SIGMOD Programming Contest 2014

Efficient Implementation of Social Network Analysis System

team: H minor free

T. Akiba, T. Hayashi, S. Hirahara, T. Ikuta, Y. Iwata, Y. Kawata, N. Ohsaka, K. Oka, Y. Yano (The University of Tokyo, Japan)
R. Okuta (Tohoku University, Japan)

Construct a social network analysis system

Synthetic social networks

Generated by

social network benchmark dataset generator produced by Linked Data Benchmark Council

- Persons Vertices of the graph
 - Posts (Comments and Replies)
 - Interests
- Friendship between persons Edges of the graph

#persons ~ 1,000,000

Four types of queries Query1 Shortest distance over frequent communication paths Query2 Interests with Large Communities Query3 Socialization Suggestion Query4 Most Central People Contest goal

Execute all the queries as quick as possible

Evaluation Environment

- Processor 2.67 GHz Intel Xeon E5430
- Configuration 2 processors (8 cores total)
- ➤ L2 Cache Size 12MB
- ➤ Main Memory 15 GB

Overview of our system

- Implement one program for each query type.
- Run programs simultaneously.
- Parallel computing using OpenMP

 Query1 (1-2 threads), Query2(1 thread)

 Query3 (1 thread) Query4(6-8 threads)

Query1: Shortest distance over frequent communication paths

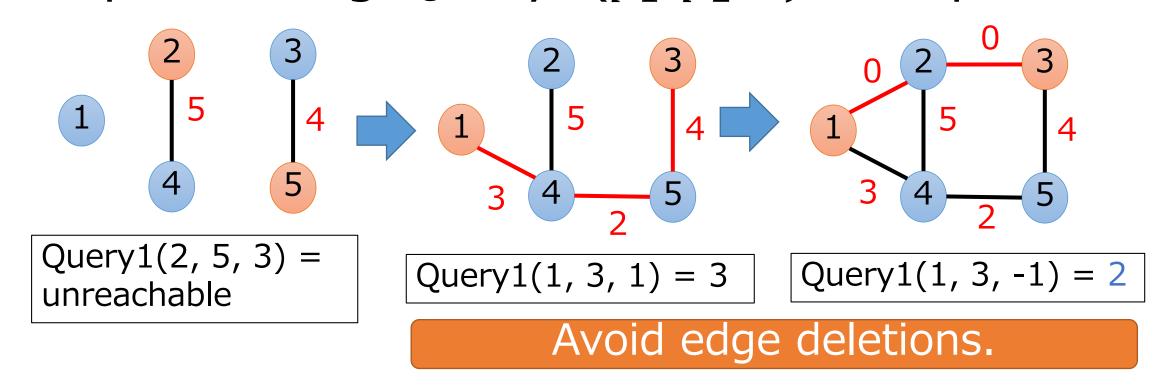
Query1(p_1, p_2, x)

- Answer the shortest path distance between persons p_1 and p_2 .
- Only use edges whose communication frequency is more than x.

Process the queries in the degreasing order of x.

- 1. Query1(1, 3, 1)
- 2. Query1(2, 5, 3)
- 3. Query1(1, 3, -1)
- 3) | -1)
- 1. Query1(2, 5, 3)
- 2. Query1(1, 3, 1) 3. Query1(1, 3, -1)

Before processing Query1(p_1, p_2, x), add passable edges.



Calculate communication frequency

The most time-consuming part is to compute the communication frequency $F(p_1, p_2)$ for all edges.

$$F(p_1, p_2) := \min (\#(p_1'\text{s replies to } p_2'\text{s comments}), \\ \#(p_2'\text{s replies to } p_1'\text{s comments}))$$

To compute $F(p_1, p_2)$, we must read two large files.

- comment_hasCreator_person.csv.
 - Pairs of a comment and a person who makes this comment.
- comment_replyOf_comment.csv.
 Pairs of a reply and a comment that receives this replycomment_replyOf_comment.csv

Linear time graph construction without sorting.

- Use the technique similar to bucket sort.
- Use the hash to get the person who makes the comment.
- Use SIMD instructions to achieve fast file reading.

Query3: Socialization Suggestion

Query3(k, h, p):

- Answer the top-k similar pairs of persons based on the number of common interest tags.
- For each of k pairs,
 - \triangleright The two persons must be located in p.
 - The shortest distance between two persons is not more than h.

Our solution

- 1. Compute the pairs with high similarity values
 - Guarantee two persons are located in p
 - Do not check the distance between them.
- 2.Check the distance between the pairs in the decreasing order of similarity values
 - Output first k pairs that the distance between two persons is not larger than h

Query2:

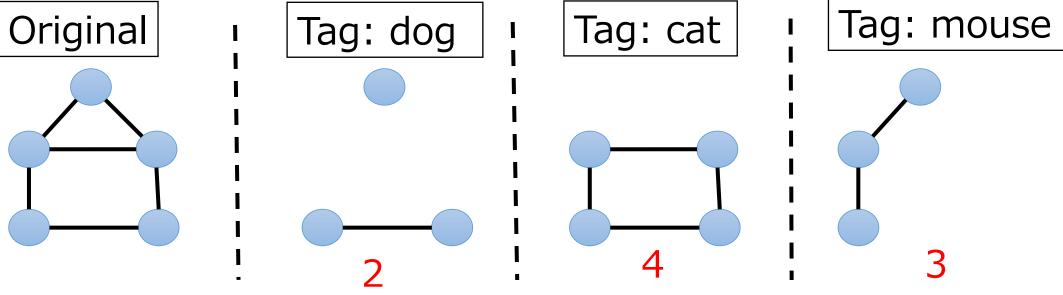
Interests with Large Communities

Query2(k,d):

Answer the k interests tags with the largest range.

The range of an interest tag t is the largest connected component in the graph induced by persons who

- Have the interest tag t
- Were born on d or later



Sort the queries in the decreasing order of d.

For each query Query2(k,d),

- Add persons whose birthday are d or later.
- Maintain the connected components for each tag t by using an Union-Find tree.

Query4: Most Central People

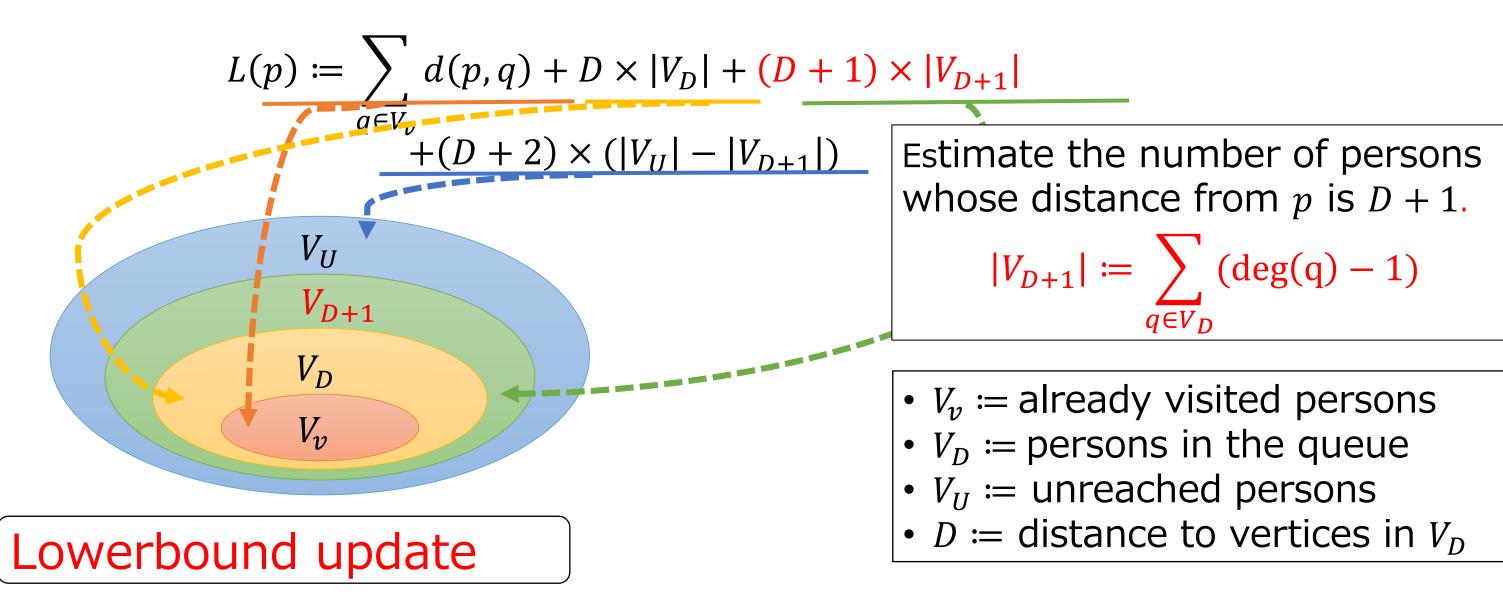
Query4'(k,t):

Answer k persons with the smallest sum of shortest distances in the induced subgraph depending on t.

Prune the BFS by using current top-k values.

- Conduct BFSs in the decreasing order of degree in the graph. For a person p, S(p) and deg(p) are strongly related.
- Estimate lowerbound of S(p) during the BFSs If the lowerbound of S(p) is large, the BFS is pruned.

How to estimate good lowerbound?



After scanning the edges from $q \in V_Q$,

