

Preparing Data Sets For Data Mining Using CASE, PIVOT And SPJ

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Abstract--Data mining plays an important role in real time applications for extracting business intelligence from business data and make expert decisions. Datasets are used in order to mine data for the purpose of discovering knowledge from data. However, preparing datasets manually is a tedious task. The reason behind it is that it involves aggregation of relations and other complex operations. Another important reason for the difficulty is the fact that SQL aggregations do not provide datasets. Instead they can give only single value results that are not suitable for data mining. Data in horizontal layout is required for data mining purposes.

I. INTRODUCTION

RDBMS is widely used database management system for storing valuable business data of organizations. Generally regular data is stored in RDBMS. For data mining purposes, this data has to be transformed and moved to data warehouses. The data in warehouses is suitable for data mining. However, preparing data sets for data mining is very important for the successful mining of data. Data mining can't be done directly with regular databases. For preparing datasets SQL aggregate functions can be used but they produce single row output. Therefore it is not suitable for data mining purposes. What is expected for data mining purposes is the preparation of data with horizontal layout. The data in horizontal layout is suitable for data mining. Vertical aggregations such as SUM(), MAX(), MIN(), AVG() and COUNT() can't produce horizontal aggregations. Therefore they can't be used directly [1]. In case of statistical algorithms vertical aggregations are useful [2], [3]. Many data mining techniques like classification, clustering, regression, and so on need data in horizontal layout for data mining [4].

To overcome this problem in this paper we built new constructs like CASE, SPJ and PIVOT. These constructs are used to produce horizontal aggregations that give rise to data in horizontal layout. These constructs internally use SQL commands along with some business logic as required. Thus the generated datasets can be used for data mining in case of OLAP applications. We also

For this reason, in this paper we focus on the horizontal aggregations that can produce datasets. Towards it we build three constructs that can be used along with SQL queries to produce datasets automatically. The novel aggregations include SPJ, CASE and PIVOT constructs. We built a prototype for making experiments and the results revealed that the proposed aggregations are able to produce datasets required.

Index Terms– SQL, aggregations, horizontal aggregations

built a prototype application that can demonstrate the usefulness of the constructs developed in this paper. They are able to produce the datasets that can be used in data mining. The underlying logic is built in stored procedures which are pre-compiled objects that work faster than normal queries. The proposed application has web based and user-friendly interface. The remainder of this paper is organized as follows. Section II reviews literature. Section III describes horizontal aggregations. Section IV presents experimental results while section V concludes the paper.

II. RELATED WORK

SQL has been around for many decades which are the de facto standard to interact with relational databases. In all kinds of applications of all platforms this is commonly used language. The commands of SQL are generally categorized into DCL, TCL, DML and DDL. The aggregation functions supported by SQL are widely used to generate summary of data. In the process they can also be used for producing outputs but they are not suitable for data mining purposes. The reason behind is that they produce single row output. The format expected by data mining applications is horizontal layout [5]. Data mining techniques like association rule mining [6] are used to mine data. However, they operate on datasets to produce patterns from the data [7]. In this paper we focus on producing new aggregate functions that make use of SQL aggregations internally to produce datasets that can be used in data mining. The new aggregations produced by us are CASE, SPJ, and

PIVOT. In [5] SQL queries which are used in clustering algorithms. In [8] spreadsheet like operations are proposed through SQL. They also provided optimizations for joins and other SQL commands. New class of aggregations such as PIVOT, SPJ, and CASE are produced in this paper which is based on the traditional relational algebra [9], [10]. Tree based plans are traditionally used in optimizing queries [11]. Lot of research went on aggregations. The literature has cross and cube tabulations as well [12]. Unpivoted relational tables are presented in [13]. Transformations can be made from normal aggregations to horizontal aggregations [14]. The operations like TRANSPOSE and Unpivot are similar. Number of operations is less in case of transpose when compared it with PIVOT. Inverse relationship exists between them. Using them vertical aggregations can be produced that are widely used in decision trees in case of data mining domain. Relational databases support both of the operations [15]. In [16] and [17] also horizontal aggregations are worked out but they have limitations. One of the limitations is that they produce data that can't be directly used for data mining purposes. In this paper we produced new operators such as CASE, SPJ and PIVOT.

III. HORIZONTAL AGGREGATIONS

These are the operations that make use of aggregate functions to generate data in horizontal layout. However, it can't be done using normal SQL aggregations. Towards it in this paper we built new constructs namely CASE, SPJ and PIVOT that can be used with SQL commands in order to produce output with horizontal layout.

K	D ₁	D ₂	A
1	3	X	9
2	2	Y	6
3	1	Y	10
4	1	Y	0
5	2	X	1
6	1	X	null
7	3	X	8
8	2	X	7

D ₁	D ₂	A
1	X	null
1	Y	10
2	X	8
2	Y	6
3	X	17

D ₁	D ₂ X	D ₂ Y
1	null	10
2	8	6
3	17	null

Fig. 1 – Input table (a), traditional vertical aggregation (b), and horizontal aggregation (c)
 As seen in fig. 1, sample data is given in input table. Vertical aggregation result is presented in (b). In fact the result generated by SUM function of SQL is presented in (b). Horizontal aggregation results are presented in (c).

Steps Used in All Methods

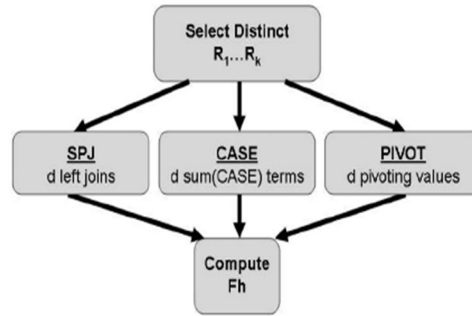


Fig. 2 shows steps on all methods based on input table

As seen in fig. 2, for all aggregations such as PIVOT, CASE and SPJ certain steps are carried out. However, the first step of all operations starts with SELECT query. Then based on the operation other activities are performed for computing horizontal aggregations.

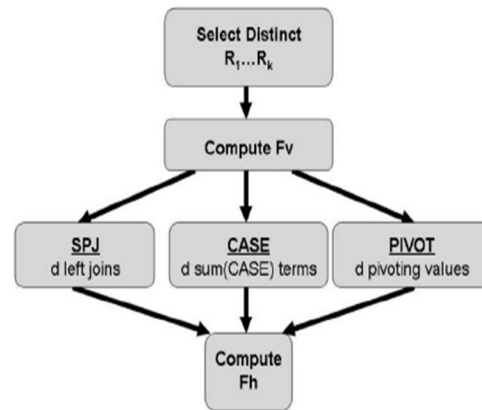


Fig. 3 shows steps on all methods based on table containing results of vertical aggregations

As seen in fig. 2, for all aggregations such as PIVOT, CASE and SPJ certain steps are carried out. However, the first step of all operations starts with SELECT query. Then based on the operation other activities are performed for computing horizontal aggregations.

SPJ Method

Vertical operations are used in SPJ method. For every column one table is generated in this model. Afterwards, the tables generated are joined in order to obtain final horizontal aggregations. The procedure followed is as given in [18].

PIVOT Method

provided by SQL such as SUM, MAX, MIN, AVG, and COUNT are unable to provide results in horizontal layout, in this paper, we built horizontal aggregations such as PIVOT, CASE and SPJ that can produce desired results that can be used for data mining operations. Our new aggregations are the programming constructs that make use of SQL aggregations internally along with certain business logic. Moreover our constructs are pre-compiled objects that work faster than normal aggregations. We built a prototype application that demonstrates the proof of concept. The empirical results revealed that the proposed aggregations are very useful in producing datasets with horizontal layout that can be used in data mining operations.

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