

교과목:데이타베이스시스템

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Decision Support and OLAP

- Information technology to help knowledge worker(executive,manager,analyst) make faster and better decisions
 - >> What were the sales volumes by region and product category for the last year?
 - >> Which orders should we fill to maximize revenues?
 - >> Will a 10% discount increase sales volume sufficiently?
- On-line analytical processing (OLAP) is an element of decision support system (DSS).

Evolution

- 60's: Batch reports
 - >> hard to find and analyze information
 - >> inflexible & expensive, reprogram every new report
- 70's: Terminal-based DSS and EIS(executive information system)
 - >> still inflexible, not integrated with desktop tools
- 80's: Desktop data access and analysis tools
 - >> query tools, spreadsheets, GUIs
 - >> easier to use, but only access operational databases
- 90's: Data warehousing with integrated OLAP engines and tools

What is OLAP?

- A category of software technology that enables analysts, managers and executives to gain insight into data through fast, consistent, interactive access to a wide variety of possible views of information that has been transformed from raw data to reflect the real dimensionality of the enterprise as understood by the user
 - from The OLAP Council's Definition
- Goal of OLAP is to support ad-hoc querying for the business analyst
- Multidimensional view of data is the foundation of OLAP

OLTP vs. OLAP

		OLTP	OLAP
•	User	Clerk, IT professional	Knowledge worker
•	Function	Day to day operations	Decision support
•	DB design	Application-oriented	Subject-oriented
•		(ER-based)	(Star, Snowflake)
•	Data	Current, Isolated	Historical, Consolidated
•	View	Detailed, Flat relational	Summarized, Multidimensional
•	Usage	Structured, Repetitive	Ad hoc
•	Unit of work	Short, simple transaction	Complex query
•	Access	Read/Write	Read mostly
•	Operations	Index/hash on prim.key	Lots of scans
•	# Records accessed	Tens	Millions
•	#Users	Thousands	Hundreds
•	Db size	100MB-GB	100GB-TB
•	Metric	Trans.throughput	Query.throughput, response

DSS(Decision Support System)

• A system used <u>to support managerial decisions.</u> Usually DSS involves the analysis of many units of data in a heuristic fashion. As a rule, DSS processing does not involve the update of data



Modification, Re-Request

< Legacy DSS cycle >

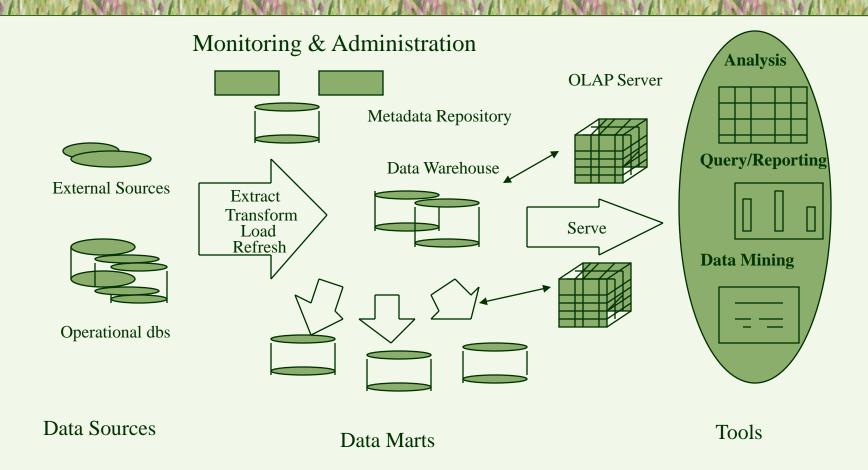
Data Warehouse

- A <u>decision support database</u> that is maintained separately from the organization's operational databases
- A data warehouse is a
 - subject-oriented
 - integrated
 - time-varying
 - non-volatile

collection of data that is used primarily in organizational decision making

- W.H.Inmon, Building the Data Warehouse,1992

Data Warehousing Architecture



Three-Tier Architecture

- Warehouse database server
 - >> almost always a relational DBMS; rarely flat files
- OLAP servers
 - >> Relational OLAP(ROLAP): extended relational DBMS that maps operations on multidimensional data to standard relational operations
 - >> Multidimensional OLAP(MOLAP) : special purpose server that directly implements multidimensional data and operations
- Clients
 - >> Query and reporting tools
 - >> Analysis tools
 - >> Data mining tools(e.g., trend analysis, prediction)

OLAP Server Requirements

- Data Types
 - >> text, calendar, numeric
- Calculation Engine
 - >> domain-specific computations(calendar)
 - >> rich library of aggregate functions
- Data at different granularities
 - estimated and actual sale for each product
- Data Load and Refresh
 - >> write must update precomputed aggregates
 - >> write/load must be accompanied by data cleaning

Data Mart & Data Mining

- **Enterprise warehouse**: collects all information about subjects (customers, products, personnel) that span the <u>entire organization</u>
 - >> Requires extensive business modeling
 - >> May take years to design and build
- **Data Marts**: <u>Departmental subsets</u> that focus on selected subjects: Marketing data mart: customers, products, sales.
 - >> Faster roll out, but complex integration in the long run
- **Data Mining**: New technology that <u>recognizes patterns</u> in the data to help you describe existing data and <u>predict</u> <u>future behaviors</u> based on current characteristic

Metadata Repository

• Administrative metadata

- source databases and their contents
- · warehouse schema, view&derived data definition
- dimensions, hierarchies
- · data extraction, cleansing, transformation rules, defaults
- · security: user authorization, access control
- etc...

Business data

- business terms and definitions
- ownership of data
- · charging policies

Operational metadata

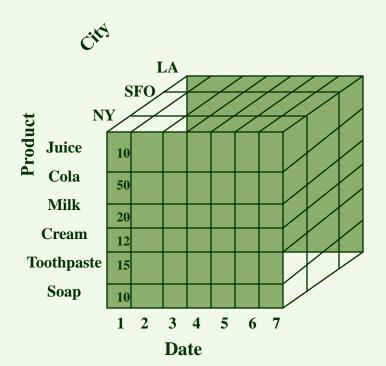
- data lineage: history of migrated data and sequence of transformations applied
- · currency of data :active, archived, purged
- monitoring information: warehouse usage statistics, error reports, audit trails

Multidimensional Data Model

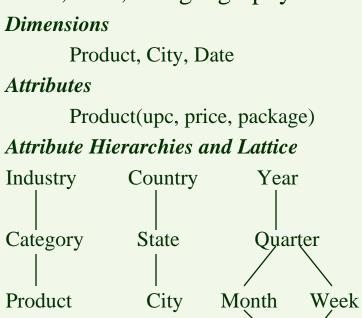
- Database is a set of *facts(points)* in a multidimensional space
- A fact has a **measure** dimension >> quantity that is analyzed, e.g., sales, budget
- A set of **dimensions** wrt which data is analyzed >> e.g., store, product, date, associated with a sales amount
- Dimensions form a sparsely populated coordinate system
- Each dimension has a set of *attributes*>> e.g., owner, city and country of store
- Attributes of a dimension may be related by partial order
 - >> **Hierarchy**: e.g., street ->> country ->> city
 - >> Lattice: e.g., date -> month -> year, date -> week->year

Multidimensional Data

Sales volume as a function of product, time, and geography



Fact data: Sales volume in \$100

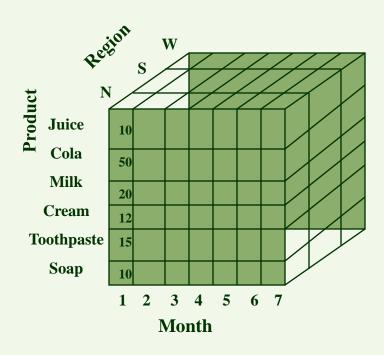


Date

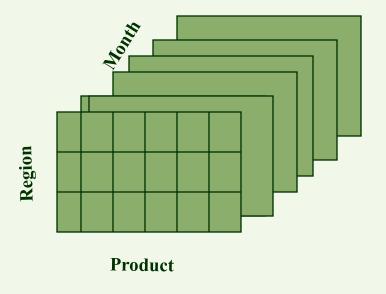
Operations in Multidimensional Data Model

- Aggregation (*roll-up*):
 - dimension reduction : e.g., total sales by city
 - summarization over aggregate hierarchy: e.g., total sales by city and year -> total sales by region and by year
- Selection (*slice*) defines a subcube
 - e.g., sales where city = Palo Alto and date = 1/15/96
- Navigation to detailed data (*drill-down*)
 - e.g., show supporting data for total sales figure for NW region
- Calculation and ranking
 - e.g., (sales-expense) bt city, top 3% of cities bt average income
- Visualization Operations(e.g., Pivot)

A Visualization Operation : Pivot (Rotate)



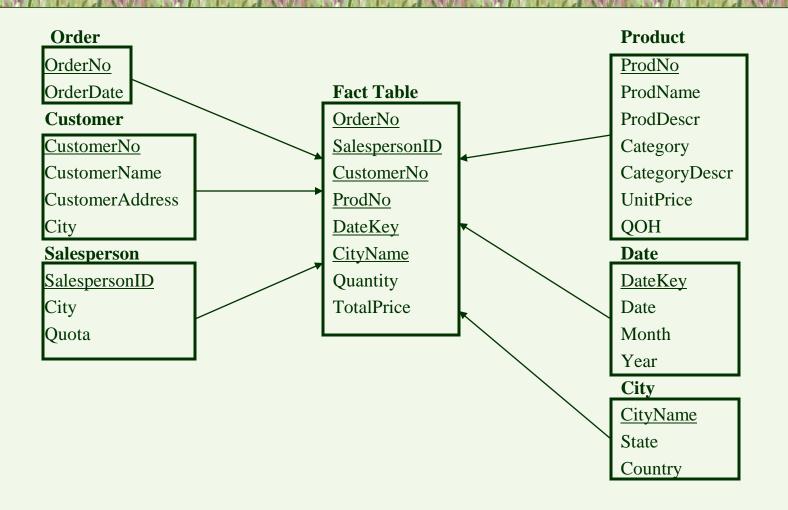
Fact data: Sales volume in \$100



Warehouse Database Schema

- ER design techniques not appropriate
- Design should reflect multidimensional view
 - >> Star Schema
 - >> Snowflake Schema
 - >> Fact Constellation Schema

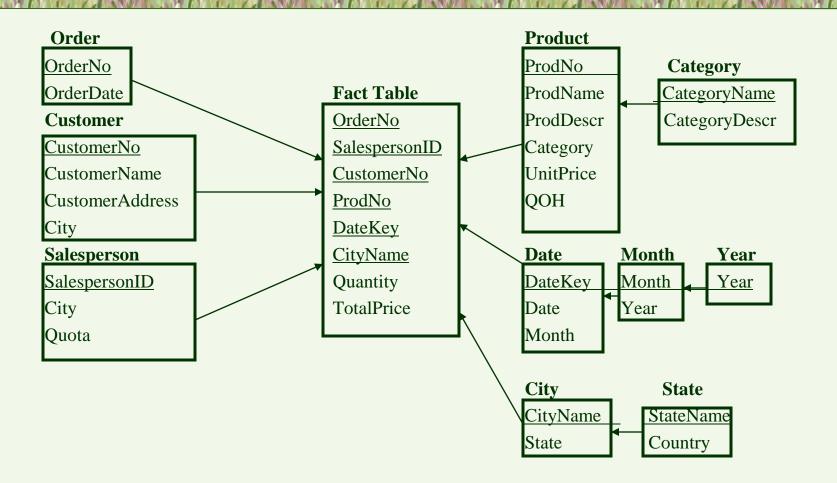
Example of Star Schema



Star Schema

- A single fact table and a single table for each dimension
- Every fact points to one tuple in each of the dimensions and has additional attributes
- Does not capture hierarchies directly
- Generated keys are used for performance and maintenance reasons
- **Fact Constellation**: Multiple Fact tables that share many dimension tables
 - >> Example : Projected expense and actual expense may share dimension tables

Example of a Snowflake Schema



Snowflake Schema

- Represent dimensional hierarchy directly by normalizing the dimension tables
- Easy to maintain
- Saves storage, but it is alleged that it reduces effectiveness of browsing (Kimball)

References

- W.H. Inmon: Building the Data Warehouse(2nd Edition). John Wiley, 1996
- R.Kimball: The Data Warehouse Toolkit, John Wiley, 1996
- E.F. Codd, S.B. Codd, C.T. Salley: Providing OLAP(On-line Analytical Processing) to User Analysts: An IT Mandate. Available from Arbor Software's web site
- Articles in trade journals: Datamation, Databased Advisor, Database Programming and Design, DBMS Magazine.
- Web sites of all product vendors.
- Web site of the OLAP Council.

Http://pwp.starnetic.com/larryg/articles.html has many references.

Http://www.olapcouncil.org/research/whitepapco.htm