# **Query Optimization Issues for Data Retrieval in Cloud Computing**

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#### Abstract

Cloud data storage redefines the issues targeted on customer's out-sourced data (data that is not stored/retrieved from the availability [5], the reports of privacy breach and service customers own servers). In this work we observed that, from a customer's point of view, the data need to be accessed with in no time the user given the request ,Even though the data is stored on cloud severs the effective query optimizations may not be defined to access the data in an efficient way. This paper proposes the efficient available Query optimization techniques for efficient retrieval of data to satisfy the customer needs.

**Keywords:** Cloud computing, storage, Cloud service provider, Query Optimization.

#### Introduction

The industrial information technology towards a subscription based or pay-per-use service business model known as cloud computing. This paradigm provides users with a long list of advantages, such as provision computing capabilities; broad, heterogeneous network access; resource pooling and rapid elasticity with measured services .Huge amounts of data being retrieved from geographically distributed data sources, and non-localized data-handling requirements, creates such a change in technological as well as business model. One of the prominent services offered in *cloud computing* is the *cloud* data storage, in which, subscribers do not have to store their own data on their servers, where instead their data will be stored on the cloud service provider's servers. In cloud computing, subscribers have to pay the providers for this storage service. This service does not only provides flexibility and scalability data storage, it also provides customers with the benefit of paying only for the amount of data they needs to store for a particular period of time, without any concerns of efficient storage mechanisms and maintainability issues with large amounts of data storage[3]. In addition to these benefits, customers can easily access their data from any geographical region where the Cloud Service Provider's network or Internet can be accessed [1]. An example of the cloud computing is shown in Fig. 1. Since cloud service providers (SP) are separate market entities, data integrity and privacy and retrieval are the most critical issues that need to be addressed in *cloud* computing. Even though the cloud service providers have standard regulations and powerful infrastructure to ensure

Customer's data privacy, data retrieval and provide a better outage have been apparent in last few years



Fig. 1. Cloud computing architecture example

In this work we observed that, from a customer's point of view, relying upon data retrieving which he needs by performing an effective query optimization In addition, providing reliability availability are crucial and equally important to query optimization. Query Optimization can be achieved by implementing the optimization techniques for effective retrieval of data.

To address optimization issues in this paper, we proposed the techniques for optimizing the queries to provide customers with fast data retreival [2]. In our model,

Query processing: A 3-step process that transforms a highlevel query (of relationa lcalculus/SQL) into an equivalent and more efficient lower-level query (of relational algebra).

#### 1. Parsing and translation

- Check syntax and verify relations.

- Translate the query into an equivalent relational algebra expression.

#### 2. Optimization

- Generate an optimal evaluation plan (with lowest cost) for the query plan.

#### 3. Evaluation

- The query-execution engine takes an (optimal) evaluation plan, executes that plan and returns the answers to the query.



The success of storage is due, in part to the availability

- of declarative query languages that allow to easily express complex queries without knowing about the details of the physical data organization and of advanced query processing technology that transforms the high-level user/application - Select and join at Site 5 queries into efficient lower-level query execution strategies.

The query transformation should achieve both correctness and efficiency

- The main difficulty is to achieve the efficiency

- This is also one of the most important tasks of any distributed system

• Cloud storage system query processing: Transform a highlevel query (of relational calculus/SQL) on a stored database (i.e., a set of global relations) into an equivalent and efficient lower-level query (of relational algebra) on relation fragments.

• Cloud storage system query processing is more complex

- Fragmentation/replication of relations
- Additional communication costs

Parallel execution

**Example:** Transformation of an SQL-query into an RA-query. Relations: EMP(ENO, ENAME, TITLE), ASG(ENO, PNO, RESP, DUR)

Query: Find the names of employees who are managing a project?

- High level query

SELECT ENAME FROM EMP, ASG

WHERE EMP.ENO = ASG.ENO AND DUR > 37

- Two possible transformations of the query are:

Expression

\_ENAME(\_DUR>37^ EMP.ENO=ASG.ENO(EMP × ASG))

\_ Expression 2: \_ENAME(EMP ⋊ KENO (\_DUR>37(ASG)))

- Expression 2 avoids the expensive and large intermediate • Strategy 1: Cartesian product, and therefore typically is better.

We make the following assumptions about the fragmentation[8]

- Data is (horizontally) fragmented and distributed

Site1:  $ASG1 = \_ENO \le "E3"(ASG)$ 

Site2: ASG2 = ENO > "E3"(ASG)

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- Site3:  $EMP1 = \_ENO \leq "E3"(EMP)$
- Site4: EMP2 = ENO>"E3"(EMP)
- \_ Site5: Result
- Relations ASG and EMP are fragmented and distributed in the same way
- Relations ASG and EMP are locally clustered on attributes **RESP** and **ENO** respectively

Now consider the expression ENAME(EMP ⋊KENO ( DUR>37(ASG)))

- Strategy 1 (partially parallel execution)[2]:
- Produce ASG'1 and move to Site 3
- Produce ASG' 2 and move to Site 4
- Join ASG' 1 with EMP1 at Site 3 and move the result to Site 5
- Join ASG' 2 with EMP2 at Site 4 and move the result to Site 5
- Union the result in Site 5
- Strategy 2:
- Move ASG1 and ASG2 to Site 5
- Move EMP1 and EMP2 to Site 5
- For simplicity, the final projection is omitted.



Calculating the cost of the two strategies under the following assumptions:

- Tuples are uniformly distributed to the fragments; 20 tuples satisfy DUR>37

- size(EMP) = 400, size(ASG) = 1000

- 1: tuple access cost = 1 unit; tuple transfer cost = 10 units
  - ASG and EMP have a local index on DUR and ENO

- Produce ASG's: (10+10) \* tuple access cost 20 data - Transfer ASG's to the sites of EMPs: (10+10)
  - \* tuple transfer cost 200 - Produce EMP's: (10+10) \* tuple access cost \* 2 40
  - Transfer EMP's to result site: (10+10) \* tuple transfer cost

200

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| Ergeneering Assessme   |                                       |  |
|--|---------------------------------------|--|
| – Total cost   | 460                                   | <b>Types of Query Optimizers wrt Search Techniques</b> [2]<br>– Exhaustive search  |
| • Strategy 2:  |                                       | Cost-based<br>Optimal  |
| - Transfer EMP1, EMP2 to site 5: 400 * tuple transfer cost 4.000   |                                       | Combinatorial complexity in the number of relations<br>– Heuristics  |
| - Transfer ASG1, ASG2 to site 5: 1000 * tuple transfer cost 10,000   |                                       | Not optimal<br>Regroups common sub-expressions   |
| - Select tuples from ASG1 [ ASG2: 1000 * tuple access cost 1,000   |                                       | Performs selection, projection first<br>Replaces a join by a series of semijoins   |
| <ul><li>Join EMP and ASG': 400 * 2</li><li>Total cost</li></ul>  | · · · · · · · · · · · · · · · · · · · | Reorders operations to reduce intermediate relation size<br>Optimizes individual operations  |
| Query optimization is a crucial and difficult part of the overall  |                                       |  |
| <ul> <li>query processing[2]</li> <li>Objective of query optimization is to minimize the following cost function: I/O cost + CPU cost + communication cost</li> <li>Two different scenarios are considered:</li> </ul> |                                       | <ul> <li>Static</li> <li>Query is optimized prior to the execution</li> <li>As a consequence it is difficult to estimate the size of the intermediate results</li> </ul> |
| <ul> <li>Wide area networks</li> <li>Communication cost dominates</li> <li>low bandwidth</li> <li>low speed</li> </ul>   |                                       | Typically amortizes over many executions<br>– Dynamic<br>Optimization is done at run time<br>Provides exact information on the intermediate relation                     |
| high protocol overhead<br>Most algorithms ignore all other cost components   |                                       | sizes<br>Have to re-optimize for multiple executions   |
| <ul> <li>Local area networks</li> <li>Communication cost not that dominant</li> </ul>  |                                       | <ul> <li>Hybrid</li> <li>First, the query is compiled using a static algorithm</li> </ul>  |
| Total cost function should be considered<br><b>Ordering of the operators</b> of relational algebra is crucial for  |                                       | Then, if the error in estimate sizes greater than threshold,   |
| <ul><li>efficient query processing</li><li>Rule of thumb: move expensive operators at the end of query</li></ul>   |                                       | Statistics   |
| <ul><li>Cost of RA operations:</li></ul>   |                                       | <ul> <li>Relation/fragments</li> <li>Cardinality</li> <li>Size of a tuple</li> </ul>   |
| Operation  | Complexity                            | Fraction of tuples participating in a join with another  |
| Select, Project  | O(n)                                  | relation/fragment  |
| (withoutduplicate elimination)   |                                       | – Attribute  |
| Project (with duplicate elimination)   |                                       | Cardinality of domain<br>Actual number of distinct values  |
| Group ,Join,Semi-join<br>Division,Set Operators  | O(n log n)                            | Distribution of attribute values (e.g., histograms)<br>– Common assumptions<br>Independence between different attribute values   |

#### **Query Optimization Issues**[2]

Several issues have to be considered in query optimization • Types of query optimizers

– Wrt the search techniques (exhaustive search, heuristics)

O(n2)

- Wrt the time when the query is optimized (static, dynamic)
- Statistics
- Decision sites

Cartesian Product

- Network topology
- Use of semi joins

#### **Decision sites**

- Centralized

Single site determines the "best" schedule Simple

Independence between different attribute values

Knowledge about the entire distributed database is needed

Uniform distribution of attribute values within their domain

- Distributed

Cooperation among sites to determine the schedule

Only local information is needed

Cooperation comes with an overhead cost

– Hybrid

One site determines the global schedule Each site optimizes the local sub-queries

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#### Network topology

- Wide area networks (WAN) point-to-point

- Characteristics
- · Low bandwidth
- $\cdot$  Low speed
- · High protocol overhead

Communication cost dominate; all other cost factors are distributed, hybrid) ignored

Global schedule to minimize communication cost

Local schedules according to centralized query optimization

- Local area networks (LAN)

Communication cost not that dominant

Total cost function should be considered

Broadcasting can be exploited (joins)

Special algorithms exist for star networks

# Use of Semi joins[2]

Reduce the size of the join operands by first computing semijoins

Particularly relevant when the main cost is the Reference communication cost

Improves the processing of distributed join operations by reducing the size of data

exchange between sites

However, the number of messages as well as local processing time is increased



Ouery processing transforms a high level query (relational calculus) into an equivalent lower level query (relational algebra). The main difficulty is to achieve the efficiency in the transformation

• Query optimization aims to minimize the cost function: I/O cost + CPU cost + communication cost

• Query optimizers vary by search type (exhaustive search, heuristics) and by type of the

algorithm (dynamic, static, hybrid). Different statistics are collected to support the query

optimization process

• Query optimizers vary by decision sites (centralized,

• Query processing is done in the following sequence: query decomposition data localization global optimization local optimization.

# Conclusion

In this paper, we proposed a various issues related to the Query Optimization in cloud computing, which seeks to provide each customer with better data retrieval from cloud data storage.

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