In previous work, we have shown how to use unified modeling language (UML) as the primary representation mechanism to model conceptual design, logical design, modeling of extraction, transformation, loading (ETL) processes, and defining online analytical processing (OLAP) requirements of data warehouses (DW). Continuing our work on using UML throughout the DW development lifecycle, in this chapter, we present our modeling techniques of physical design of DW using component diagrams and deployment diagrams of UML. Our approach allows the DW designer to anticipate important physical design decisions that may reduce the overall development time of a DW such as replicating dimension tables, vertical and horizontal partitioning of a fact table, and the use of particular servers for certain ETL processes. We illustrate our techniques with a case study.
Introduction

In the early 90s, Bill Inmon (Inmon, 2002) coined the term *data warehouse* (DW): “A data warehouse is a subject-oriented, integrated, time-variant, non-volatile collection of data in support of management’s decisions” (p. 33). This definition contains four key elements that deserve a detailed explanation:

- **Subject orientation** means that the development of the DW will be done in order to satisfy the analytical requirements of managers that will query the DW. The topics of analysis differ and depend on the kind of business activities; for example, it can be product sales focusing on client interests in some sales company, the client behavior in utilization of different banking services, the insurance history of the clients, the railroad system utilization or changes in structure, and so forth.

- **Integration** relates to the problem that data from different operational and external systems have to be joined. In this process, some problems have to be resolved: differences in data format, data codification, synonyms (fields with different names but the same data), homonyms (fields with the same name but different meaning), multiplicity of data occurrences, nulls presence, default values selection, and so forth.

- **Non-volatility** implies data durability: Data can neither be modified nor removed.

- **Time-variation** indicates the possibility to count on different values of the same object according to its changes in time. For example, in a banking DW, the average balances of client’s account during different months for the period of several years.

DWs provide organizations with historical information to support a decision. It is widely accepted that these systems are based on multidimensional (MD) modeling. Thus, research on the design of a DW has been mainly addressed from the conceptual and logical point of view through multidimensional (MD) data models (Blaschka, Sapia, Höfling, & Dinter, 1998, Abelló, Samos, & Saltor, 2001). During the few last years, few efforts have been dedicated to the modeling of the physical design (e.g., the physical structures that will host data together with their corresponding implementations) of a DW from the early stages of a DW project.

Nevertheless, the physical design of a DW is vitally important and highly influences the overall performance of the DW (Nicola & Rizvi, 2003) and the following maintenance; even more, a well-structured physical design policy can provide the perfect roadmap for implementing the whole warehouse architecture (Triantafillakis, Kanellis, & Martakos, 2004).
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