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Query-Driven Method for Improvement of Data Warehouse Conceptual Model

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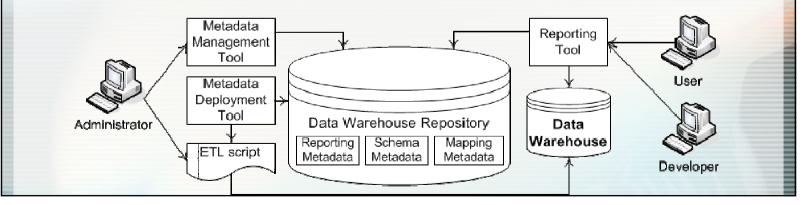
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Multidimensional Model of Data Warehouses

- Data warehouses (DW) are based on multidimensional models, which contain:
 - facts (the goal of the analysis),
 - measures (quantitative data),
 - dimensions (qualifying data),
 - dimension attributes that form classification hierarchies.
- Approaches to the development of conceptual models of DW :
 - supply driven (also known as data-driven)
 - demand driven
 - named according to the requirements elicitation method,
 - user-driven, process-driven, goal-driven etc.

Data Warehouse Evolution

- Data warehouse conceptual models tend to evolve, because of changes in :
 - data sources of DW
 - information requirements of users
- In our approach:
 - a data warehouse is a part of the data warehouse evolution framework
 - we use metadata about existing data warehouse schema, mappings of its elements to data source elements, as well as existing reports defined on the data warehouse schema



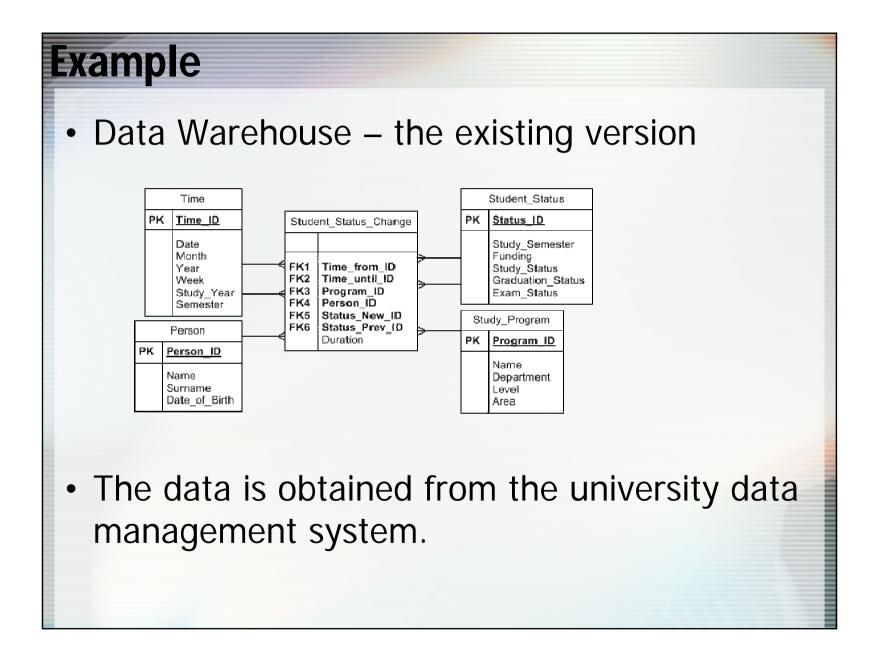
Research problem and our proposal

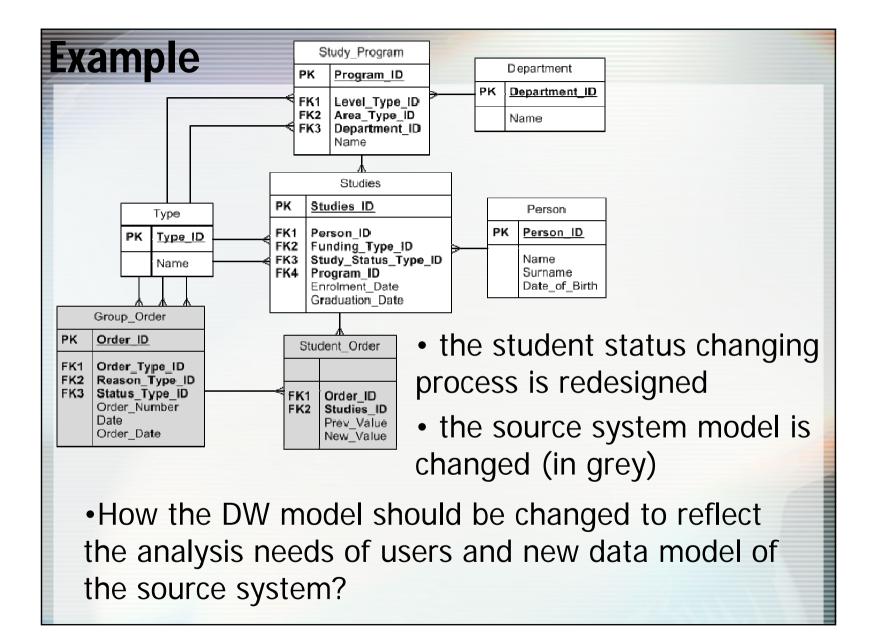
The problem addressed in our research is:

- how to reconcile new information requirements with the existing conceptual model of the data warehouse
- We propose a new type of demand-driven method:
 - query-driven method
 - elicits the requirements from existing queries on data sources and their usage statistics,
 - combines the needs of users with existing data supply
 - method recommends changes to existing data warehouse schemata
 - new version of DW can be constructed according to the recommendations

Related work

- Our method for eliciting user information requirements presumes that the queries against the source database reflect the interests and needs of users
- Some other works also exploit similar idea:
 - to improve query performance, but not to get a conceptual model,
 - analyze expected queries (documented during interviews) instead of real queries
 - analyze only the structure of implemented queries in source systems, regardless of the real usage of the queries
- The proposed method:
 - We take into account not only structure but also the usage statistics of queries in source systems
 - Our method is tailored to a definite purpose to recommend the necessary changes in existing data warehouse schema





Information Sources for Query Usage Study (1)

- We will examine a source system that collects different logs about functions and data usage
- The function usage:
 - Logging is invoked when the function is performed
 - Log format: *Function_name, user, date.*
 - This log can exhibit the most frequently used functions
 - the data manipulation statements that are used within these functions should be analyzed
- The data changes:
 - Each table has its own trigger for logging insert, update and delete statements
 - Log format: Table_name, user, date, change_type, change_text
 - The statistics about changes for each table and attribute is computed

Information Sources for Query Usage Study (2)

• A special kind of functions:

- reports with a wide range of parameters, that user can change
- The queries in reports represent also the information requirements of users
- format of such log is the following:

Function_name, user, date, parameter_text,

- where parameter_text contains values in the format: p1=<value1>... pN=<valueN>,
- where *p1...pN* are in the format: <*table_name.column_name>*.

Information Sources for Query Usage Study (3)

- The structure of data manipulation statements used in functions of source systems is examined
- Two kinds of information sources can be prepared.
 - INSERT and UPDATE statements :
 - Function_name, Action_Type, List_of_columns, Where_condition,

– SELECT statements :

• Function_name, Joins, Where_condition, Group_By, Columns_Select, Aggregation,

Query-Driven Method (1.step)

- Pre-processing of Information Sources
 - two usage tables are constructed
 - data modification table contains data about the frequency of INSERT and UPDATE statements performed by functions with tables
 - Function_name, List_of_columns, Where_condition, count.
 - data selection table contains data about the frequency of SELECT statements performed by functions with tables
 - Function_name, Joins, Where_condition, Group_By, Columns_Select, Aggregation, count.

Query-Driven Method (2.step)

- Analysis of Data Modification Statements
 - data modification table is analysed.
 - For each column, we calculate the number of times it was updated or inserted
 - we summarize *count* for every occurrence of the column in *List_of_columns* of the data modification table
 - we choose the Top-N columns with the biggest number of times they were updated or inserted and consider such columns as **potential measures**.
 - we identify the potential dimension attributes from the Top-N columns with the biggest number of times they were used in *Where_condition* in the data modification table.
 - We assume that potential fact measures could be analysed together with potential dimension attributes if they are used together in one data modification statement.

Query-Driven Method (3.step)

- Analysis of Data Selection Statements
 - the Top-N most frequently used columns in *Where_condition* or *Group_By* are considered as **potential dimension attributes**.
 - The Top-N columns with the biggest number of times they were used in *Aggregation* are considered the potential measures.
 - Several potential measures belonging to one source table are united into a potential data warehouse fact table,
 - Several potential attributes belonging to one source table are united into a potential data warehouse dimension table

Query-Driven Method (4.step)

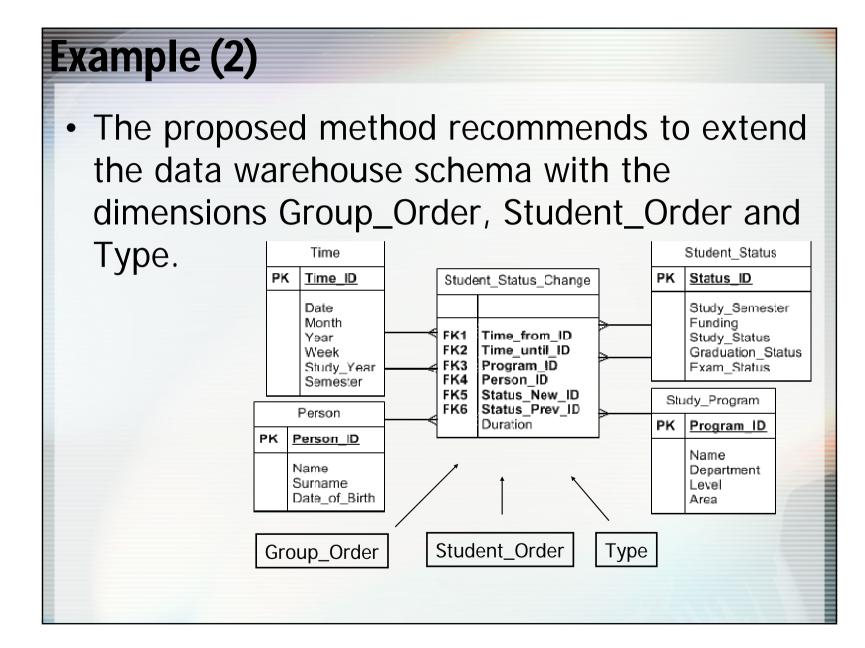
- Conflict Resolution
 - it is possible that the same source column can be identified as potential fact measure **and** as a potential dimension attribute
 - In such case we analyse the data type of the column:
 - If the data type of the column is **numeric**, then we consider such column as a **measure**,
 - If the data type of the column is character or date, then we refer to such column as to a dimension attribute

Query-Driven Method (5.step)

- Generation of Potential Improvements:
 - the potential data warehouse schema elements are compared with the existing data warehouse schema
 - Mappings between source table columns and existing data warehouse elements are used in this comparison.
 - Non-existing potential data warehouse schema elements are recommended to be added to the existing data warehouse schema.
 - Schema changes supported by the data warehouse evolution framework:
 - add a new dimension attribute or fact measure,
 - add a new dimension or fact table,
 - connect dimension to a fact table,
 - add a new dimension hierarchy or new hierarchy level.
 - Recommended changes should be accepted by the DW administrator

Example (1)

- The proposed method was applied to the student status change data warehouse and student management system
- Student status changing process was logged and analyzed:
 - The following columns are encountered frequently in *Where_Condition, Group_By* and *Columns_Select* of the data selection table:
 - Attribute Name of the table Type;
 - Order_Date and Order_Number of the table Group_Order;
 - Prev_Value and New_Value of the table Student_Order.
 - these columns are considered as potential dimension attributes



Example (3)

- After the **manual revision** of recommended improvements:
 - it was decided to unite dimensions Group_Order, Student_Order and Type in one dimension Order with the following attributes:
 - Order_Date, Order_Number, Prev_Value, New_Value.
 - the column Name of the table Type was added as 3 attributes of the dimension Order:
 - Order_Type, Reason_Type and Status_Type.

Conclusions and Future Work

- The proposed method:
 - allows to keep track of actual analysis tasks performed in operational data sources
 - adjust the existing data warehouse schema according to the most frequent queries.
 - the method is used to recommend necessary changes for a new data warehouse schema version.
 - a data warehouse schema also serves as a representation of the existing data warehouse analysis needs.
- Future work:
 - We are working on implementation of the method in more complicated cases and evaluation of constructed models to be convinced that the new version of the data warehouse is adjusted to real analysis needs

