



# Business intelligence

Unit 2 – Datawarehouse and OLAP

S2-1 – Datawarehouse



#### Introduction to datawarehouse



- Objective: to provide infrastructure for the DSS (Decision Support System) in an organization
  - We start from basic systems to organizational processes.
  - These systems may have several operational databases.
- DSS have new requirements
  - We want to extract knowledge from the databases and historical operational.
  - In order to:
    - Analysis of the organization
    - Make predictions
    - Define strategies



#### Introduction to datawarehouse



- You can still use the traditional system:
  - Maintaining daily transactional work in the original information systems (known as OLTP, On-Line Transactional Processing).
  - Basic data analysis is done in real time on the same database (known as OLAP, On-Line Analytical Processing).

#### but:

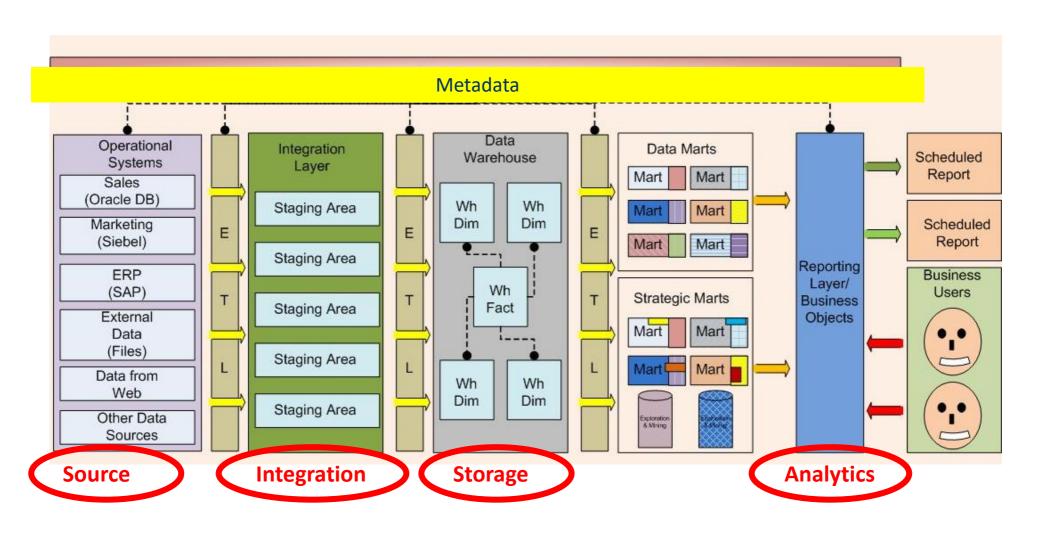
- Efficiency problems in daily work due to complex queries that are made when there is low load.
- Efficiency problems in the analysis because there is no any specific design. Not possible in real time.



# Conceptual Architecture and Components



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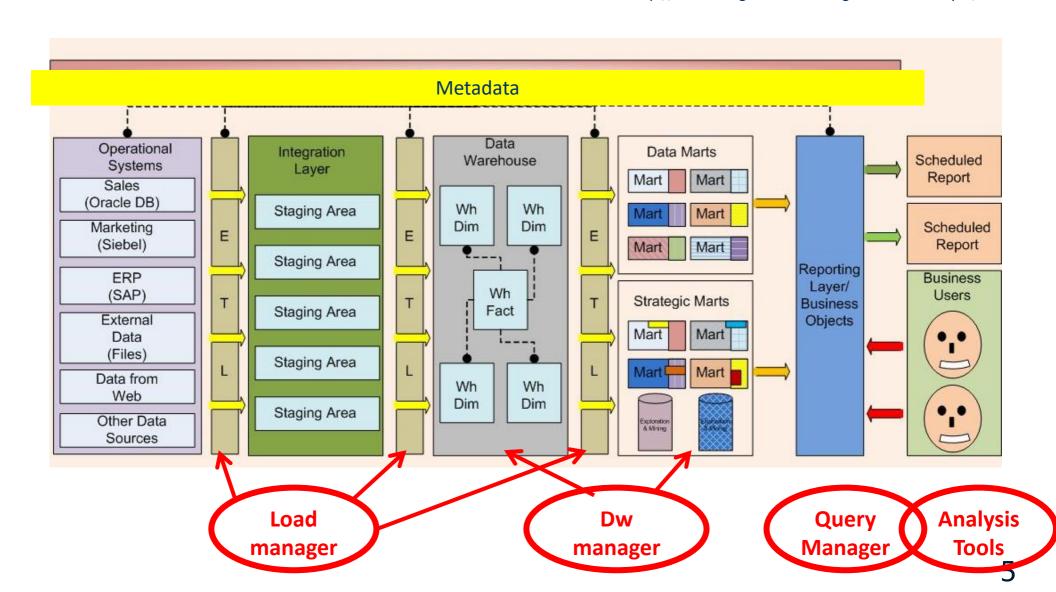




# Conceptual Architecture and Components



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## **DW Components**



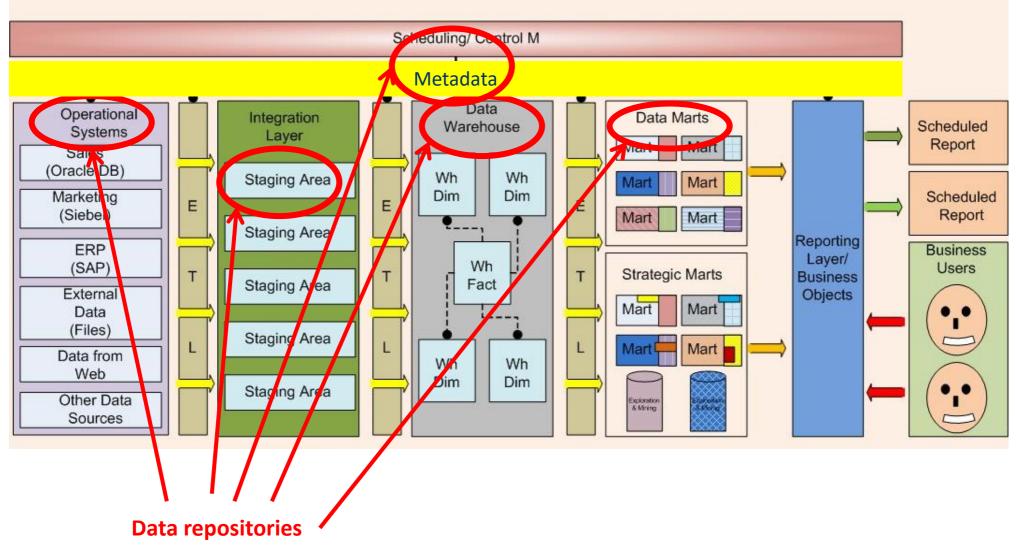
- Load manager: runs ETL tasks
  - Extraction
  - Transformation
  - Load
- DW manager (server): it allows to define and maintain the datawarehouse: data definition, aggregation, views, index, backup, etc..
- Query manager: Query execution, monitoring, ad-hoc forms, etc.
- Access tools: tools to design queries and reports, tools to develop end-user applications, OLAP tools, data mining tools, enterprise Information Systems (EIS)



# Conceptual Architecture and Components



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#### Data



- Data sources: files, www, xls, databases, ...
- Staging area: integrated conceptual model of information.
  - E-R, Relational model.
  - Not compulsory to have it implemented, but usually convenient.
- Datawarehouse: data collection for decision making
  - Multidimensional model
- Data mart: departamental dw
- Metadata: describes an organization in terms of its business activities, the business objects, and rules on which the business activities are performed.
  - Technical meta data needs to be mapped to the business meta data.
  - Includes documentation about data sources (origen, description, aggregation level, storage, ...)



#### Datawarehouse: Features of a DW



- The core of a BI architecture is the Datawarehouse
- Data Warehouse: A collection of data designed to support the decision-making processes:
  - Information Oriented (not processes)
  - integrated
  - Variable over time
  - Nonvolatile



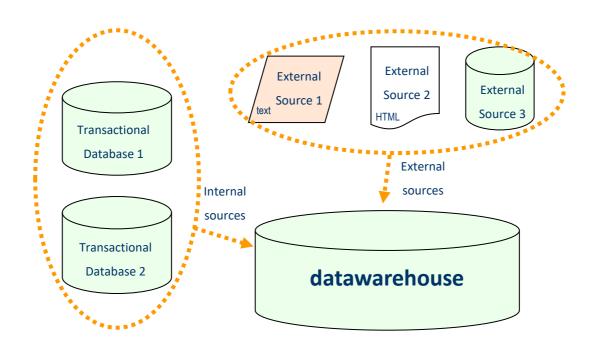


- Information oriented (not processes): dw is designed to efficiently view information on the basic activities (sales, purchasing, production, ...) of the organization, not to support the processes that take place in it (order management, billing, etc).
- Necessary information is extracted from transactional systems and efficiently stored for analysis.
- It leaves out irrelevant information.





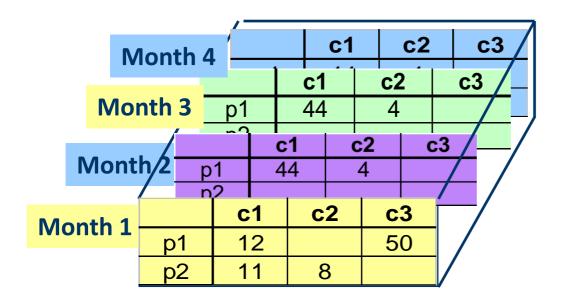
 Integrated: it collects data not only from the transactional databases, but it can also include external sources







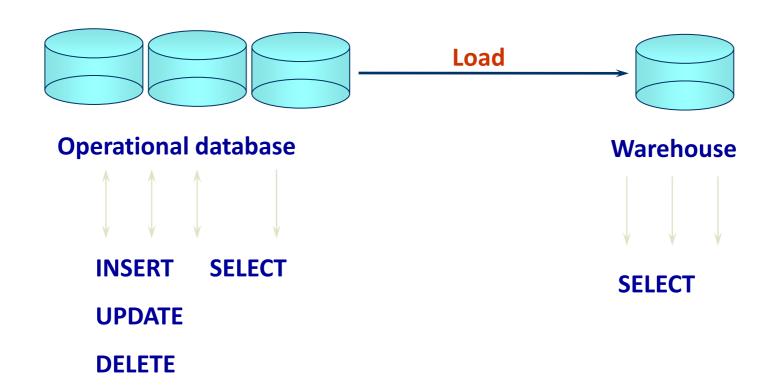
 Variable in time: the data are relative to a period of time and must be periodically increased







Non-volatile: the stored data are not updated, only increased



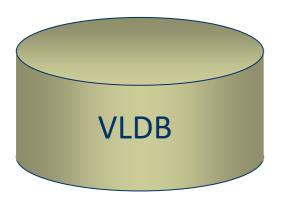


#### Introduction



 Advances in technology have encouraged the development of data warehouse technology

- Big data technology
  - Parallelism
  - Hardware
  - Distributed operating systems
  - Database
  - Query languages



- VLDB
- Big memory
- Indexing techniques
- Open systems (interoperability)
- Specialized hw and sw for DW
- Tools for data analysis



# OLTP vs DW



•	OLTP system	Datawarehouse	
•	Stores current data	historical data (trend->include current data)	)
•	Stores detailed data	Stores summarized or aggregated data	
•	Data are dynamic (updatable)	Data are static	
•	Repetitive processes	Ad-hoc unforeseeable processes	
•	Predictable usage pattern	Unpredictable usage pattern	
•	High rate of transaction	Medium or low rate of transactions	
•	Low response time (seconds)	Variable (usually long) response time	
•	Directed by transactions	Directed by data analysis	
•	Application or process-oriented	Information oriented	
•	Supports daily decisions	Supports strategic decisions	
•	High number of users (administrative)	Few users (managers)	
•	Medium-size databases	Large-size databases	15



## Advantages of using DW



- The advantages of using DW are, among others:
  - High ROI
  - Competitive advantages:
    - Information non previously available, unknown or difficult to extract and incorporate
  - Higher productivity in decision making personnel:
    - More integrated information with easy Access
  - Better data quality and consistency



## Problems



#### Problems:

- Understatement of resources to load
- Complexity of integration
- Hidden problems of source systems
- High demand for resources
- High cost of ownership
- Required data are not captured
- Increased demand from end users
- Homogenization data
- Data ownership
- Long-term projects





- Multidimensional model:
  - it models an activity which is subjected to analysis (fact) and dimensions that characterize the activity.
  - relevant information about the event (activity) is represented by a set of indicators (measures or fact attributes).
  - descriptive information for each dimension is represented by a set of attributes (dimension attributes).

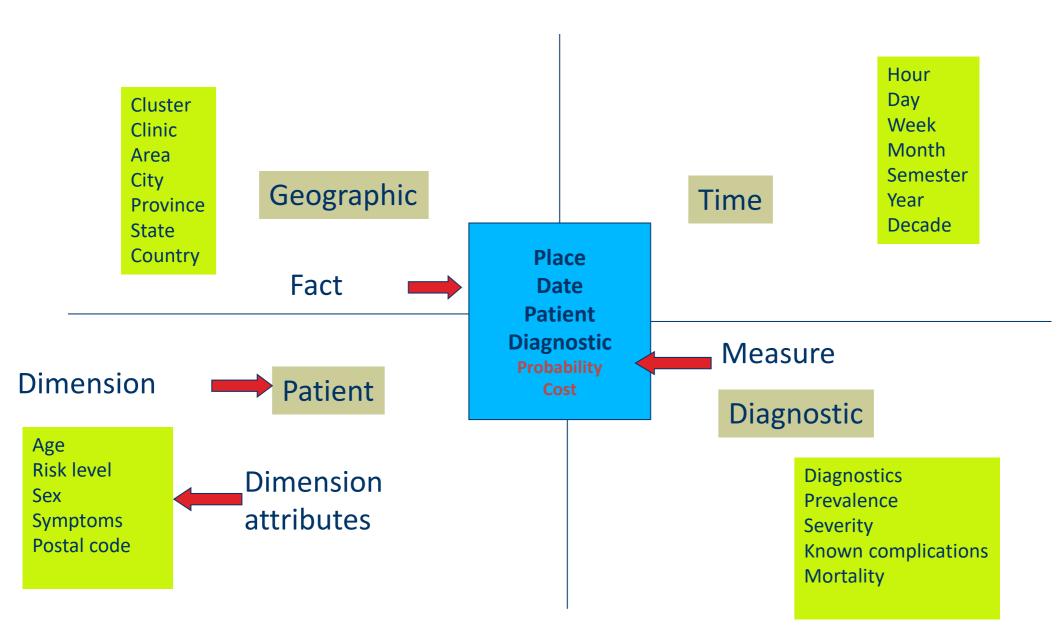




- Activity analized: Diagnostics.
- Information recorded about diagnostic: "Avian influenza diagnostics has been realized in the clinic "Morales" on Oct, 11th 2012 with a probability of 85% to the patient "Joseph".
- The geographic and temporal context are important, not only the concrete diagnostic to a person.



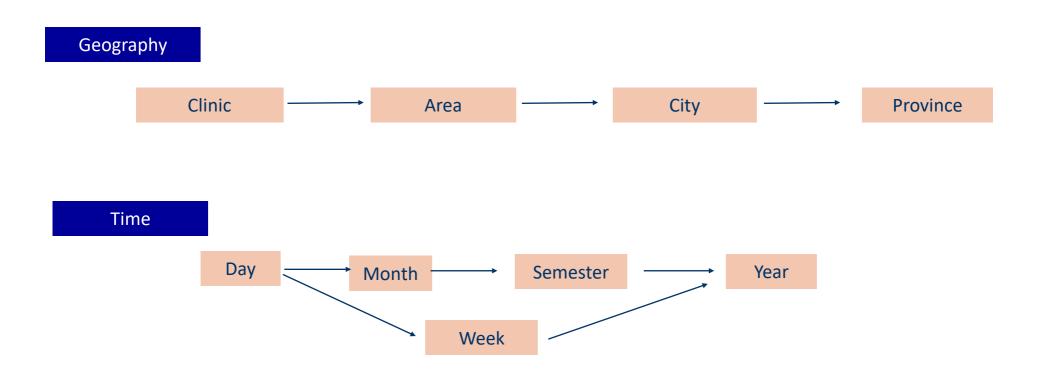








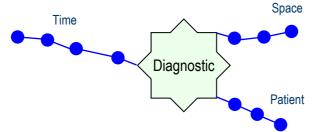
Hierarchy: sorting between the dimension's attributes



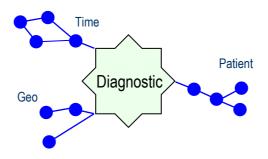




- Basic structures
- Star: lineal relation between dimension's attributes



Snowflake: non linear hierarchy

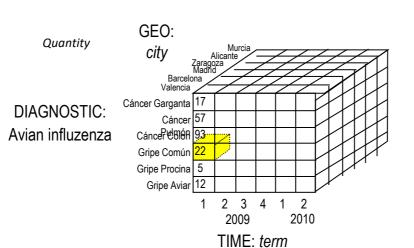


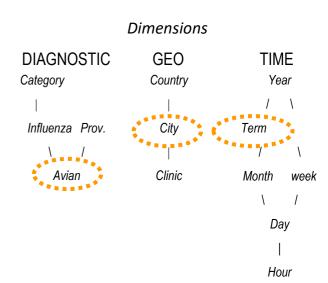




- Facts can be queried at different aggregation levels:
  - Query measures about the facts parametrized by dimensions' attributes and constrained by values of those attributes

FACT: "The first term of 2010 22 cases of avian influenza were diagnosed in Murcia



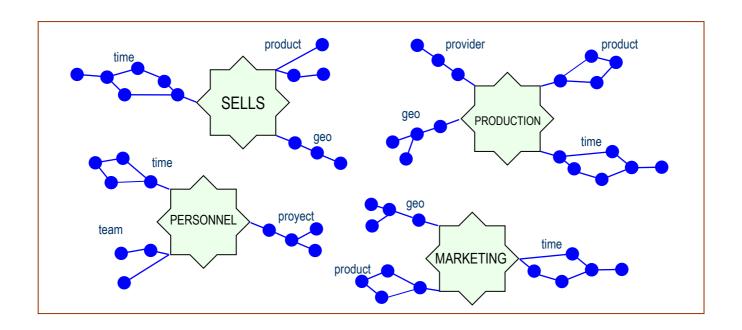


An aggregation level for a set of dimesions is called cube.





- Not all the information can be stored in a single schema
  - Some departamental datawarehouses are needed
  - Each of these is called datamart.





## Design strategies



#### Datamart:

- defined to meet the needs of a department or subdivision of the organization.
- contains less detail and more aggregated information.
- Up-down (Inmon)
  - First define the data warehouse of the whole organization and then define data marts on him
- Bottom-up (Kimball)
  - predefine departmental data marts and then integrate them into a data warehouse for the organization



#### **Datamart**



- Datamart: subset of the datawarehouse
- The datawarehouse can be formed by several datamarts and, optionally, additional tables.
- Are defined to meet the needs of a department or subdivision of the organization.
- Contains less detailed information and more aggregate information.
- They are easier to understand and use.
- They may be the intermediate step between the data warehouse and transactional system



#### **Datamart**



#### Why datamarts:

- Provide users with access to the most commonly analyzed
  - With better response time
  - Offloads the primary DWH
- The data will already be adapted for OLAP functions or DM.
- More control over their data mart by department users.
- Being simple, ETL processes are too.
- Low cost. Faster implementation.
- Greater involvement of users in the overall data warehouse.
  Better data quality, security, relevance,...



#### Data lakes



- In DW data are transformed and structured, and we keep only the data needed for the analysis
- Are we losing data?
- Data lakes store ALL source data in original format (may be not transformed)
- Easy to transform later for new DW or for data analysis
  - But schema-on-read. Is it good?
- Created for data scientist and explorers, not for business users