

# Business Intelligence 2

## Introduction to Traditional Data Warehouses

Slides adapted from Jiawei Han, Micheline Kamber, Jian Pei, (2011), Data Mining: Concepts and Techniques, Third Edition, The Morgan Kaufmann Series in Data Management System.

# The need for a Data Warehouse

- Business generated vast amounts of data.
- Organisations spread globally.
- Competition between business grow.
  - Operational systems capture, store and process large amount of data use daily but managers require access to consolidate data and information that can be used for decision making.

# Why are Data Warehouses Used?

- Allow managers to track the productivity of their organization.
- Managers can use the operational systems vast data resources to make decisions.
- Data analysis can produce information used in the decision-making process.

# Where are Data Warehouses Used?



# Examples of Data Warehouse Usage

- Improves competitive advantage and aid strategic decision making.
- i.e.
  - Reduce churn.
  - Increase market share .
  - Identify new products.
  - Improve customer satisfaction.
  - Increase productivity.
  - Increase global sales.

# What is a Data Warehouse?

- Many explanations.
  - A large quantity of data used to support decision making.
  - Central store of current and historical data used by managers.
- “A data warehouse is a subject-oriented, integrated, time-variant, and nonvolatile collection of data in support of management’s decision-making process.” —W. H. Inmon
- Data warehousing:
  - The process of designing, implementing and using a data warehouse

# Characteristics of a Data Warehouse

- Subject Oriented.
- Integrated.
- Time – Variant.
- Non-volatile
- Relational Structure or Multidimensional
- Client/Sever Architecture
- Metadata

# Subject-Oriented

- Organized around major business entities (subjects).
- Used for visualizing and analysis business data for decision making
- Only considers data relevant to the entities
- E.g. Sales:
  - Customers
  - Products



# Integrated

- Data sourced from multiple, heterogeneous data sources.
  - E.g. relational databases, flat files, on-line transaction records
- Data requires cleaning and integrating.
  - E.g. Consistency in naming conventions, data format, attribute measures, etc.
  - E.g., Customers: customers name, address, contact details, etc.
- ETL process used to convert the data.

# Time Variant

- Data Warehouses stores data over a period
  - Operational database: store current regularly used data
  - Data warehouse data: store current and historical data (e.g., past 10+ years)
- Data in a data warehouse contains a time element, explicitly or implicitly.
- Data in an operational system may not contain a “time element”

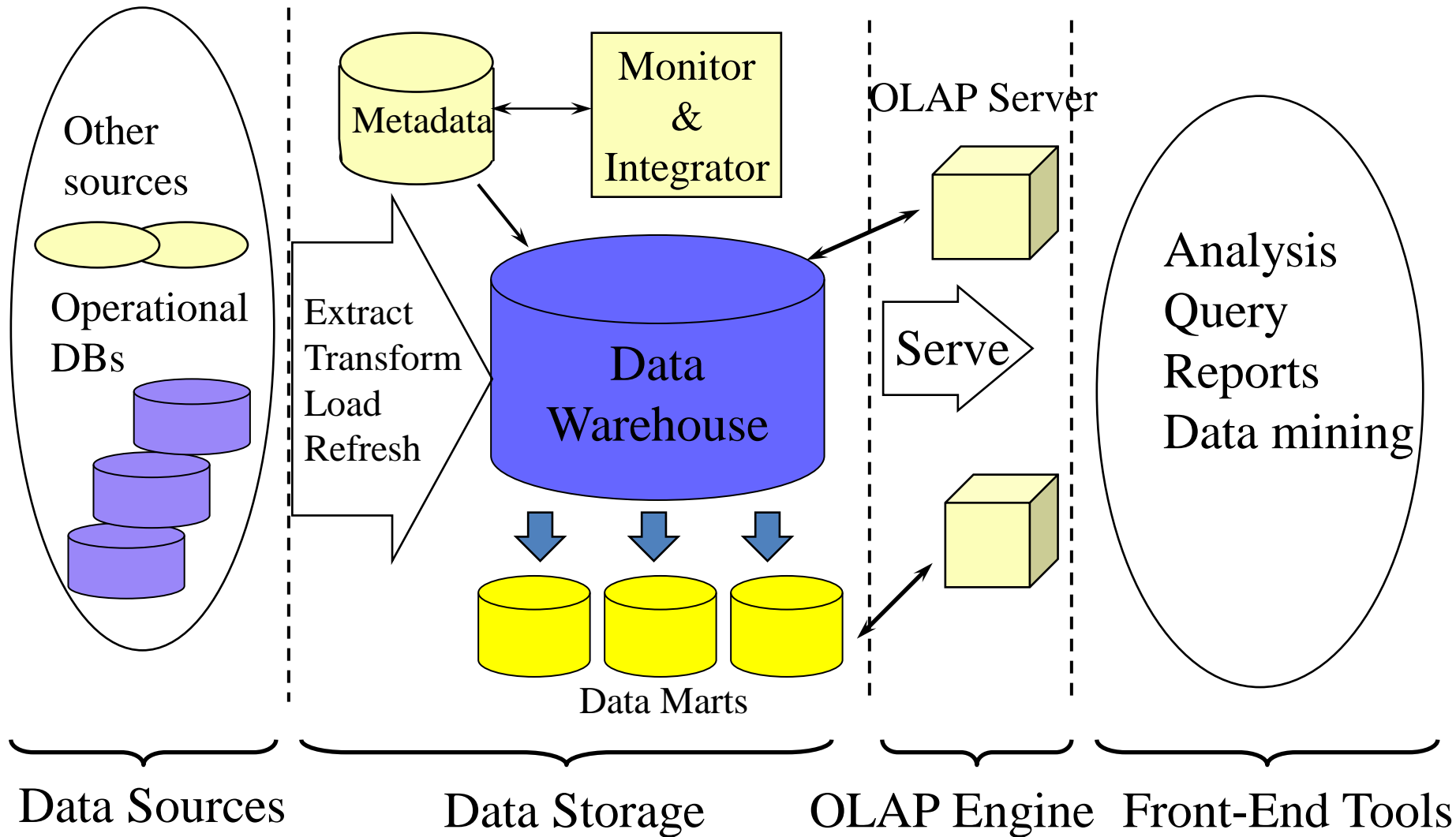
# Nonvolatile

- In a Data Warehouse Environment
  - Data is stored separately from operational systems data.
  - Data is not updated.
  - Data is loaded and accessed (two operations).
  - Operational processes such transaction processing, recovery, and concurrency control mechanisms are not required.

# Why a Separate Data Warehouse?

- Efficiency of both systems
  - DBMS— tuned for OLTP: day to day transactions
  - Warehouse—tuned for OLAP: analysis of data
- Decision Support systems require:
  - historical data
  - consolidation (aggregation, summarization) of data from heterogeneous sources
  - Quality data – anomalies removed

# Traditional Data Warehouse Environment



# Data Warehouse Models

- Enterprise warehouse
  - Large scale data warehouse used enterprise wide
- Data Mart
  - Smaller, focuses on data for a department or subject
    - Independent – created separately from the data warehouse
    - Dependent - a subset of data from the data warehouse
- Virtual warehouse
  - Works with the operational databases
  - Not all data utilised

# Reading

- Chapter 4, section 4.1 of:
  - Jiawei Han, Micheline Kamber, Jian Pei, (2011), Data Mining: Concepts and Techniques, Third Edition, The Morgan Kaufmann Series in Data Management System.
- Chapter 3, section 3.2 of:
  - Sharda, Delen, Turban (2018), Business Intelligence Analytics and Data Science: A management Perspective