ELEN 4020 Project - Comparison of Parallel

Equi-Join using MPI and OpenMP

Uyanda Mphunga - 1168101, Darren Blanckensee - 1147279, Ashraf Omar - 710435 and Amprayil Joel Oommen - 843463

Abstract—

I. Introduction

The join operation concerns the combining of two different tuples on a common join attribute [1]. It is important in information systems, in particular the use of databases since the join operation is the most expensive operation in database query operations in terms of time and data-intensity [2]. This project explores two different approaches to performing a join of two very large tables. The problem is described in Section II and the two different approaches discussed in Sections IV and III. The method of experimentation is outlined in Section V and the results of the experiment are analysed in Section VI. Section VII reviews the project and concludes this document.

II. PROBLEM DESCRIPTION

The objective of this project is to perform an equijoin of two very large tables. An equi-join is a type of join that uses the equality operator as a basis for the join [3] that is if the join attribute in $R_1(A, B)$ is strictly equal to the join attribute in $R_2(A, C)$ the result of the join is inserted into a third table $R_3(A, B, C)$. An illustrated example can be seen in Figure 1. The join

School of Electrical & Information Engineering, University of the Witwatersrand, Private Bag 3, 2050, Johannesburg, South Africa

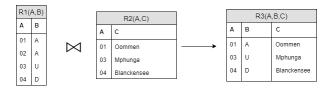


Fig. 1. An example of equi-join between two relational tables

must be done using two different algorithms, one that is based in MPI (Message Passing Interface), and one that uses another high-level parallel programming model. The programming model chosen in this project is OpenMP. The two programs need to be compared with one another in terms of speed-up and scalability when increasing the number of processors and nodes that the program uses. The speed-up comparison is performed by running the two programs with increasing number of processors and the scalability comparison is conducted by increasing the number of nodes the program uses in a cluster. Based on the literature, there are multiple join algorithms. The algorithms chosen by the authors are the hash algorithm and the merge algorithm, which form part of the equijoin family of join algorithms and are widely used [4].

The join algorithms are designed to read data from file of a specific format and to perform the required operations as per specific algorithm. The algorithms read data of the same file in order to be able to compare their performance. The algorithms are both implemented in C++. This allows for accurate comparisons of the algorithm performances and consequently prevents the possible time lags that may arise from the use of different programming languages. C++ was used instead of C due to the authors finding easier to use C++ under the given time constraints. Another reason that why C++ was found more favourable is that is offers more functionality than C and due to its Object Orientated Design (OOD) capabilities and the also due to the fact that the authors are familiar with C++. These algorithms were implemented using parallelism.

III. OPENMP JOIN ALGORITHM - MERGE-JOIN

The OpenMP join algorithm being implemented is merge-join also known as sort-merge join [5]. The two table are sorted by join attribute. Then the table are scanned and join attributes are compared to one another. For an equi-join, if the join attribute of one entry is strictly equal to the join attribute of the other table's entry, the entries are joined in the output table [6], [7]. The implemented algorithm assumes that the sorting of the data has already taken place and proceeds to only perform the join algorithm under that assumption.

References [4], [8] take into account possible errors that may occur due to data skew for the merge join algorithm. According to [4], data skew can result in some of the processors being over-utilised and others underutilised. Reference [4] proposes the use of a divide-and-conquer technique as a solution to handle data skew. Reference [8] approaches the issue of data skew in the merge algorithm by adding an extra scheduling phase. The implemented merge join does not take into account the possibility of data skew and instead relies on OpenMP's computational ability to process certain chunks of data and ability to make numerous threads [9]. This technique was chosen due to the resources available, namely a cluster, and time constraints.

IV. MPI JOIN ALGORITHM - HASH-JOIN

A hash join can be generalised into two stages [10], the split phase and the join phase [11]. The reviewed literature for the hash join is also concerned about data skew. Some of the said literature include references [8], [10] and other documents. Reference [8] addresses the issue of data skew by implementing multiple hash phases along with the use of a heuristic optimisation algorithm in order to isolate skew elements[8].

An overview of the algorithm being used is listed in Algorithm 1.

```
Data: Table R_1(A,B), Table R_2(A,C)

Result: R_3(A,B,C)

for all key-value pairs in R_1 do

read key;

calculate hash as sum(ASCII) of key;

calculate (hash%(no. of processes - 1) + 1);
```

end

send key-value pairs of $R_1(A, B)$ to slave nodes in vector;

for all key-value pairs in R_2 do

```
read key;
calculate hash as sum(ASCII) of key;
calculate (hash%(no. of processes - 1) + 1);
//This is to determine which slave node the
key-value pair is sent to.
```

end

send key-value pairs of $R_2(A, C)$ to slave nodes in vector;

IN EACH SLAVE NODE:

for all key-value pairs in each vector do

if key from $R_1 == key$ from R_2 then

write key-value output to R_3 end

end

Algorithm 1: MPI Implementation of hash-join

V. EXPERIMENT DESCRIPTION

VI. ANALYSIS OF RESULTS

VII. CONCLUSION

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