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Data warehouse modeling

- Data warehouse modeling is the process of designing the schemas of the detailed and summarized information of the data warehouse.
- The goal of data warehouse modeling is to develop a schema describing the reality, or at least a part of the fact, which the data warehouse is needed to support.

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Data warehouse modeling

- Data warehouse modeling is an essential stage of building a data warehouse for two main reasons.
 - Firstly, through the schema, data warehouse clients can visualize the relationships among the warehouse data, to use them with greater ease.
 - Secondly, a well-designed schema allows an effective data warehouse structure to emerge, to help decrease the cost of implementing the warehouse and improve the efficiency of using it.

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Data warehouse modeling

- Data modeling in data warehouses is different from data modeling in operational database systems.
- The primary function of data warehouses is to support DSS processes.
- Thus, the objective of data warehouse modeling is to make the data warehouse efficiently support complex queries on long term information.

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Data warehouse modeling

- data modeling in operational databases targets supporting simple transactions the database such as retrieving, inserting, deleting, and changing data.
- data warehouses are designed for the customer with general information knowledge about the enterprise, whereas operational database systems are more oriented toward use by software specialists for creating distinct applications..

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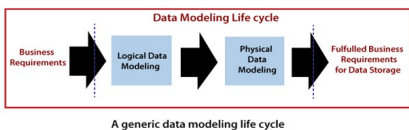
Data Modeling Life Cycle

- a data modeling life cycle is a process of transforming the business requirements to fulfill the goals for storing, maintaining, and accessing the data within IT systems.
- The result is a logical and physical data model for an enterprise data warehouse.

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Data Modeling Life Cycle

- The objective of the data modeling life cycle is primarily the creation of a storage area for business information.
- That area comes from the logical and physical data modeling stages, as shown in Figure:



A generic data modeling life cycle

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Conceptual Data Model

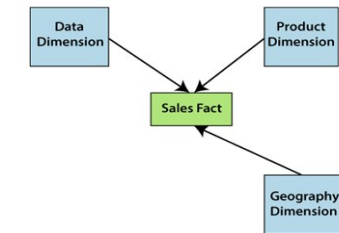
- A conceptual data model recognizes the highest-level relationships between the different entities.
- Characteristics of the conceptual data model
 - It contains the essential entities and the relationships among them.
 - No attribute is specified.
 - No primary key is specified.

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Conceptual Data Model

- It can be seen that the only data shown via the conceptual data model is the entities that define the data and the relationships between those entities.
- No other data, as shown through the conceptual data model.

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Example of Conceptual Data Model

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Logical Data Model

- A logical data model defines the information in as much structure as possible, without observing how they will be physically achieved in the database.
- The primary objective of logical data modeling is to document the business data structures, processes, rules, and relationships by a single view - the logical data model.

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Logical Data Model

- **Features of a logical data model**
 - It involves all entities and relationships among them.
 - All attributes for each entity are specified.
 - The primary key for each entity is stated.
 - Referential Integrity is specified (FK Relation).

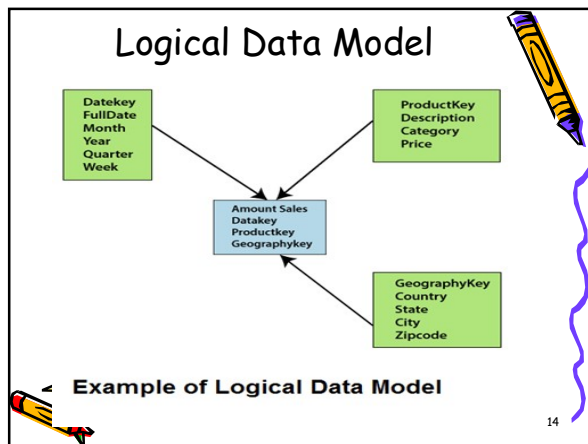
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Logical Data Model

- The phase for designing the logical data model which are as follows:
 - Specify primary keys for all entities.
 - List the relationships between different entities.
 - List all attributes for each entity.
 - Normalization.
 - No data types are listed




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Physical Data Model


- Physical data model describes how the model will be presented in the database.
- A physical database model demonstrates all table structures, column names, data types, constraints, primary key, foreign key, and relationships between tables.



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Physical Data Model


- The purpose of physical data modeling is the mapping of the logical data model to the physical structures of the RDBMS system hosting the data warehouse.
- This contains defining physical RDBMS structures, such as tables and data types to use when storing the information.
- It may also include the definition of new data structures for enhancing query performance.



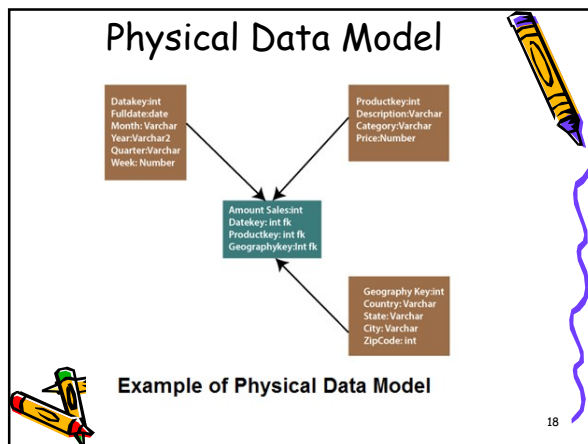
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Physical Data Model

- Characteristics of a physical data model
 - Specification all tables and columns.
 - Foreign keys are used to recognize relationships between tables.
- The steps for physical data model design which are as follows:
 - Convert entities to tables.
 - Convert relationships to foreign keys.
 - Convert attributes to columns.



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Data warehouse Schemas

- The following schemas will be discussed in this course:
 - **Star schema:** One fact table and a set of dimension tables
 - **Snowflake schema:** Avoids redundancy of star schemas by normalizing dimension tables
 - **Starflake schema:** Combination of the star and snowflake schemas, some dimensions normalized, other not
 - **Constellation schema:** Multiple fact tables that share dimension tables

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Star schema

- **Star Schema** in data warehouse, is a schema in which the center of the star can have one fact table and a number of associated dimension tables.
 - It is known as star schema as its structure resembles a star.
 - The Star Schema data model is the simplest type of Data Warehouse schema.
 - It is also known as Star Join Schema and is optimized for querying large data sets.

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Fact Tables

- A Fact table in a star schema contains facts and is connected to dimensions.
- A fact table has two types of columns:
 - A column that includes Facts
 - Foreign Key to Dimensions Table
- Generally, the primary key of a fact table is a composite key that is made up of all the foreign keys that make up the table.

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Fact Tables

- Fact tables can contain detail-level facts or aggregated facts. Fact tables that include aggregated facts are often called summary tables.
- Fact tables usually contain facts that have been aggregated to some level.

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Dimension Tables

- A dimension is an architecture that categorizes data in a hierarchy.
 - A dimension without hierarchies and levels is called a flat dimension or list.
 - Each dimension table's primary key is part of the composite primary key of the fact table.
 - A dimension attribute is a descriptive, textual attribute that helps describe a dimensional value.

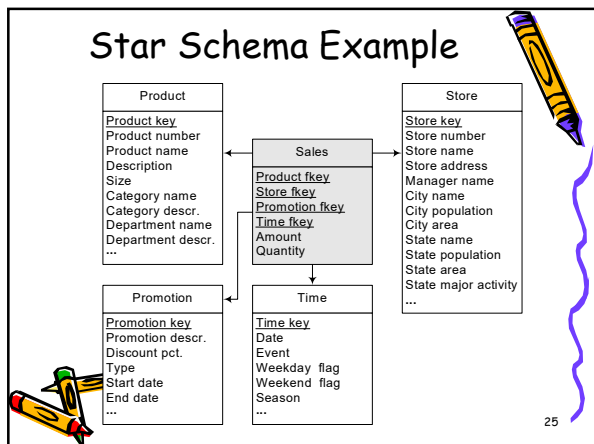
Fact tables are usually larger than dimension tables.

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Example of Star Schema Data Modelling

- In the following Star Schema example, the fact table is at the center which contains keys to every dimension table such as product key, store key, promotion key, time key & other attributes like amount and quantity.

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- ### Characteristics of Star Schema
- Every dimension in a star schema is represented with the only one-dimension table.
 - The dimension table should contain the set of attributes.
 - The dimension table is joined to the fact table using a foreign key
 - The dimension table are not joined to each other
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- ### Characteristics of Star Schema
- Fact table would contain key and measure
 - The Star schema is easy to understand and provides optimal disk usage.
 - The dimension tables are not normalized.
 - The schema is widely supported by BI Tools
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- ### Advantages of Star Schema
- Star schemas have a more straightforward join logic compared to other schemas for fetching data from highly normalized transactional schemas.
 - As opposed to highly normalized transactional schemas, the star schema simplifies common business reporting logic, such as reporting and period-over-period.
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- ### Advantages of Star Schema
- Star schemas are widely used by OLAP systems to design cubes efficiently.
 - A star schema can be used as a source without designing a cube structure in most major OLAP systems.
 - By enabling specific performance schemes that can be applied to queries, the query processor software in Star Schema can offer better execution plans.
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- ### Disadvantage of Star Schema
- Since the schema is highly de-normalized, data integrity is not enforced well.
 - Not flexible in terms of analytical needs.
 - Star schemas do not reinforce many-to-many relationships within business entities.
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