CS729 Efficient Subgraph Matching

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Agenda

SPath

SPath query ordering with respect to GraphQL/VF2 Plus

CPI

- Query decomposition
- Combination of decomposition and CPI construction
- CPI construction in Neo4j
- Experimental results
- Lessons learned



- A new graph indexing technique towards resolving the graph query problem efficiently on large networks.
- It maintains for each query vertex a neighborhood signature a structure that stores decomposed shortest path information within the vertex's vicinity.
- SPath, revolutionizes the way of graph query processing from vertex-at-a time to path-at-a-time, which proves to be more cost effective than traditional graph matching methods.



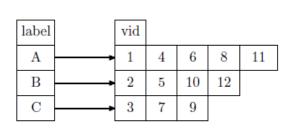
The process is as follows:

- Decompose a query graph into a set of shortest paths, among which a subset of candidate paths with high selectivity is picked by a graph query optimizer.
- The query is further processed by joining candidate shortest paths in order to reconstruct the original query graph.

The principle of SPath is to use shortest paths within the k-neighborhood subgraph of each vertex of the graph to capture the local structural information around the vertex.



The structure used to store SPath:



Histogram			ID-List				
distance	label	count		vid			
1	A	3		1	4	6	
1	В	2		2	5		•
2	A	1		8			
	C	2		7	9		

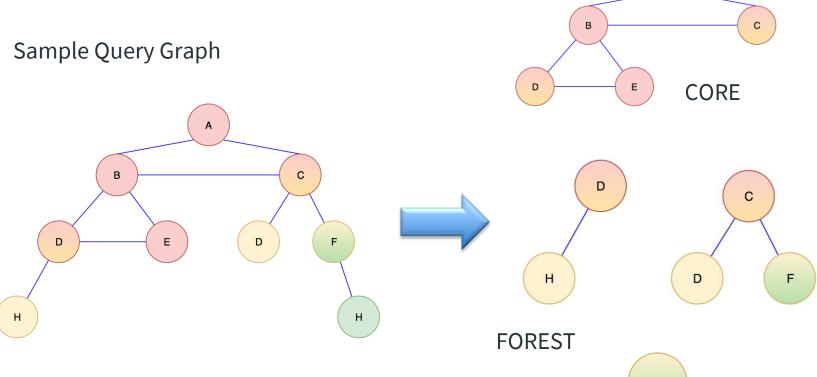
(a) Global Lookup Table (b) Histogram and ID-List

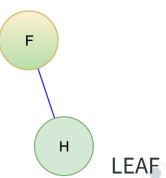
Figure 2: The Global Lookup Table $\mathcal H$ and the Histogram and ID-List of $NS(u_3)$, $u_3 \in V(G)$ $(k_0 = 2)$



 They claim SPath is the first scalable graph indexing mechanism which supports effective path-at-a-time graph query processing on large networks, and thus achieves far better query performance, compared with other traditional vertex-at-a-time graph matching methods.



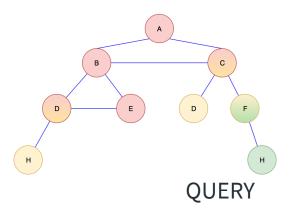


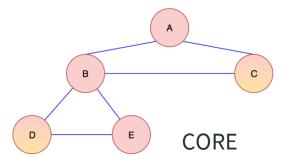


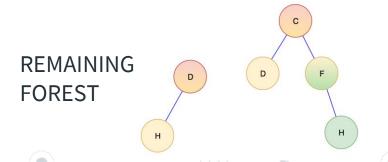
Core-Forest Decomposition

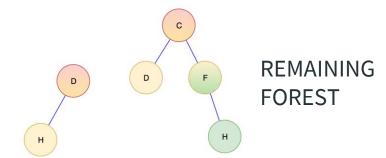
As mentioned in the paper,

- . Iteratively delete all nodes with degree 1
- 2. Remaining nodes represent CORE





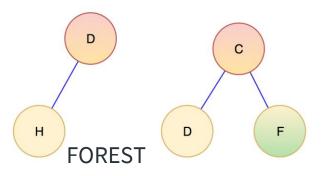


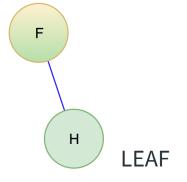


Forest-Leaf Decomposition

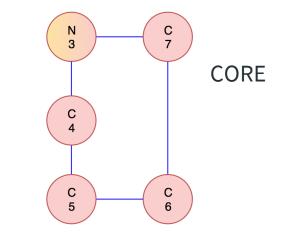
Different approach,

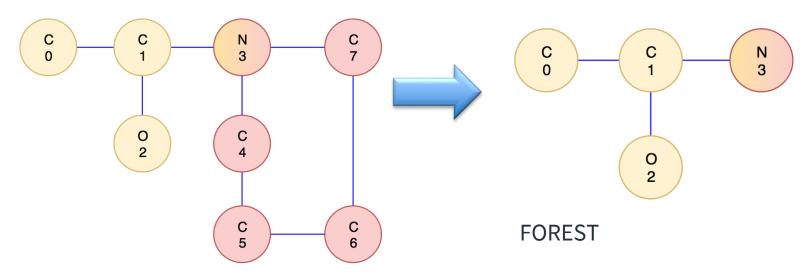
- 1. Leaf nodes have only one neighbor, thus degree of leaf node should be 1
- Also, leaf node or its neighbor(parent) should not be the part of CORE. This, can be confirmed by looking at the degree of these two nodes
- 3. Now, if the neighbor is not a part of the Core, then its degree should be less than or equal to 2





Query graph from proteins database - "ecoli_1RF8.8.sub.grf"

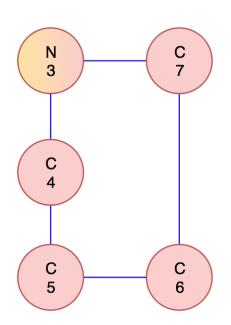


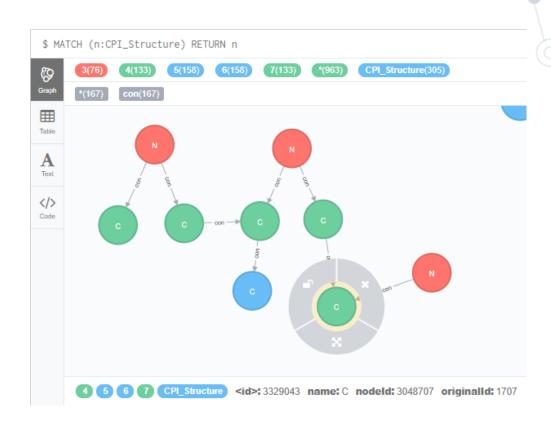


NO LEAF

Combination of Decomposition and CPI Construction

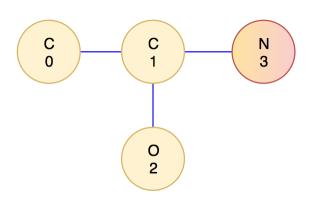
CPI for CORE

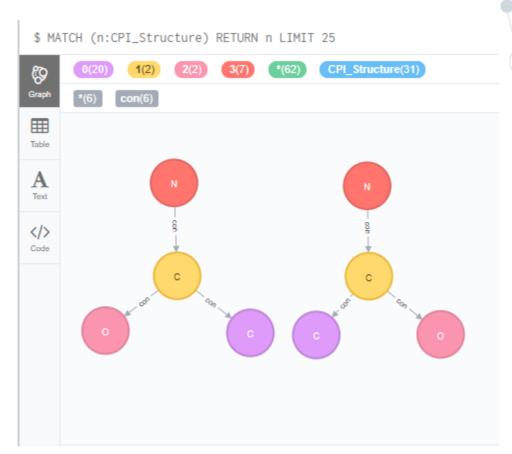




Combination of Decomposition and CPI Construction

CPI for FOREST





CPI construction in Neo4j

Top Down CPI construction:

Directly generate node in neo4j with following properties:

Label: CPI_Structure, 0

<id>: 3290308,

Property: name:N, nodeld: 3047906, originalld: 906

Check as line 26-28, add edge between node if not present

```
/* Lines 24-28: Adjacency List Construction */

for each query vertex u at level lev do

u_p \leftarrow u.p;
for each vertex v_p \in u_p.C do

for each vertex v \in N_G(v_p) with label l_q(u) do

if v \in u.C then N_u^{u_p}(v_p) \leftarrow N_u^{u_p}(v_p) \cup \{v\};

29 return CPI;
```

CPI construction in Neo4j

Bottom-Up Refinement:

Remove/update graph in Neo4J (Remove edge, Remove node)

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for each vertex v \in u.C do

for each child u' of u in the BFS tree of q do

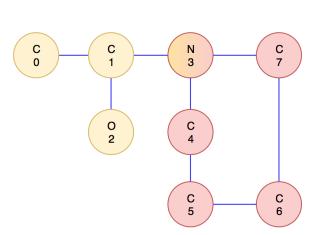
for each vertex v' \in N_{u'}^u(v) do

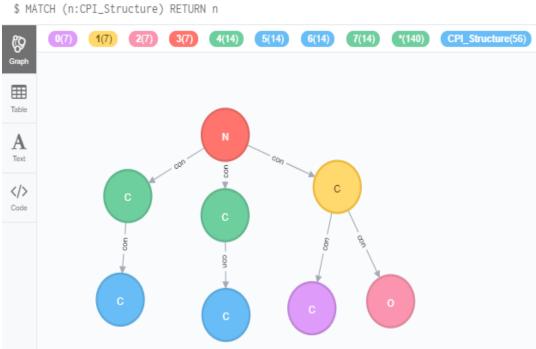
if v' \notin u'.C then N_{u'}^u(v) \leftarrow N_{u'}^u(v) \setminus \{v'\};
```



CPI construction in Neo4j

Query graph from proteins database "ecoli_1RF8.8.sub.grf":





Experimental Results

Target Graph (5):

backbones_140L.grf backbones_1NOD.grf backbones_2X3T.grf human_1B9G human_2KSA.grf

50 query per protein size

- Total 300 query
- Get the average time (Min)

Query Graph (60):

Protein 8 Protein 16 Protein 32

X

Protein 64

Protein 128

Protein 256

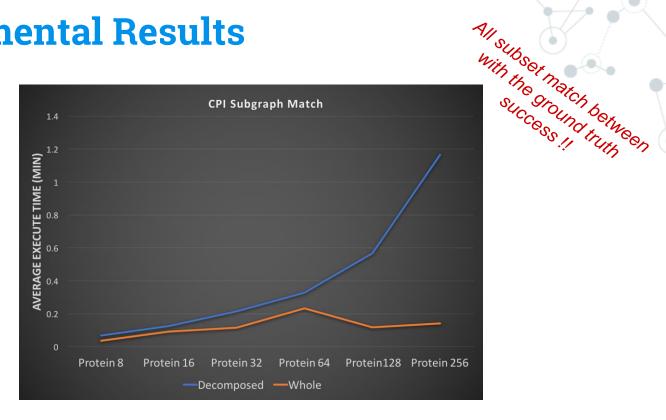


Experimental Results

Decomposed	Protein 8	Protein 16	Protein 32	Protein 64	Protein 128	Protein 256
Max (Min)	0.2547	0.3771	0.6458	0.7971	1.9941	3.7188
Min (Min)	0.1130	0.0198	0.0240	0.0293	0.0301	0.0768

Whole	Protein 8	Protein 16	Protein 32	Protein 64	Protein 128	Protein 256
Max (Min)	0.1384	0.2758	0.39365	0.6909	1.0577	1.7942
Min (Min)	0.0143	0.0170	0.0126	0.0167	0.0107	0.0227

Experimental Results



Averager time (Min)	Protein 8	Protein 16	Protein 32	Protein 64	Protein 128	Protein 256
Decomposed	0.0675	0.1269	0.2161	0.3284	0.567	1.1651
Whole	0.0362	0.0924	0.1159	0.2340	0.1178	0.1413

Lessons Learned

- **Different way to improving subgraph match**. But we can't really compare CPI performant with the previous algorithm like VF2 plus because we didn't perform the framework matching and the best search ordering.
- **Great team work.** Great teamwork experiment during this final project. Help each other to focus and avoid confusing by the complex algorithm.
- Paper digest and implementation. Learn how to digest paper and implement the algorithm step by step.

References

- Bi F, Chang L, Lin X, Qin L and Zhang W. Efficient Subgraph Matching by Postponing Cartesian Products in International Conference on Management of Data, 1199-1214, July 2016.
- P. Zhao and J. Han. On graph query optimization in large networks.
 PVLDB, 3(1), 2010.

Thanks You!

Any questions?

