INTRODUCTION

Within today’s competitive economic context, information acquisition, analysis and exploitation became strategic and unavoidable requirements for every enterprise. Moreover, in order to guarantee their persistence and growth, enterprises are forced, henceforth, to capitalize expertise in this domain.

Data warehouses (DW) emerged as a potential solution answering the needs of storage and analysis of large data volumes. In fact, a DW is a database system specialized in the storage of data used for decisional ends. This type of systems was proposed to overcome the incapacities of OLTP (On-Line Transaction Processing) systems in offering analysis functionalities. It offers integrated, consolidated and temporal data to perform decisional analyses. However, the different objectives and functionalities between OLTP and DW systems created a need for a development method appropriate for DW.

Indeed, data warehouses still deploy considerable efforts and interests of a large community of both software editors of decision support systems (DSS) and researchers (Kimball, 1996; Inmon, 2002). Current software tools for DW focus on meeting end-user needs. OLAP (On-Line Analytical Processing) tools are dedicated to multidimensional analyses and graphical visualization of results (e.g., Oracle Discoverer®); some products permit the description of DW and Data Mart (DM) schemes (e.g., Oracle Warehouse Builder®). One major limit of these tools is that the schemes must be built beforehand and, in most cases, manually. However, such a task can be tedious, error-prone and time-consuming, especially with heterogeneous data sources.

On the other hand, the majority of research efforts focuses on particular aspects in DW development, cf., multidimensional modeling, physical design (materialized views (Moody & Kortnik, 2000), index selection (Golfarelli, Rizzi, & Saltarelli, 2002), schema partitioning (Bellatreche & Boukhalfa, 2005)) and more recently applying data mining for a better data interpretation (Mikolaj, 2006; Zubcoff, Pardillo & Trujillo, 2007). While these practical issues determine the performance of a DW, other just as important, conceptual issues (e.g., requirements specification and DW schema design) still require further investigations. In fact, few propositions were put forward to assist in and/or to automate the design process of DW, cf., (Bonifati, Cattaneo, Ceri, Fuggetta & Paraboschi, 2001; Hahn, Sapia & Blaschka, 2000; Phipps & Davis, 2002; Peralta, Marotta & Ruggia, 2003).

This chapter has a twofold objective. First, it proposes an intuitive, tabular format to assist decision maker in formulating their OLAP requirements. It proposes an automatic approach for the conceptual design of DW/DM schemes, starting from specified OLAP requirements. Its automatic approach is composed of four steps: Acquisition of OLAP requirements, Generation of star/constellation schemes, DW schema generation, and Mapping the DM/DW onto data sources. In addition, it relies on an algorithm that transforms tabular OLAP requirements into DM modelled either as a star or a constellation schema. Furthermore, it applies a set of mapping rules between the data sources and the DM schemes. Finally, it uses a set of unification rules.
that merge the generated DM schemes and construct the DW schema.

BACKGROUND

There are several proposals to automate certain tasks of the DW design process (Hahn, Sapia & Blaschka, 2000). In (Peralta, Marotta & Ruggia, 2003), the authors propose a rule-based mechanism to automate the construction of the DW logical schema. This mechanism accepts the DW conceptual schema and the source databases. That is, it supposes that the DW conceptual schema already exists. In addition, being a bottom-up approach, this mechanism lacks a conceptual design methodology that takes into account the user requirements which are crucial in the DW design.

In (Golfarelli, Maio & Rizzi, 1998), the authors propose how to derive a DW conceptual schema from Entity-Relationship (E/R) schemes. The conceptual schema is represented by a Dimensional-Fact Model (DFM). In addition, the translation process is left to the designer, with only interesting strategies and cost models presented. Other proposals, similar to (Marotta & Ruggia 2002; Hahn, Sapia & Blaschka, 2000) also generate star schemes and suppose that the data sources are E/R schemes. Although the design steps are based on the operational data sources, the end-users’ requirements are neglected. Furthermore, the majority of these works use a graphical model for the Data Source (DS) from which they generate the DM schema; that is, they neither describe clearly how to obtain the conceptual graphical models from the DS, nor how to generate the multidimensional schemes.

Other works relevant to automated DW design mainly focus on the conceptual design, e.g., (Hüsemann, Lechtenbörger & Vossen 2000) and (Phipps & Davis 2002) who generate the conceptual schema from an E/R model. However, these works do not focus on a conceptual design methodology based on users’ requirements and are, in addition, limited to the E/R DS model.

MAIN FOCUS

We propose an automatic approach to design DW/DM schemes from precisely specified OLAP requirements. This approach (Figure 1) is composed of four steps: i) Acquisition of OLAP requirements specified as two/n-dimensional fact sheets producing “semi-structured” OLAP requirements, ii) Generation of star/constellation schemes by merging the semi-structured OLAP requirements, iii) DW generation schema by fusion of DM schemes, and iv) Mapping the DM/DW to the data sources.

OLAP Requirement Acquisition

Decisional requirements can be formulated in various manners, but most generally they are expressed in natural language sentences. In our approach, which aims at a computer aided design, we propose to collect these requirements in a format familiar to the decision

Figure 1. DW design starting from OLAP requirements

![Diagram of DW design process](image-url)
Related Content

**Statistical Web Object Extraction**
[www.irma-international.org/chapter/statistical-web-object-extraction/11071/](www.irma-international.org/chapter/statistical-web-object-extraction/11071/)

**Scientific Web Intelligence**
[www.irma-international.org/chapter/scientific-web-intelligence/11049/](www.irma-international.org/chapter/scientific-web-intelligence/11049/)

**Statistical Data Editing**
[www.irma-international.org/chapter/statistical-data-editing/11068/](www.irma-international.org/chapter/statistical-data-editing/11068/)

**On Modeling and Analysis of Multidimensional Geographic Databases**
Sandro Bimonte (2010). *Data Warehousing Design and Advanced Engineering Applications: Methods for Complex Construction* (pp. 96-112).
[www.irma-international.org/chapter/modeling-analysis-multidimensional-geographic-databases/36610/](www.irma-international.org/chapter/modeling-analysis-multidimensional-geographic-databases/36610/)

**Model Indentification through Data Mining**
[www.irma-international.org/chapter/model-indentification-through-data-mining/10710/](www.irma-international.org/chapter/model-indentification-through-data-mining/10710/)