## A LINEAR ALGEBRA APPROACH TO OLAP

Rogério Pontes
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Universidade do Minho

HASLab
HIGH-ASSURANCE

## DATA WAREHOUSE



## OLAP

Online analytical processing (OLAP) systems, perform multidimensional analysis of business data and provides the capability for complex calculations, trend analysis, and sophisticated data modeling.

## OLAP APPLICATIONS

- Multidimensional OLAP
- Most efficient.
- Require more storage.
- Requires additional investment.
- Relational OLAP
- Uses all the developments made on relational databases.
- Least efficient.
- One example is Hive.
- Hybrid OLAP
- Has the advantage of both solutions.
- It is more complex as it deals with MOLAP and ROLAP.


## THE PROBLEM

## Main Issue

All the solutions depend on Relational Algebra.

- Relational Algebra Lacks algebraic properties on the most common operations.
- Hard to frame OLAP operators in Codd's Relational Algebra.
- Relational Algebra does not provide qualitative and quantitative proofs for all the relational operator.


## Typed Linear Algebra

An Algebra capable of calculation the OLAP operation with equations that can provide a formal proof both on the quantitative side and the qualitative side.

Projection Function creates the relation between the attributes and their original position in the table.

| Model | Year | Sales |  |  | 0 | 1 | 2 | 3 | 4 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Chevy | 1990 | 5 | $\xrightarrow{\text { Convert }}$ | $t_{\text {Model }}$ |  |  |  |  |  |
| Chevy | 1990 | 87 |  |  |  |  |  |  |  |
| Ford | 1990 | 64 |  | Chevy | 1 | 1 | 0 | 0 | 0 |
| Ford | 1990 | 99 |  | Ford | 0 | 0 | 1 | 1 | 1 |
| Ford | 1991 | 8 |  |  |  |  |  |  |  |

Measure Matrices store the numeric values in a Diagonal Matrix.

| Model | Year | Sales |  | $\llbracket t \rrbracket_{\text {Sales }}$ | 0 | 1 | 2 | 3 | 4 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Chevy | 1990 | 5 |  | 0 | 5 | 0 | 0 | 0 | 0 |
| Chevy | 1990 | 87 | Convert | 1 | 0 | 87 | 0 | 0 | 0 |
| Ford | 1990 | 64 | $\xrightarrow{\text { Cl }}$ | 2 | 0 | 0 | 64 | 0 | 0 |
| Ford | 1990 | 99 |  | 3 | 0 | 0 | 0 | 99 | 0 |
| Ford | 1991 | 8 |  | 4 | 0 | 0 | 0 | 0 | 8 |

## CROSS TAB

## OLAP Operations in Linear Algebra

Projection functions $\left(t_{A}\right)$, Measure Matrices $\left(\llbracket t \rrbracket_{M}\right)$ and Matrix multiplications form the building blocks of the Typed Linear Algebra.

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## OBJECTIVE \& CHALLENGES

## Main goal

Implement and evaluate the performance of a Typed Linear Algebra solution in a real world scenario. Can linear algebra provide a more efficient solution?

## Challenges

- Matrix sparsity. In the worst case only $0.1 \%$ of the matrix has non-zero values.
- Attribute-Range Problem. An attribute must always be in the same row in every matrix.
- Conversion of complex SQL queries to LA equations and implementation on a distributed system.


## MATRIX FORMATS

From the many standard sparse matrix representations, two of them were selected as the most appropriate to our problem.

- Compressed Sparse Column
- Used to represent Projections Functions and Measure matrix.
- Standard formats uses 3 or 4 arrays to store the data.
- Due to the properties of linear algebra matrices we improved the format to use a single array.
- Coordinate Format
- Used to exchange values between the computing nodes.


## ATTRIBUTE RANGE

## Problem

Typed linear algebra works with the premiss that the relation between Attribute and Matrix line number is bijective. How can two concurrent machines, that are generating a matrix, assign the same line number to the same attribute?

## Solution

Using a 64base encoding, its possible to assign a unique id to every Attribute.

## Problem

Typed linear algebra initial conception only operated with queries on a Single Table. Real Word query benchmarks are far more complex.

## Solution

Typed linear algebra was extended to support the relational algebra projection, restriction and joins.

## ARCHITECTURE



## CLUSTER SETUP

The cluster is made of 5 Machines, one server is dedicated to manage the cluster, the other four are the computing nodes. Each machine is running Ubuntu 14.0464 bit on Intel Core i3-3240 @ 3.40 Ghz, 3K cache and 8GB of RAM.

## EXPERIMENTAL EVALUATION

## Objective

We seek to evaluate the job latency time and resource usage of each computing node (CPU, Memory, Disk and Network).

## Procedure

Adapt TPC-H queries, generate data with scale factors from 2 to 32 and calculate the average job latency of 30 runs.

## Baseline

Hive, an Hadoop application that converts SQL to MapReduce jobs is used as a baseline of comparison.

## QUERY1 (ADAPTED FROM TPC-H)

SELECT RETURNFLAG, LINESTATUS, sum(QUANTITY) FROM LINEITEM
WHERE SHIPDATE >= 1998-08-28
AND SHIPDATE <= 1998-12-01
GROUP BY RETURNFLAG, LINESTATUS

$$
\left.L_{\text {ReturnFlag }} \nabla L_{\text {LineStatus }} \cdot[L]_{\text {Shipdate }}^{>=1998-08-28} \cdot[L]_{\text {Shipdate }}^{<=1998-12-01} \cdot \llbracket L\right]_{\text {Quantity }} \cdot!^{\circ}
$$

## QUERY1 JOB LATENCY



## QUERY 3 (ADAPTED FROM TPC-H)

```
SELECT L_SHIPMODE,
        O_ORDERSTATUS,
        SUM(0_TOTALPRICE * L_QUANTITY)
FROM LINEITEM AS L, ORDERS AS O
WHERE L_ORDERKEY = O_ORDERKEY
GROUP BY L_SHIPMODE, O_ORDERSTATUS.
```

$\left(L_{\text {shipmode }} \cdot \llbracket L \rrbracket_{\text {quantity }} \cdot L_{\text {orderkey }}^{\circ}\right) \cdot\left(O_{\text {orderkey }} \cdot \llbracket O \rrbracket_{\text {totalprice }} \cdot O_{\text {orderstatus }}\right)$

## QUERY 3 JOB LATENCY



## CONCLUSION

- LA has an improved performance of $38 \%$ on queries that work on a single Table.
- It has a decrease of performance of $29 \%$ on queries that involve a join.
- A large subset of Relational Algebra can be converted to LA equations.
- Matrix kernel libraries can be improved to handle sparse matrices.


## FUTURE WORK

- With the extension of the algebra its possible to benchmark a larger set of TPC-H queries.
- Create an execution plan from the conversion of SQL to LA.
- Prove correctness of conversion from SQL to LA.
- Other areas worth exploring are:
- Improve matrix libraries.
- Extend other types of joins (OuterJoins, AntiJoin)


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