Assignment 6 1

UNIVERSITY OF ILLINOIS AT CHICAGO, INFORMATION AND DECISION SCIENCES

MapReduce-based PageRank Algorithm

1 ALGORITHM DESCRIPTION

In this assignment, you will implement a MapReduce-based commonly-used web link analysis algorithm: PageRank.

The input: Graph G and parameter β . The Graph G is directed without having dead end nodes. You can either write codes to generate such a graph or manually create it. Common values for β are in the range of 0.8 to 0.9. In the case of spider traps, the random surfer follows a link at random with probability β .

The output: PageRank vector \vec{r} , each component would be the ranking value of a node in the graph G.

The algorithm: The stopping criteria is that the change of PageRank vectors between current iteration and the previous iteration is less than a very small value ϵ (for example, $\epsilon = 0.05$. The sequential version of PageRank implementation (pseudo code) is shown below.

2 PSEUDO ALGORITHM FOR SEQUENTIAL VERSION

PageRank algorithm on a directed graph G

$$set: r_i^{(0)} = \frac{1}{N}, t = 1$$

N is the number of nodes in the graph G

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(1) $\forall j : r_i^{'(t)} = \sum_{i \to j} \beta \frac{r_i^{(t-1)}}{d_i}$

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r_j^{'(t)} = 0 \text{ if in-degree of } j \text{ is } 0
(2) \text{ Now re-insert the leaked PageRank:}
\forall j: r_j^{(t)} = r_j^{'(t)} + \frac{1-S}{N} \text{ where: } S = \sum_j r_j^{'(t)}
(3) \ t = t+1 \text{ od}
\text{while } \sum_j |r_j^{(t)} - r_j^{(t-1)}| > \epsilon \text{ od}
end
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3 PSEUDO ALGORITHM FOR MAPREDUCE VERSION

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/ ★ The Mapper is to invert the input ★ /
Mapper:
     \forall page_i \in (page_1, page_2, \cdots, page_k)
          output page_j \rightarrow \langle page_i, \frac{rank_i}{d_i} \rangle / / d_i is degree of node i.
     output page_i \rightarrow page_1, page_2, \cdots, page_k
/ \star The Reducer is to update the ranking using the in-links \star /
Reducer:
     Input is in a format of \triangle. The key: page_k
     \forall in-link page_i \in (page_1, page_2, \dots, page_n)
          rank_k + = \frac{rank_i}{d_i} * \beta
     output\langle page_k, rank_k \rangle \rightarrow \langle page_1, page_2, \cdots, page_n \rangle
     // page_1, page_2, \cdots, page_n are out-links of page_k.
After map function, we have temporary files in the following structure (\triangle):
page_k \rightarrow \langle page_1, rank_1 \rangle,
            \langle page_2, rank_2 \rangle,
            .....,
            \langle page_n, rank_n \rangle,
            \langle page_{k1}, page_{k2}, \cdots, page_{kn} \rangle
where page_1, page_2, ..., page_n are the in-links of page_k,
and page_{k1}, page_{k2}, ..., page_{kn} are the out-links.
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4 SUBMISSION INSTRUCTION

You need to submit one zip file, which contain all source codes (*.java files) and README file. The name of your zip file would be following the format like: Firstname_Lastname_Assignment_6.zip. Please follow this format and submit it through the blackboard. Please try to comment your codes for critical sections and make your codes as readable as possible.