Query Optimization in a Machine With Parallel Structure Using with Genetic Algorithm

Javad Sohafi-Bonab\textsuperscript{1}, Alireza Hajieskandar\textsuperscript{2}, Fattaneh Taheri Ashtiyani

Javad Sohafi-Bonab\textsuperscript{1}: Department of computer engineering, Bonab Branch, Islamic Azad University, Bonab, Iran
Alireza Hajieskandar\textsuperscript{2}: Department of computer engineering, Bonab Branch, Islamic Azad University, Bonab, Iran
Fattaneh Taheri Ashtiyani\textsuperscript{3}: Department of computer engineering, Bonab Branch, Islamic Azad University, Bonab, Iran

Abstract: Query Optimization is one of fundamental problems in grid databases. Especially, when the databases are replicated and stored in different nodes of the network, with regard to the point that query in grid databases can be processed in different sites. The problem of choosing suitable sites to execute query is very important. In this paper, we consider the queries such that including cycle and requiring a lot of times to respond to the query. We use two methods for optimizing such queries. First, by following an algorithm, we decompose graph to several trees, then distribute the trees on computers, and finally by using Genetic Algorithm, optimize it. Second, without decomposition the graph, by distributing it on computers, and then by using Genetic Algorithm, we optimize the query. After ten times executing the algorithms, we conclude first method gets better result than the second.

Keywords—Database, Query Optimization, Genetic Algorithm, Query graph, Parallel Machine

1. Introduction

Query optimization is the process of choosing the most efficient query evaluation plan from among the many strategies usually possible for processing a certain query, especially if the query is complex.

Success mystery of a DBMS especially if we deal with relation model is a optimal efficiency of a query system. In this system, an input is a query like q for DBMS by user.

Suppose S is a set of all possible strategies to response to query of q, each member of S likes has cost C(S).(regarding cpu & i/o). Aim of optimization of an algorithm is finding of one member of s\textsubscript{0} from S as\cite{13}:

\[ C(s_0) = \min_{s \in S} C(s) \]

Now we continue discussion about relation database. One strategy for being responsible for query of q, use sequential algebra operation for database relationship to find an answer for q.

Cost of one strategy is total cost of process of each operator. Among these operators the difficult of them process and optimization is join operator [which is present by \( \bowtie \) or join] This operator received two relation as input then with combination of tuples and regarding of type join produce a new relation as an output Join is a portability and associative operator.

So number of strategies for response to an query increase exponentially with number of join.\cite{2}\cite{3}\cite{4}\cite{12}\cite{15}. Generally for N relation number of possible combination is equal with \( 2(n-1)!/(n-1)! \).Due time complexity formula, N relation is equal with \( \Omega(N!) \).Because optimization of query which contain of join operator is very time consuming rather than other relation operators like select operator or project operator. Basically all optimization query algorithms related to queries with join \cite{2}\cite{4}.In modern applications of each query is usually include a few relation such as 10 relationship\cite{12}.But it is prognosticated in future application, number
of relations would be more than 16. At present optimization query techniques for support some application of database are unsuitable. For example system-R is not able to performance with more than 16 join even for less than 16 join query, it encounter is with problem. [1] It sometimes encounters a proposal design to solve problem (query optimization) is to apply random algorithm such as SA(Simulated Annealing), II(Iterative Improvement) and 2PO a combination of II and SA which up to now would be successful for query optimization and provide enough reason to apply. According to experiments for small query with 5 to 10 join there is not any difference among three random algorithms.

Basically when the size of query increases so improvement of 2PO out put is comparison able with SA and also with II. Even though at the similar time average cost of out put strategy of all algorithms has less resistance. This present that there are many cases which these algorithms loss its best state. All in All (for all that), among these three algorithms 2PO is fairly more resistance. Totally random algorithms was acceptable for optimization and also for presented problem obtain better response rather than other method.[6][11][14][18][19]

When these Algorithms were in recursive query they will answer very well. But in case of being recursive these Algorithms spend a lot of time to response (we can imagine an equal graph for every query). In fact the problem can be put forward in this way:

When query graph be a tree graph these Algorithms are suitable approach for optimization. But if query graph be a graph these Algorithms with present structure encounter with problem. Therefore to optimization a large query which its query graph include cycle, genetic algorithm is used.

In this paper we use two methods for optimization for large queries which their query graph include a cycle. In first method we have presented an algorithm according depth search which with proposal algorithms query graph decompose to several trees and have performed every tree with application of Genetic algorithm distributed in parallel on a network of computers which connected in ring state and best solution acquired of union of one by one solutions and in second method performed whole query graph without decomposition to trees with using parallel genetic algorithm distributed. Contents of this paper organized as below: in section (2) an algorithm is presented according deep search for decomposition of query graphs, in section (3) a parallel machine model is presented for finding optimization case of join relations, and in section (4) a parallel genetic algorithm is presented for finding optimum case, and in section (5) computing environment which program is executed on environment, and in section (6) result of experiments will expressed and finally summary and future task will be expressed.

2. An Algorithm Base On Deep Search For Decomposition Of Query Graph

In this paper with use of depth method search which discussed in [20][9], we presented an algorithm which by that we can decompose one graph include cycle to several trees, so apply this obtained trees by mentioned method[10] for response to query.

\[(S \preceq T) \text{ and } (S \succ P) \text{ and } (P \preceq H) \text{ and } (H \preceq T) \text{ and } (H \succ R) \text{ and } (P \succ M) \text{ and } (M \preceq R) \text{ and } (T \preceq R)\]

fig 1 Imagine query figure

For each query related query graph can be obtained. In this graph nodes show relation, edges show join of two relation, and numbers on edges show types of relation of two join.

Figure (2) show query graph(1).

Considering fig(2), it will be clear that this query graph include cycle, so response time to this query in comparison with those queries which graph of them is a tree, will be long.[14]
To solve this “query graph” we decompose query, that is, we will find a tree with elimination of existing cycle of graph. Note that after decomposition of graph some obtained trees may have cycle (be a graph). In this case we will continue above process to receive a jungle at the end.

Presented algorithm to this problem emanate from a depth search algorithm. Depth search algorithm is used to measurement of a graph. In fact this algorithm doesn't consider met nodes. So output of this algorithm will be a tree.

Appointed algorithm will act in this way:

We made a list of joins from its near nodes for every vertex. Every node of this join list include of a vertex and a join which show type of join.

We make array of above join list called graph (number of vertex) which include this join lists. In this algorithm, first a vertex like, v, is met and enter the weight, if weight of this vertex have not entered in a tree which call tree. Then consider vertex (w) near to this vertex, if this vertex don't be met already we meet it and enter its weight in related array. Otherwise eliminate it from stack and continue this process till stack be empty. At end all vertexes is met and a list called “tree” will obtained.

```
Tdfs( integer v )
{Node_ptr w ;
Visited [ v ] =TRUE ;
print( v ) ;
if ( ! Search_Weight ( w->join , tree )
   tree [I++] = w->join ;
for ( w = gragh [ v ]  ;  w  ;  w = w->next )
if ( ! visited [ w->vertex ]
   Tdfs ( w->vertex ) ;
}
```

Out put vertexes by this algorithm can be match to each other by join numbers between vertexes. The result will be a tree. If acquire remain numbers which there aren’t in array tree and match them to each other considering joined numbers, we succeed to decompose graph to several trees.

We should note that obtained trees have not cycle otherwise continue above process again till receive to a jungle. Now survey query (1) with this algorithm. Considering fig (1) ,proximity list of query graph will be like fig3:

```
M 1 p 8 R null
S 2 p 6 T null
P 1 M 2 S 3 H null
H 4 P 5 T 5 R null
T 6 S 4 H 2 R null
R 8 M 5 H 7 T null
```

Fig (3) is presentation of link list of query graph(1).Applying Tdfs algorithms for query graph, array Tree will take following numbers:

```
1 2 6 3 8
```

and met vertexes will be as below:

M ,P ,S ,T ,H ,R

Considering content of “array tree” and met vertexes can make query tree that has come in the fig 4:
Can be seen that cycle of query graph of fig (1) is eliminated and the result is a tree.

![Query Graph](image)

Fig.4: the resulting query graph

Now if consider remain weights of tree then we will have:

\[ 4, 5, 7 \]

Obtained graph or tree is like:

![Obtained Graph](image)

With continuing of above process, graph will decompose to two trees. So according to fig (2) we could decompose query (1) to three tree.

3. Studying Of Probable Problems

It is possible with increase of number of query relations number of obtained trees would be more. So below solution is advised:

Obtained trees which make jungle except main tree could change to one or some larger tree (max 16 join).

We can do this task by below method:

If there wasn’t any join between two relations with temporary label of zero connect two trees. This mechanism will be repeated till number of joins achieved to 16.

For remain trees above mechanism will be repeated.

4. Parallel Machine Model

Since proposal genetic algorithm will performed in network of processors, first we deal with structure which algorithm have done on it and we have concluded acceptable results. We suggest a parallel distributed execute, of parallel algorithms in multi-processors which every processors has its own local memory (not shared memory) and each processors do a genetic operation on set of its chromosome and relation with neighbors is minimum. Circumstances of distributed parallel execution is completely described. In [5][7][8][18].

To demonstrate mentioned method we use of eight computers which are connected in ring state to optimization queries with cycle in their queries in computer network. Advantage of using network with ring topology against structures with mesh on or hypercube in parallel distributed execute from genetic algorithm is completely described in [5].

In this part we turn to parallel distributed detail on multi processor system.

Since distinct execution of genetic algorithm on processors is not effective so we use transfer of chromosome (which we call migration) between neighbor processors. Two processors are neighbors if there was a direct link. In order select a chromosome for migration to distinct neighbor before producing next generation and after
genetic operation like crossover and mutation, we use of two method of selection of high qualities chromosome, and random way of selection to migrate. In this method before producing next generation of processor, select a chromosome by mentioned selection method and migrate to distinct neighbor. If quality of this sent chromosome was better than neighbor processors chromosome, accept it and otherwise reject it.

Also of next generation selection method in proposal algorithm is done by using SUS method. Reason for selection of SUS method to select next generation is possibility of select of first chromosome in first population which avoid rapid convergence to solution of problem. In fact algorithm engages less in local optimum solutions.

5. Application Of Ga In Parallel

At the beginning of execution, each processor produce set of its chromosome based on GA parameters (mutation and crossover) which is sent from host processor then each processor operate genetic operation like crossover, leaping on its own population in each generation similar to ordinary genetic algorithm [17]. Before producing next of generation or action selection, moving chromosome are done by neighboring processors, if contain the condition of displacement. That we call migration condition.

We introduce migration condition which each processor send high quality or random chromosome to its neighbor and if quality of received chromosome was better than quality of its chromosome. It will accept it and or not it will reject chromosome. Procedure for each processor is like below:

Procedure PGA:
begin
Receive GA parameters and a signal for starting execution from the host processor;
Initialize a population;
repeat
Evaluate all the chromosomes;
If the migration condition holds then
Take place the chromosomes migration;
endif;
Select the chromosomes for the next generation by the SUS way with the elite strategy or random;
Create new chromosomes by applying Crossover rate the mutation rate;
until the termination condition holds;
Send result to the host processor;
end;

6. Applied Computation Environment

All computation in a network environment with Linux operating system is described in table (1). Proposal genetic algorithm is implemented by using c++ and v2.0 is used as a interface to transfer messages and communication between genetic algorithms between different computers.
7. Result of Experiments

Before expressing result of experiments we turn to some applied expressions and needed parameters to execute genetic algorithms.

In migration step, some chromosome migrates from a sub population to neighbor sub population. We migration frequency by \( M_{freq} \) which described as a number of generation. For example if \( M_{freq} = 4 \) chromosome will migrate to neighbor one time every four generation. Also we show number of migration by \( M_{rate} \). For example if it was 5 it shows in each migration five chromosome are migrated we also use selection SUS method to select chromosome for next generation and use of two method of best selection of chromosome or random selection to choice a chromosome to transfer to neighbor. Also genetic algorithm spam show number of produced generation and \( N_{relation} \) use for number of applied relations to optimization. Fix applied parameters experiment is shown below:

- Generation span = 100
- Crossover rate = 0.8
- Mutation rate = 0.01
- \( M_{rate} = 1 \)
- Select method = SUS
- Migrated chromosome is a best fitness chromosome or random
- \( N_{relation} = 8 \)

We have supposed below parameters to better evaluate of found solutions variable:

- Number of chromosome per sub population = 3, 6, 8
- \( M_{freq} = 1, 4 \)

As it is already mentioned we use two methods to optimization query for query graph. Each method is described separately and we turn to result of them at below.

In first method query graph is decomposed with using Tdfs algorithm to several tree and each tree is solved by mentioned parameters with using PGA algorithm. Finally after execution of all trees final results obtained of union of answers. For example one method to select parameters of variable genetic algorithms to execute can be as below:

- Number of chromosome per sub population = 3
- \( M_{freq} = 4 \)

This means size of primary population in each processor 3, and every four generation selection of chromosome to migrate is done. Obtained result of execution of first method is shown in table 2, 3.
method query graph execute without decomposition to tree with using mentioned parameters in first method and results shown in table 4.5.

By comparison of obtained result we can understand optimal result is obtained first method with four chromosome parameter in each sub population and \( M_{freq} = 4 \).

### 8. Conclusion

Because time of execution of query graph with cycle is longer than other relation query graph we turn to solve query graph on network with using two proposal methods and applying distributed parallel genetic algorithm with different parameters. In first method we present an algorithm that we decompose graph with cycle and change it to different tree and then distribute every tree separately in executing processors and each processors with using genetic algorithm which is called PGA we would be able to find optimal case. Final results obtained at the end from union of results. In second method without decomposition of query graph to tree we solve it with using genetic algorithm. Also applied parameters in distributed parallel genetic algorithm is divided to two fix and variable part, fix parameters in all execution of two mentioned method in paper are fix and presented result in this paper is based on variable parameters for example migration rate and number of chromosome in primary population.

We execute to presented method in this paper ten times and separately. In first time we suppose migration frequency of chromosome one and execute genetic algorithm with using number of existing chromosome in primary population. Second time migration frequency chromosome is supposed four and execute genetic algorithm as before from obtained result and comparing them we found out that in first number of chromosome in primary population, most primary solutions in obtained and time of finding results is very worthy in comparison with ordinary genetic algorithm.

### TABLE II: Number of found solution for the method of without decomposition query graph to tree

<table>
<thead>
<tr>
<th>AVG</th>
<th>RUN9</th>
<th>RUN8</th>
<th>RUN7</th>
<th>RUN6</th>
<th>RUN5</th>
<th>RUN4</th>
<th>RUN3</th>
<th>RUN2</th>
<th>RUN1</th>
<th>RUN0</th>
<th>#sub pop</th>
</tr>
</thead>
<tbody>
<tr>
<td>2287.2</td>
<td>2658</td>
<td>2457</td>
<td>2036</td>
<td>1989</td>
<td>2785</td>
<td>2217</td>
<td>1957</td>
<td>2652</td>
<td>2014</td>
<td>2134</td>
<td>3</td>
</tr>
<tr>
<td>2312.4</td>
<td>2123</td>
<td>2583</td>
<td>2042</td>
<td>2145</td>
<td>2004</td>
<td>2954</td>
<td>2471</td>
<td>2652</td>
<td>2145</td>
<td>1996</td>
<td>4</td>
</tr>
<tr>
<td>2056</td>
<td>1891</td>
<td>2486</td>
<td>1652</td>
<td>1980</td>
<td>2104</td>
<td>2753</td>
<td>1932</td>
<td>2037</td>
<td>1584</td>
<td>2141</td>
<td>8</td>
</tr>
</tbody>
</table>

### TABLE III: Number of found solution for the method of without decomposition query graph to tree

<table>
<thead>
<tr>
<th>AVG</th>
<th>RUN9</th>
<th>RUN8</th>
<th>RUN7</th>
<th>RUN6</th>
<th>RUN5</th>
<th>RUN4</th>
<th>RUN3</th>
<th>RUN2</th>
<th>RUN1</th>
<th>RUN0</th>
<th>#sub pop</th>
</tr>
</thead>
<tbody>
<tr>
<td>2432.2</td>
<td>2984</td>
<td>2387</td>
<td>2057</td>
<td>2139</td>
<td>2785</td>
<td>2217</td>
<td>2853</td>
<td>2541</td>
<td>2225</td>
<td>2134</td>
<td>3</td>
</tr>
<tr>
<td>2699.4</td>
<td>3047</td>
<td>2883</td>
<td>2143</td>
<td>2314</td>
<td>2594</td>
<td>2957</td>
<td>2987</td>
<td>2647</td>
<td>2965</td>
<td>2457</td>
<td>4</td>
</tr>
<tr>
<td>2570</td>
<td>2994</td>
<td>2967</td>
<td>2035</td>
<td>2019</td>
<td>2636</td>
<td>2753</td>
<td>2876</td>
<td>2738</td>
<td>2541</td>
<td>2141</td>
<td>8</td>
</tr>
</tbody>
</table>

### TABLE IV: Number of found solution for the method of without decomposition query graph to tree

<table>
<thead>
<tr>
<th>AVG</th>
<th>RUN9</th>
<th>RUN8</th>
<th>RUN7</th>
<th>RUN6</th>
<th>RUN5</th>
<th>RUN4</th>
<th>RUN3</th>
<th>RUN2</th>
<th>RUN1</th>
<th>RUN0</th>
<th>#sub pop</th>
</tr>
</thead>
<tbody>
<tr>
<td>183</td>
<td>206</td>
<td>204</td>
<td>159</td>
<td>147</td>
<td>98</td>
<td>145</td>
<td>258</td>
<td>180</td>
<td>243</td>
<td>190</td>
<td>3</td>
</tr>
<tr>
<td>501</td>
<td>281</td>
<td>286</td>
<td>212</td>
<td>207</td>
<td>236</td>
<td>264</td>
<td>284</td>
<td>247</td>
<td>230</td>
<td>203</td>
<td>4</td>
</tr>
<tr>
<td>488.5</td>
<td>512</td>
<td>542</td>
<td>465</td>
<td>500</td>
<td>498</td>
<td>563</td>
<td>502</td>
<td>498</td>
<td>398</td>
<td>407</td>
<td>8</td>
</tr>
</tbody>
</table>

### TABLE V: Number of found solution for the method of without decomposition query graph to tree

<table>
<thead>
<tr>
<th>AVG</th>
<th>RUN9</th>
<th>RUN8</th>
<th>RUN7</th>
<th>RUN6</th>
<th>RUN5</th>
<th>RUN4</th>
<th>RUN3</th>
<th>RUN2</th>
<th>RUN1</th>
<th>RUN0</th>
<th>#sub pop</th>
</tr>
</thead>
<tbody>
<tr>
<td>119.6</td>
<td>114</td>
<td>107</td>
<td>146</td>
<td>151</td>
<td>101</td>
<td>92</td>
<td>141</td>
<td>123</td>
<td>123</td>
<td>98</td>
<td>3</td>
</tr>
<tr>
<td>189.1</td>
<td>296</td>
<td>201</td>
<td>195</td>
<td>203</td>
<td>163</td>
<td>196</td>
<td>175</td>
<td>178</td>
<td>142</td>
<td>141</td>
<td>4</td>
</tr>
<tr>
<td>572.6</td>
<td>580</td>
<td>601</td>
<td>578</td>
<td>599</td>
<td>587</td>
<td>537</td>
<td>612</td>
<td>596</td>
<td>532</td>
<td>504</td>
<td>8</td>
</tr>
</tbody>
</table>

http://dx.doi.org/10.17758/UR.U0915127
9. References

[12] Y.E.Ioannidis Query Optimization,supported by Oracle ,IBM, Research 1996
http://dx.doi.org/10.1145/38713.38722