SIGMOD Programming Contest 2014

Efficient Implementation of Social Network Analysis System

Team: H_minor_free

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R. Okuta (Tohoku University, Japan)

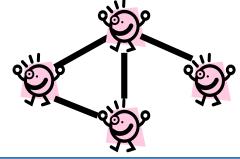
Speaker: Takanori Hayashi

Social networks in this contest.

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

Contest goal

Execute all the queries as quick as possible.

- 1. Shortest distance over frequent communication paths
- 2. Interests with Large Communities
- 3. Socialization Suggestion
- 4. Most Central People

Evaluation Environment

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

Our team H_minor_free

The University of Tokyo

T. Akiba Y. Iwata

Y. Yano Y. Kawata

Shortest distance query and

reachability query

PLL [SIGMOD'13]

PPL [CIKM'13]

PHL [ALENEX'14]

HPLL [WWW'14]

T. Ikuta T. Hayashi S. Hirahara

N. Ohsaka K. Oka

Good at implementing graph algorithms

Tohoku University

R. Okuta

5th place in SIGMOD Programming Contest 2013

Tasks of SIGMOD Programming Contest 2014 are queries on the social networks

Especially, Query1 is the shortest distance query!



We can win?

Our strategy

T. Hayashi

Y. Iwata Query1

Shortest Distance over frequent communication paths

T. Ikuta

Query3 S. Hirahara

Socialization Suggestion

Y. Yano

Query2

K. Oka

Interests with large communities

Y. Kawata

Query4 N. Ohsaka

Most central people

Analyze the networks:

T. Akiba

Optimize programs

R. Okuta



At March 18th, We became 1st place

Overview of our system

- Implement one program for each query type.
- Run programs simultaneously.



Total Elapsed Time

= max(total elapsed time of query type $i \mid 1 \le i \le 4$)

The most difficult query

At March 27th, (about 20 days before the deadline)

In the middle size network (#persons = 100,000),

- ➤ Query1 (1,000 queries) → 9 sec (2 threads)
- ➤ Query2 (10 queries) → 2 sec (1 thread)
- ➤ Query3 (9 queries) → 6 sec (1 thread)
- ➤ Query4 (8 queries) → 18 sec (8 threads)

Query4 is the most difficult query



Query4: Most Central People (Top-k Closeness Centrality)

Query4(k,t):

Answer k persons with the highest centrality values in the induced subgraph depending on a tag name t

The centrality value C(p) is defined as follows

$$C(p) \coloneqq \frac{(R(p) - 1)^2}{(n - 1) \times S(p)}$$

$$R(p) \coloneqq \#(\text{reachable persons from } p)$$

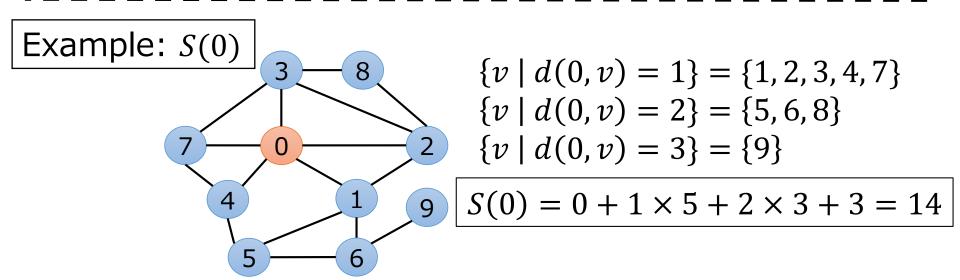
$$S(p) \coloneqq \sum_{q: reachable from p} d(p, q)$$

 $C(p) \propto \frac{1}{S(p)}$ in a connected graph.

Query4: Most Central People (Top-k Closeness Centrality)

Query4'(k,t):

Answer k persons with the smallest sum of shortest distances in the induced subgraph depending on a tag name t. $S(p) \coloneqq \sum_{q:reachable\ from\ p} d(p,q)$



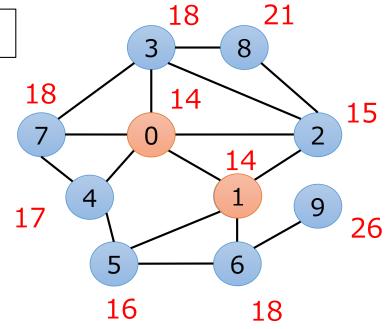
Query4: Most Central People (Top-k Closeness Centrality)

Query4'(k,t):

Answer k persons with the smallest sum of shortest distances in the induced subgraph depending on t a tag name S(n) = S

$$S(p) \coloneqq \sum_{q:reachable\ from\ p}\ d(p,q)$$

Example: Query4′(2, t)



 $Query4'(2,t) = \{0,1\}$

Naïve approach

 $n \coloneqq \#persons$ $m \coloneqq \#friendships$

Conduct BFSs from all persons.

- 1. S(p) can be computed in O(m) time.
- 2. Pick up k persons with highest centrality values.

Too inefficient

• O(nm) time. (n > 50,000 in large networks)

Our algorithm in Query4

Our algorithm should be

Exact

Answer correct values on any instance

Simple

Easy to parallelize Easy to optimize

Note: O(km) time algorithm is not known.

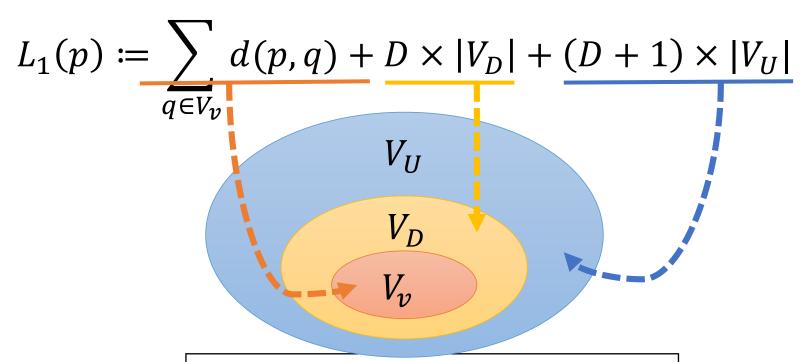
Our algorithm in Query4

Prune the BFS by using current top-k values.

- ➤ Conduct BFSs in the decreasing order of degree.
- 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?

Straightforward lowerbound



 $V_v := \#(\text{already visited persons})$

 $V_D := \#(\text{persons in the queue})$

 $V_U \coloneqq \#(\text{unreached persons})$

 $D := \text{distance to vertices in } V_D$

Straightforward lowerbound Example

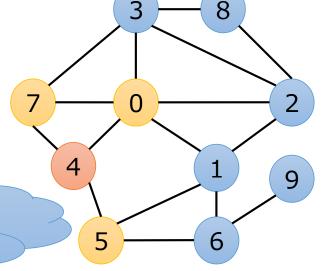
Example: BFS from a person 4

Assume
$$k = 2$$
 and $S(0) = S(1) = 14$

$$L_1(4) = 0 + 1 \times 3 + 2 \times 6 = 15 > 14$$

$$V_{v} = \{4\}$$

$$V_{D} = \{0, 5, 7\}$$

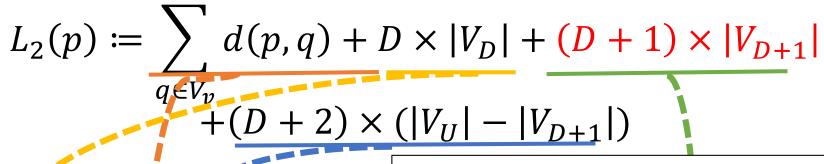


BFS from a person 4 is pruned!

Tighter lowerbound

 V_{D+1}

 V_D



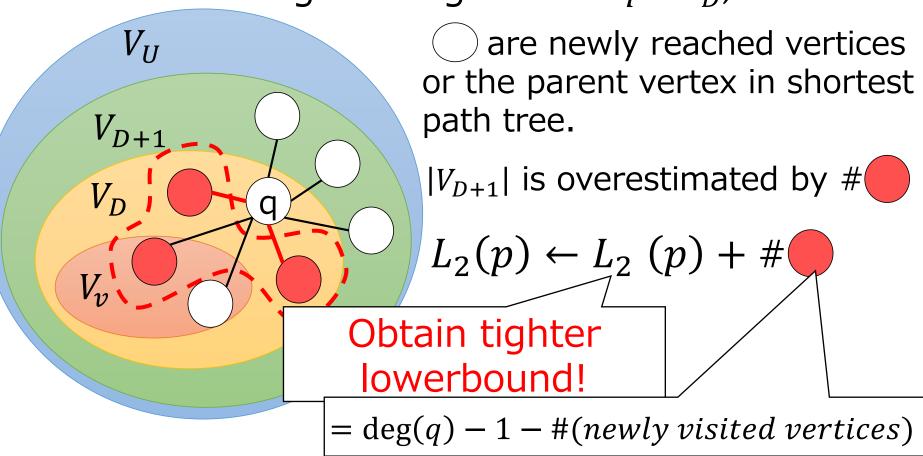
Estimate the number of persons whose distance from p is D + 1.

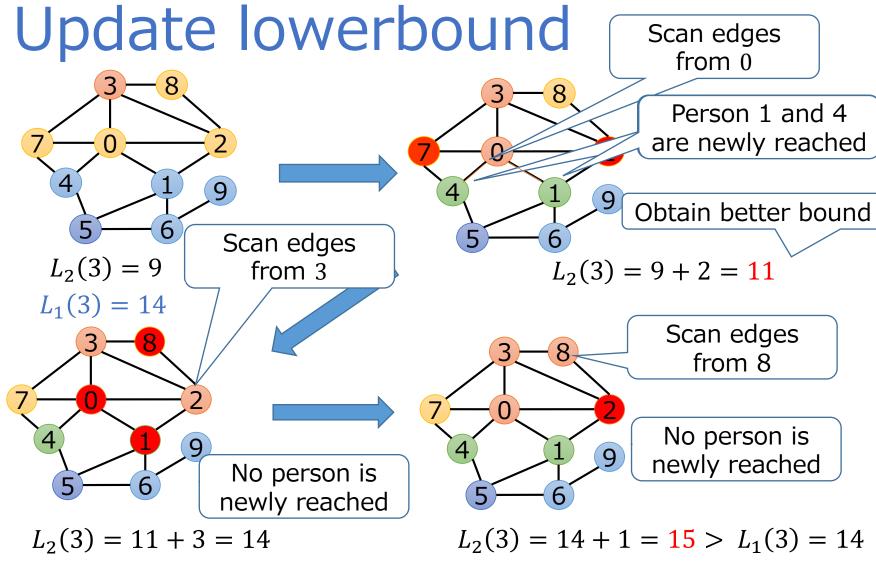
$$|V_{D+1}| \coloneqq \sum_{q \in V_D} (\deg(q) - 1)$$

 $V_v \coloneqq \#(\text{already visited persons})$ $V_D \coloneqq \#(\text{persons in the queue})$ $V_U \coloneqq \#(\text{unreached persons})$ $D \coloneqq \text{distance to vertices in } V_D$ $V_{D+1} \coloneqq \#(\text{persons whose})$ $V_{D+1} \coloneqq \#(\text{persons whose})$

Update lowerbound

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





BFS from a person 3 is pruned before scanning the edges from person 7!

Speedup by pruning

In 5,000 persons network induced by a tag *t* from 10,000 persons network,

Naïve: 425 ms

Pruning(L_1): 35 ms

Pruning(L_2): 28 ms

15x faster!

Results

In the middle size network (#people = 100,000),

Match 27th

Query1: 9 sec.

Query2: 2 sec.

Query3: 6 sec.

Query4: 18 sec.

April 15th

Query1: 6 sec.

Query2: 2 sec.

Query3: 2 sec.

Query4: 6 sec.

Structure of the system is still naïve...



Each program is written in the different way.



Summary T. Hayashi

Query1 Y. Iwata

Shortest Distance over frequent communication naths

Query3 T. Ikuta
S. Hirahara Socialization Suggestion

Y. Yano

K. Oka Query2

Interests with large communities

Y. Kawata

Query4 N. Ohsaka

Most central people

Analyze the networks

T. Akiba

Optimize programs

R. Okuta

At March 18th, 1st place

At April 15th, 3th place

- BFS with pruning
- Lowerbound estimation
- Other optimization