Speedup Graph Processing by Graph Ordering

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Graph Ordering

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Background

- Motivation
- Graph Access Patterns



Algorithm

- The GO Algorithm
- GO-PQ



Section 1

Background

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Subsection 1

Motivation

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Graphs are important

- Graph is a popular data model for Big Data social graph, web graph, knowledge graph
- Large graph is emerging with increasing size Friendship connection in Facebook, Twitter,

CPU cache performance is key issue in efficiency in DBS

 cache miss latency takes a half of the execution time in database systems

Design general optimization approach for graph model



Subsection 2

Graph Access Patterns

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- 1: for each node $v \in N_O(u)$ do
- 2: the program segment to compute/access v

Common Relationships: Neighbours / Sibling Key Idea here is that sibling relationship is a dominating factor.

$$\binom{d_O(u)}{2} \gg d_O(u)$$

Measure The Problem With Function

• Measure the Closeness of two points:

$$S(u,v) = S_s(u,v) + S_n(u,v)$$

• Our problems turns into a permutation problem!



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Section 2

Algorithm

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Theorem 2.1: Maximizing $F(\phi)$ to obtain an optimal permutation $\phi(\cdot)$ for a directed graph G is NP-hard.

The proof sketch is given in Appendix.

- All algorithms here are all approximate algorithms.
- If we set w = 1, the problem here is actually equivalent to the maximum traveling salesman problem, denoted as maxTSP for short.

Algorithm 1 $GO(G, w, S(\cdot, \cdot))$

1: select a node v as the start node, $P[1] \leftarrow v$; 2: $V_R \leftarrow V(G) \setminus \{v\}, i \leftarrow 2;$ 3: while i < n do 4: $v_{max} \leftarrow \emptyset, k_{max} \leftarrow -\infty$: 5: for each node $v \in V_R$ do i = 16: $k_v \leftarrow \sum S(P[j], v);$ $i = \max\{1, i-w\}$ 7: if $k_v > k_{max}$ then 8: $v_{max} \leftarrow v, k_{max} \leftarrow k_v;$ 9: $P[i] \leftarrow v_{max}, i \leftarrow i+1;$ 10: $V_R \leftarrow V_R \setminus \{v_{max}\};$

Theorem 3.1: The algorithm GO gives $\frac{1}{2w}$ -approximation for maximizing $F(\phi)$ to determine the optimal graph ordering.

	w=3		<i>w</i> =	= 5	w=7		
	F_{go}	\overline{F}_w	F_{go}	\overline{F}_{w}	F_{go}	\overline{F}_w	
Facebook	149,073	172,526	231,710	275,974	308,091	373,685	
AirTraffic	2,420	3,468	2,993	4,697	3,465	5,545	

Table 1: F_{go} and \overline{F}_w

• Go is expensive. Time complexity:

$$O(w \cdot d_{max} \cdot n^2)$$

Reasons for the inefficiency

 (1)Repeatedly computes score function w times for the same pair (v_j, v) while v_j is in the window of size w (2)it scans every node v in the set of remaining nodes VR in every iteration

Explain For What we do in the new algorithm



GOPQ algorithm

Algorithm 2 GO-PQ $(G, w, S(\cdot, \cdot))$

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1: for each node v \in V(G) do
2:
        insert v into Q such that \text{key}(v) \leftarrow 0;
3: select a node v as the start node, P[1] \leftarrow v, delete v from Q;
4: i \leftarrow 2;
5: while i \leq n do
6:
        v_e \leftarrow P[i-1];
 7:
        for each node u \in N_O(v_e) do
8:
            if u \in \mathcal{Q} then \mathcal{Q}.incKey(u);
9:
        for each node u \in N_I(v_e) do
10:
             if u \in \mathcal{Q} then \mathcal{Q}.incKey(u);
11:
            for each node v \in N_O(u) do
12:
                 if v \in \mathcal{Q} then \mathcal{Q}.incKey(v);
13:
         if i > w + 1 then
14:
             v_b \leftarrow P[i-w-1];
15:
            for each node u \in N_O(v_b) do
16:
                 if u \in \mathcal{Q} then \mathcal{Q}.decKey(u);
17:
            for each node u \in N_I(v_b) do
18:
                 if u \in \mathcal{Q} then \mathcal{Q}.decKey(u);
19:
                 for each node v \in N_O(u) do
20:
                    if v \in \mathcal{Q} then \mathcal{Q}.decKey(v);
21:
         v_{max} \leftarrow \mathcal{Q}.\mathsf{pop}();
22:
         P[i] \leftarrow v_{max}, i \leftarrow i+1;
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Implementation Of the Priority Queue



Figure 7: The Priority Queue: Q, Q_h , and top

- Lazy Update, use key and update instead of key
- Maintain the double link list until there should be changes
- Use the head and end to get the position for the first element and the last element of the subqueue with same key

Algorithm 3 decKey (v_i)

1: $update(v_i) \leftarrow update(v_i) - 1;$

Algorithm 4 incKey (v_i)

1: update
$$(v_i) \leftarrow$$
 update $(v_i) + 1$;

- 2: if update $(v_i) > 0$ then
- 3: update $(v_i) \leftarrow 0, x \leftarrow \overline{\text{key}}(v_i), \overline{\text{key}}(v_i) \leftarrow \overline{\text{key}}(v_i) + 1;$

4: delete
$$v_i$$
 from Q ;

- 5: insert v_i into Q in the position just before head[x];
- 6: update the head Q_h array accordingly;

7: if
$$\overline{\text{key}}(v_i) > \overline{\text{key}}(\text{top})$$
 then

8: $top \leftarrow v_i;$

Algorithm 5 pop ()

- 1: while update(top) < 0 do
- 2: $v_t \leftarrow top;$
- 3: $\overline{\text{key}}(v_t) \leftarrow \overline{\text{key}}(v_t) + \text{update}(v_t);$
- 4: update $(v_t) \leftarrow 0;$
- 5: if $\overline{\text{key}}(\text{top}) \leq \overline{\text{key}}(\text{next}(\text{top}))$ then
- 6: adjust the position of v_t and insert v_t just after u in Q, such that $\overline{\text{key}}(u) \ge \overline{\text{key}}(\text{top})$ and $\overline{\text{key}}(\text{next}(u)) < \overline{\text{key}}(\text{top})$;
- 7: $top \leftarrow next(top);$
- 8: update the head array;
- 9: $v_t \leftarrow top;$
- 10: remove the node pointed by top from Q and update top \leftarrow next(top);
- 11: return v_t ;

Section 3

Evaluation

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Implementation Of the Priority Queue



(a) $F(\cdot)$ by Different Orderings

Order	L1-ref	L1-mr	L3-ref	L3-r	Cache-mr
Original	11,109M	52.1%	2,195M	19.7%	5.1%
MINLA	11,110M	58.1%	2,121M	19.0%	4.5%
MLOGA	11,119M	53.1%	1,685M	15.1%	4.1%
RCM	11,102M	49.8%	1,834M	16.5%	4.1%
DegSort	11,121M	58.3%	2,597M	23.3%	5.3%
CHDFS	11,107M	49.9%	1,850M	16.7%	4.4%
SlashBurn	11,096M	55.0%	2,466M	22.2%	4.3%
LDG	11,112M	52.9%	2,256M	20.3%	5.4%
METIS	11,105M	50.3%	2,235M	20.1%	5.2%
Gorder	11,101M	37.9%	1,280M	11.5%	3.4%

Table 3: Cache Statistics by *PR* over Flickr (M = Millions)

Order	NQ	BFS	DFS	SCC	SP	PR	DS	Kcore	Diam
Original	50.8	15.3	5.4	7.8	21.5	52.1	21.9	20.8	14.9
MINLA	51.8	18.0	5.5	8.1	24.6	58.1	22.1	21.5	17.9
MLOGA	41.7	16.3	5.1	7.2	21.9	53.1	21.1	20.6	16.4
RCM	49.1	12.1	4.6	6.6	15.9	49.7	20.3	20.2	12.4
DegSort	45.7	16.7	4.8	7.0	24.9	58.3	21.4	18.6	17.0
CHDFS	42.1	12.3	4.1	5.8	18.5	49.9	21.1	20.6	12.9
SlashBurn	46.2	16.0	4.5	6.2	22.1	55.0	20.7	21.3	15.8
LDG	50.7	15.9	5.8	8.2	21.8	52.9	22.4	21.2	14.9
METIS	63.0	18.2	7.7	10.1	20.8	50.3	23.0	21.7	16.7
Gorder	35.4	11.1	3.6	5.2	12.8	37.9	18.7	18.1	10.9

Table 6: L1 Cache Miss Ratio on Flickr (in percentage %)

Order	Pokec	Flickr	LiveJ	wiki	G+	pld	twitter	sd
Original	1,187	750	3,040	10,503	23,128	21,961	48,238	50,784
MINLA	1,176	843	3,471	10,322	25,543	20,698	44,536	48,218
MLOGA	1,181	787	3,427	10,121	24,229	20,829	45,698	48,899
RCM	1,091	673	2,883	7,272	20,371	18,657	40,730	36,334
DegSort	1,188	815	3,281	9,500	24,799	21,278	44,228	47,723
CHDFS	1,107	690	2,934	7,600	21,150	17,732	39,517	36,585
SlashBurn	1,219	810	3,452	10,031	24,616	21,564	44,261	45,134
LDG	1,168	793	2,940	10,137	23,569	22,740	47,841	55,234
METIS	1,131	843	3,232	-	-	-	-	-
Gorder	1,003	620	2,556	5,932	17,936	14,389	32,808	30,202

Table 10: Running Time of Diam (in second)

The End

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