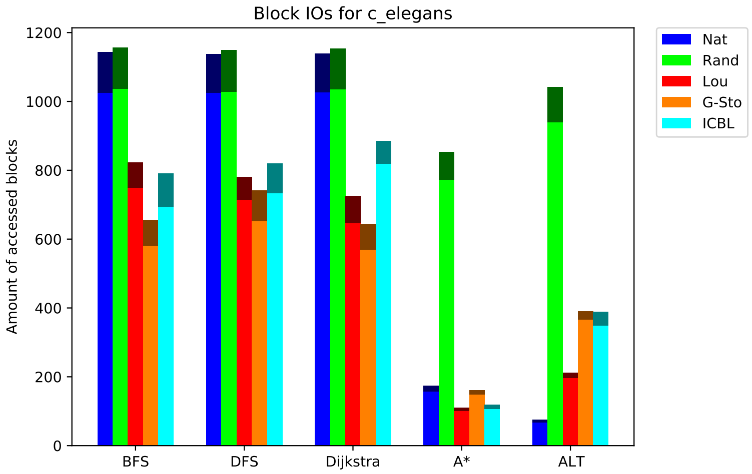


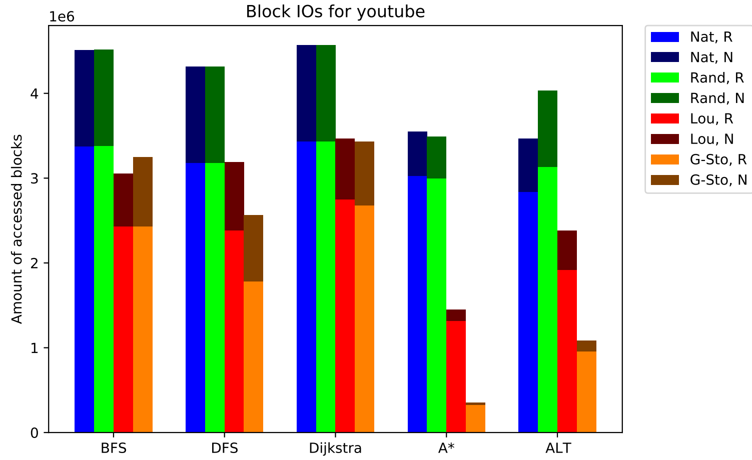
Setup II

- Datasets: $[131, 1'134'890]$ nodes, $[764, 2'987'624]$ edges, average degree $[2.6, 25.5]$

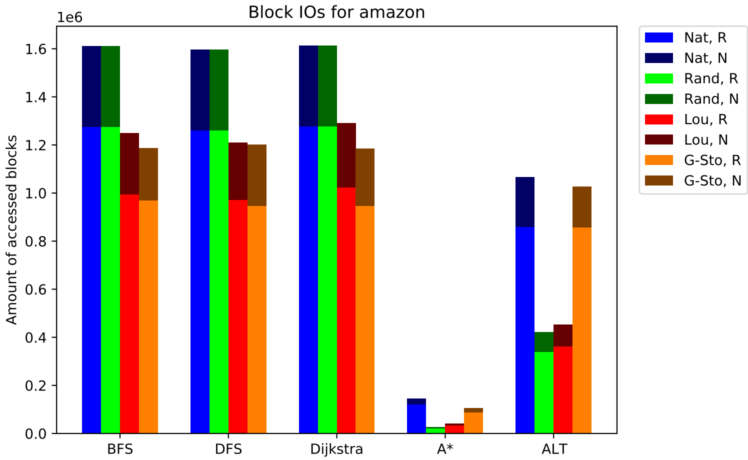
- Domains include biological neural net, e-mails, co-authors, frequent item sets, video channel subscriptions.

Results I

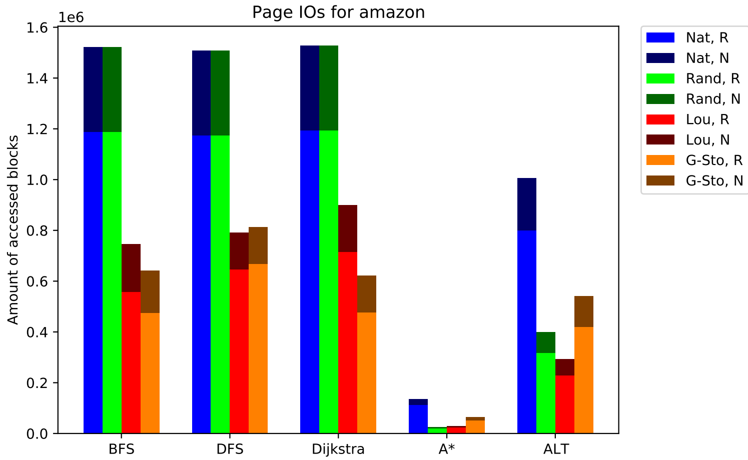

Block IOs for c_elegans
Amount of accessed blocks
0
200
400
600
800
1000
1200
BFS
DFS
Dijkstra
A*
ALT
Nat
Rand
Lou
G-Sto
ICBL


Block IOs for youtube
1e6
Amount of accessed blocks
0
1
2
3
4
BFS
DFS
Dijkstra
A*
ALT
Nat, R
Nat, N
Rand, R
Rand, N
Lou, R
Lou, N
G-Sto, R
G-Sto, N

Results II

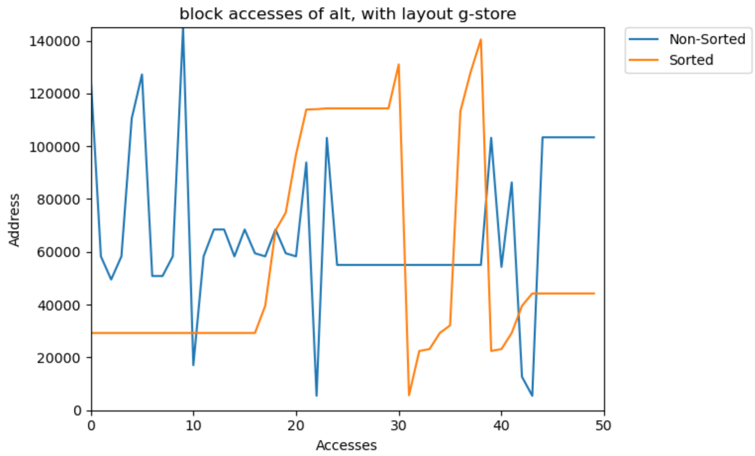


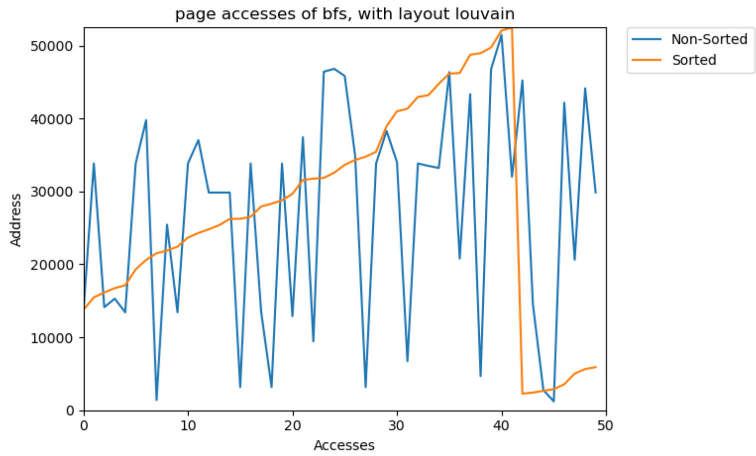
Block IOs for amazon
1e6
Amount of accessed blocks
0.0
0.2
0.4
0.6
0.8
1.0
1.2
1.4
1.6
BFS
DFS
Dijkstra
A*
ALT
Nat, R
Nat, N
Rand, R
Rand, N
Lou, R
Lou, N
G-Sto, R
G-Sto, N



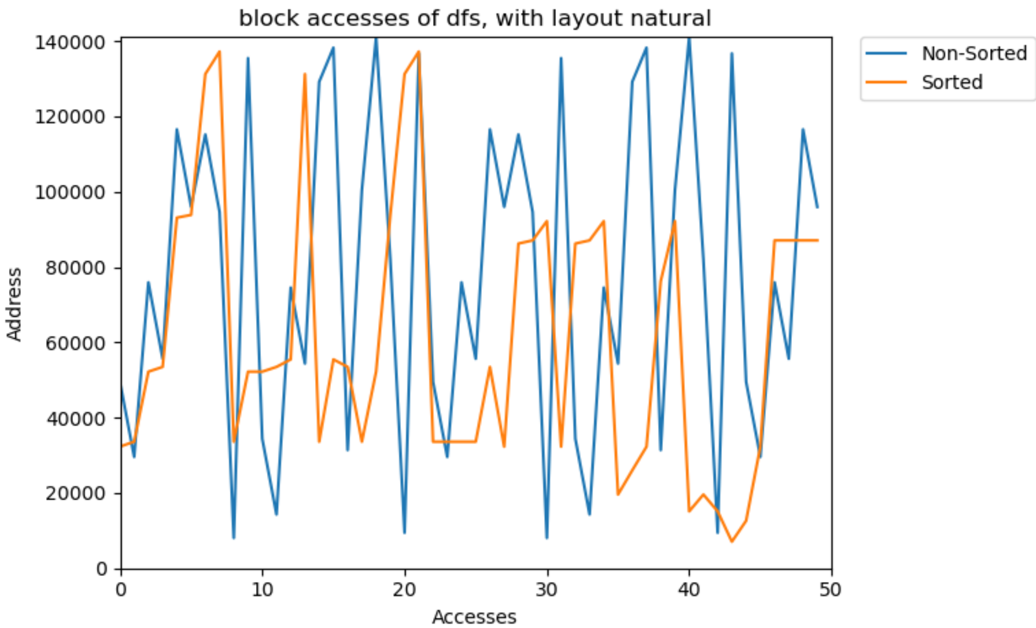
Page IOs for amazon
1e6
Amount of accessed blocks
0.0
0.2
0.4
0.6
0.8
1.0
1.2
1.4
1.6
BFS
DFS
Dijkstra
A*
ALT
Nat, R
Nat, N
Rand, R
Rand, N
Lou, R
Lou, N
G-Sto, R
G-Sto, N

Results III


block accesses of alt, with layout g-store
Non-Sorted
Sorted
Address
Accesses
0
20000
40000
60000
80000
100000
120000
140000
0
10
20
30
40
50


page accesses of bfs, with layout louvain
Non-Sorted
Sorted
Address
Accesses
0
10000
20000
30000
40000
50000
0
10
20
30
40
50

Results IV


block accesses of dfs, with layout natural
Non-Sorted
Sorted
Address
Accesses
0
20000
40000
60000
80000
100000
120000
140000
0
10
20
30
40
50

Summary

- Static rearrangement methods decrease number of block accesses.
  $\Rightarrow$ increase locality

- Ordering the blocks is crucial for spatial locality.

- Sorting the incidence lists leads to more sequential access sequences.

Future Work

- Leiden instead of Louvain
- RCM-based rearrangement
- Dynamic Rearrangement
  - Query-based
  - History-based