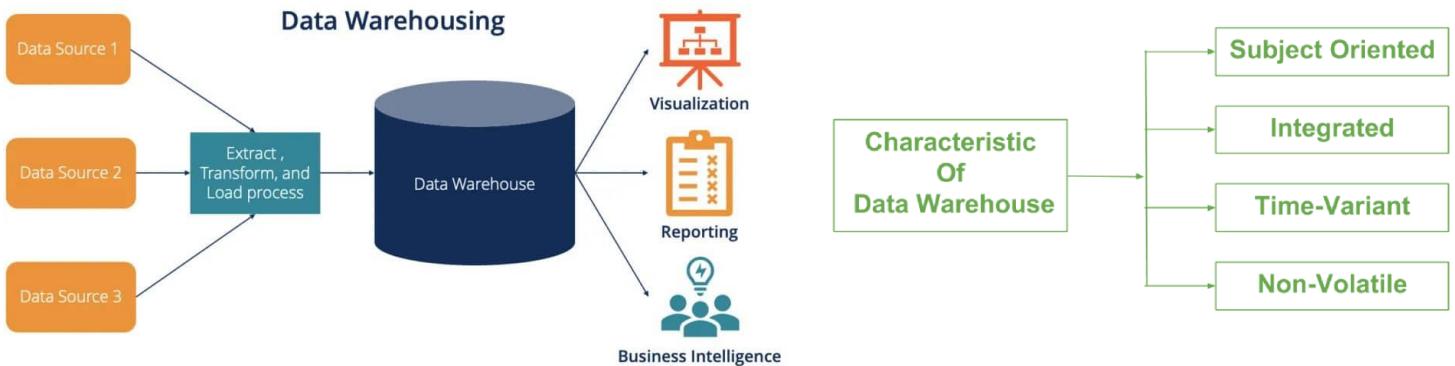


## [ Data Warehouse ]

### What is a data warehouse?

A data warehouse is a [central repository](#) of information that can be [analyzed](#) to make [reports](#) and [informed decisions](#). A Data warehouse is typically used to connect and analyze business data from [heterogeneous sources](#). The data warehouse is the core of the [BI](#) system which is built for data analysis and reporting.



### What are the characteristics of a data warehouse?

- [Subject-oriented](#): DW typically provides information on a topic (Sales, HR, Supply chain)
- [Time-variant](#): Time variant keys (e.g., for the date, month, time) are typically present.
- [Integrated](#): A data warehouse combines data from various sources. These may include a cloud, relational databases, flat files, structured and semi-structured data.
- [Non-volatile](#): Prior data isn't deleted when new data is added. Historical data is preserved for comparisons, trends, and analytics

### What is a database vs. a data warehouse?

#### Database

- Database stores the [current data](#) required to power an application.
- Highly [normalized](#), [static](#) schemas

#### Data warehouse

- [Data warehouse](#) stores [current](#) and [historical](#) data for the purpose of [analyzing](#) the data.
- [Denormalized](#) schemas, such as the [Star](#) schema or [Snowflake](#) schema

### What is a bigdata vs. a data warehouse?

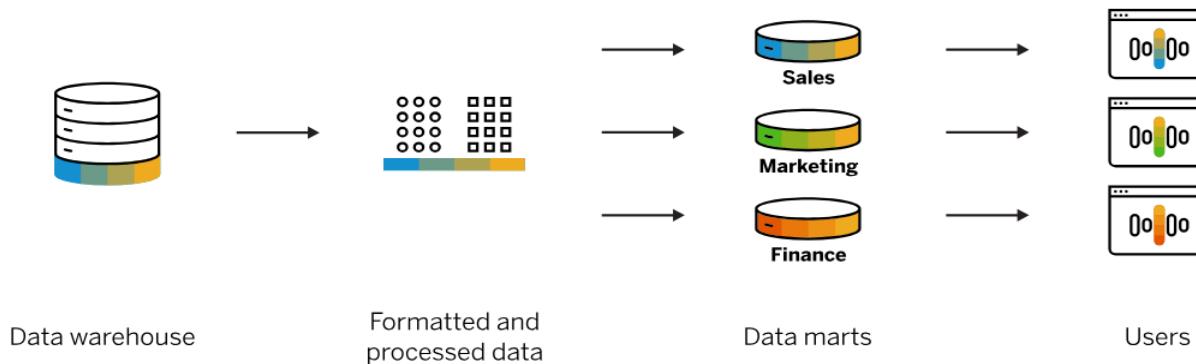
#### Big data

- Big data is [a technology](#) to store and manage large amounts of data.
- It takes [structured](#), [non-structured](#) or [semi-structured](#) data as input
- Big data doesn't follow any [SQL queries](#) to fetch data from database

#### Data warehouse

- Data warehouse is [an architecture](#) used to organize data from heterogeneous sources
- It only takes [structured data](#) as an input
- In data warehouse we use [SQL queries](#) to fetch data from relational database

## Data warehouse vs Data mart?



### Data mart

- Decentralized, specific subject area
- A single community or department
- A single or a few sources, or a portion of the data warehouse
- Small, generally up to 10's of gigabytes

### Data warehouse

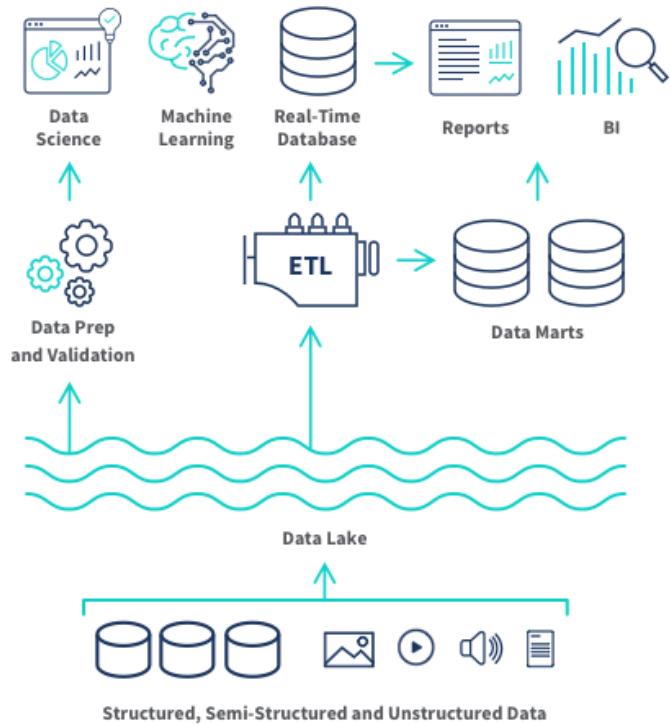
- Centralized, multiple subject areas
- Organization-wide
- Many sources
- Large, can be 100's of gigabytes to petabytes

## Data warehouse vs Data Lake?

### Data lake

- Takes Structured, semistructured and unstructured data, from sensors, apps, websites, etc.
- May not have a predefined purpose, typically used for machine learning & deep analysis
- Used by Data engineers & Data Scientists
- Schema-on-Read

### Data Lake



### Data warehouse

- Takes Structured, processed data, from operational databases, applications and transactional systems.
- Predefined purposes for business intelligence batch reporting and visualization
- Used by Data engineers, business analysts and data analysts
- Schema-on-Write

## What is the difference between OLTP and OLAP?

OLTP stands for **Online Transaction Processing**.

OLTP has the work to administer day-to-day transactions in any organization. The main goal of OLTP is data **processing** not data analysis.

OLAP stands for **Online Analytical Processing**.

OLAP systems have the capability to **analyze** database information of multiple systems at the current time. The primary goal of OLAP Service is data analysis and not data processing.

OLTP (RDBMS)	OLAP (DW)
Consists of operational <b>current &amp; detailed data</b>	Consists of <b>historical &amp; summarized data</b>
<b>application-oriented</b> . Used for business tasks.	<b>Subject-oriented</b> . Used for Analytics & Mining
OLTP DB are <b>isolated</b> as applications	OLAP <b>integrated</b> per subject area (data mart)
Relatively <b>small</b> , data is archived in <b>MB, and GB</b> .	<b>Large</b> amount of data, stored typically in <b>TB, PB</b>
Supports <b>CRUD</b> (Create, Read, Update, Delete)	Supports <b>only Read</b>
It is <b>volatile</b>	It is <b>non-volatile</b>
<b>Normalized</b> Schema (3NF)	<b>Denormalized</b> Schema

## What are the processes that can be done in the data warehouse?

### 1. Data Extraction:

extracting data from various data sources, such as databases, applications, and other sources, and transforming it into a format that can be loaded into the data warehouse.

### 2. Data Transformation:

cleaning, filtering, merging, and transforming the data extracted from various sources to ensure consistency and accuracy.

### 3. Data Loading:

loading the transformed data into the data warehouse. This can be done through various methods such as batch processing, real-time data integration, or incremental loading.

### 4. Data Aggregation:

summarizing or consolidating the data into meaningful information that can be used for analysis and reporting.

### 5. Data Analysis:

querying and analyzing the data in the data warehouse to gain insights and make informed decisions. This can be done through various tools such as SQL, OLAP, and data mining.

### 6. Data Visualization:

presenting the analyzed data in a visual format such as charts, graphs, and dashboards, to help stakeholders understand the information easily.

## Data modeling

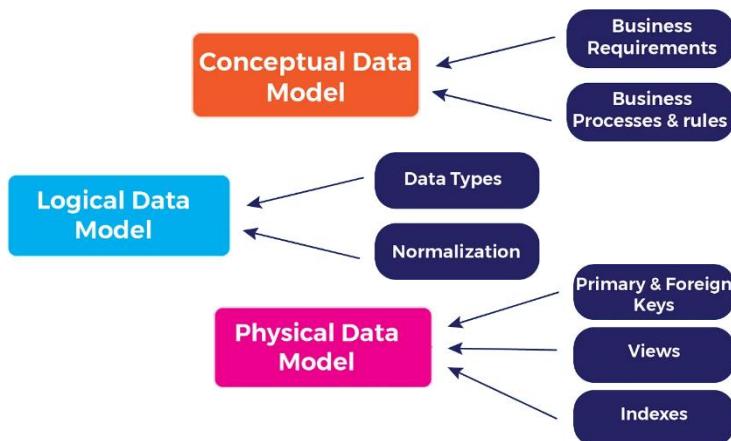
The process of developing a visual representation of an entire information system or sections to express connections between data points and structures

## Types of data modeling

1. Conceptual Data Model

2. Logical Data Model

3. Physical Data Model



## Data warehouse modeling

the process of [designing the schema](#) or structure of the data warehouse, including the tables, columns, relationships, and constraints that will be used to [store](#) and [organize](#) the data.

The goal of data modeling in a data warehouse is to create a structure that is [optimized for reporting and analysis](#), and that can support [complex queries and aggregations](#).

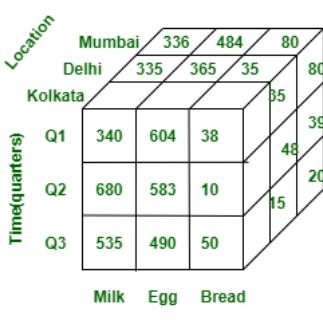
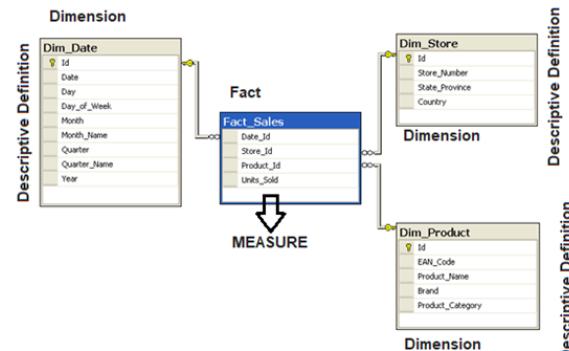
## Types of Data warehouse modeling

- 1- Entity-Relationship (ER) modeling
- 2- Data Vault modeling
- 3- Enterprise Multi-Dimensional (EDW) modeling

the most commonly used approach in data warehousing.

It involves organizing data into **dimensions** and **facts**.

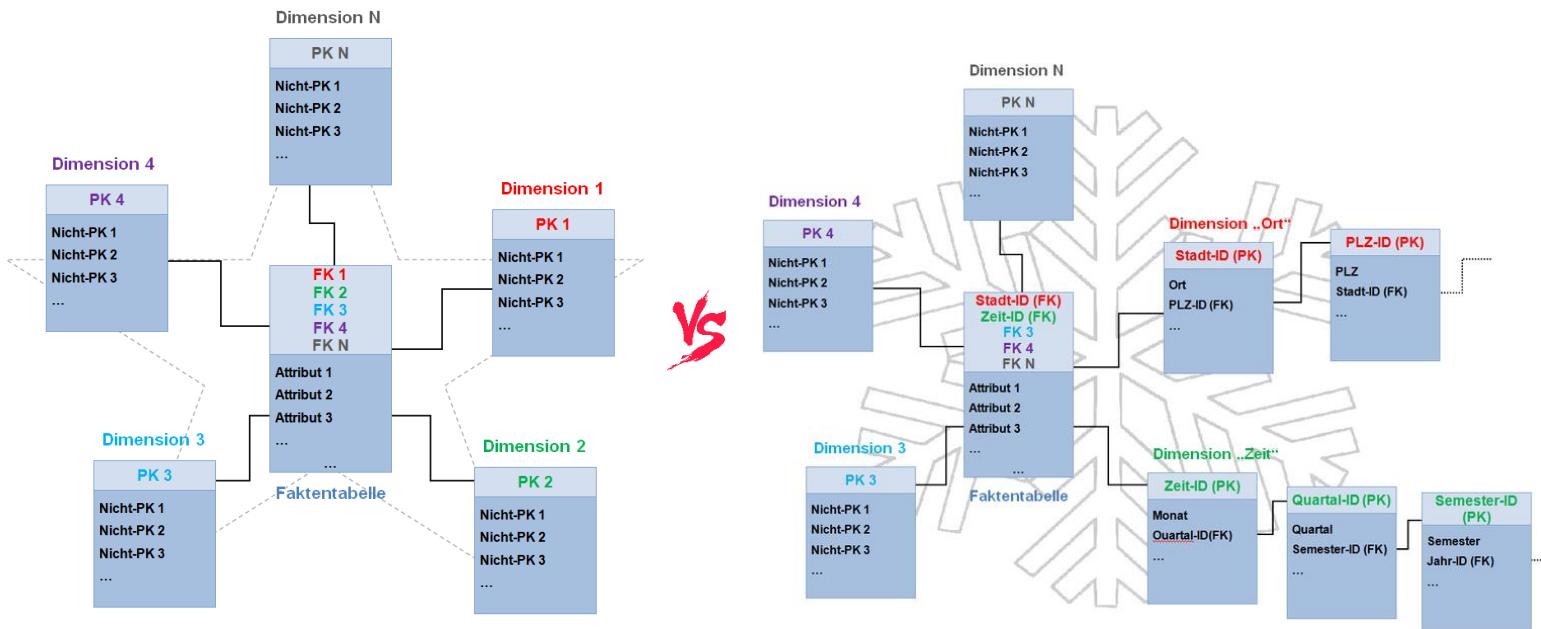
- o Dimensions tables ([Descriptive Definition](#))  
represent the various attributes of the data such as time, location, and product.
- o Fact table ([Numerical Quantities](#))  
represent the measures or metrics that will be analyzed, such as sales or revenue
- o Data cube ([data structure](#))  
represent the multidimensional relationships between measures and dimensions.  
They provide a fast and efficient way to [retrieve](#) and [analyze](#) data.



3D data cube  
Represented in  
2D data table

Time	Location="Kolkata"			Location="Delhi"			Location="Mumbai"		
	item			item			item		
	Milk	Egg	Bread	Milk	Egg	Bread	Milk	Egg	Bread
Q1	340	604	38	335	365	35	336	484	80
Q2	680	583	10	684	490	48	595	594	39
Q3	535	490	50	389	385	15	366	385	20

# Multi-Dimensional Modelling Techniques



## Star Schema

The most common technique and basic modelling type and is easy to understand. In which **Fact** table **connects** with **all Dimension** tables. Used to develop DWH and Data marts

## Star Schema

- **more redundant data, difficult to change** or maintain.
- **less complex, easy to understand.**
- **All dimension tables are directly connected to the fact table**
- **fewer foreign keys, query execution is faster** and takes lesser time.
- Better for **one to one**, or **one to many** relationships
- Both the fact table and dimension tables are **denormalized**.
- follows a **top-down** approach.

## Snowflake Schema

An **extension** of the Star Schema where that the dimensional table is **normalized** into **multiple lookup tables**, resulting in a more **complex** structure, in snowflake schema **not** all dimension tables are directly related to fact table

## Snowflake Schema

- **less redundant data, easier to change** and maintain
- **more complex, difficult to understand.**
- **Not all dimension tables are directly connected to the fact table**
- **more foreign keys, query execution is slower** and takes more time
- Better for complex relationships i.e., **many to many** relationships
- **fact** table is **denormalized**, while dimension tables are **normalized**.
- follows a **bottom-up** approach.