

The architecture of DW & BI.

• Data Warehousing Concept -

- process of constructing & using a data warehouse.
- Integrating data from multiple heterogeneous sources → analytical reports

- Data Warehouse → store historical info from multiple sources to allow you to analyze & report on related data.

Goals →

- reporting / analysis
- maintain → organization's historical info.
- foundation → decision making.

* organizations → info → data warehouse.

- Increasing customer focus
- Repositioning products & managing products
- Analysing operations & looking for sources of profit
- managing customer relationships, making environmental corrections & managing cost of corporate assets.

* Characteristics of DW →

- Subject oriented
- integrated
- Non-volatile
- Time Variant
- access & high speed query.
- end user → time sensitive
- large amt → historical data
- queries → large amt of data.

Architecture of DW →

- Single tier architecture
- two-tier "
- three-tier " (multi-tier archi)

Need for DW →

- Business user
- store historical data
- Make strategic decisions
- For data consistency & quality
- High response time
- Data integration
- BI
- Performance
- Data quality

Benefits →

- Improved decision making
- Increased efficiency
- Better data quality
- Historical analysis (large data)
- cost saving
- competitive advantage
- Improved data governance

Data warehouse types →

- Enterprise warehouse
- Data mart
- Virtual warehouse

A ETL process → Extraction, Transformation & Loading

Business Intelligence (BI)

Data Warehouse (DW)

- set of strategies & technologies to analyze & visualize data to make decisions
- data → user's behaviour belongs to BI.
- set of technologies & strategies
- present data → reports, charts & graphs.
- top executives & senior managers use BI.
- example - Datapine.
- create business insights.
- provides user-friendly tools for data analysis & visualization.

- central location → store consolidated data from multiple data sources
- customer consumption → data warehouse
- storage
- presents data in tables
- data engineers, data & business analysts use data warehouses.
- example - Amazon Redshift
- storing data from several sources.
- Typically accessed via SQL queries or reporting tools.

ROLAP

MOLAP

HOLAP

implementation

- based on relational DBMS

- based on multidimensional DBMSs

- both relational & multidimensional techniques

Adv →

- handle large data
- can leverage functionalities
- inherit in the relational database

- excellent performance
- perform complex calculations

- HOLAP tools can utilize both pre-calculated cubes & roll-up sources

Disadv →

- Performance → slow
- Limited by SQL functionalities

- limited data handling
- requires additional investment

- supports disadv of MOLAP.

ROLAP

MOLAP

SCHEMA

- User Star Schema
- Additional dimensions can be added dynamically.

- User Data cubes

- Add dimensions require reorganization of data base

Database size

- Medium to large

- Small to medium

Architecture

- client/server

- client/server

Access -

- support ad-hoc requests
- Good with small data sets
- Avg for medium to large data set.

- limited to pre-defined dimensions
- Faster for small to medium data
- Avg for large datasets.

Speed →

Flexibility & Scalability

- High

- Low

OLTP

OLAP

- support long transⁿ → complex queries.
- tradⁿ DBMS to accommodate a large volume of real-time transaction.
- Tables → OLTP database are normalized.
- are designed for use for frontline workers.
- contains current data.
- applⁿ → current data.
- DB size → 100GB to TB
- used for day-to-day operations.
- flexibility → low
- priority → high performance
- performance high → few seconds per query
- Updating → continuous & irregular
- short transⁿ → query & updates.
- has multidimensional schema
- tables → are not normalized.
- for data scientists, business analyst & knowledge.
- contains historical data.
- focuses on information out.
- DB size - 100MB to GB
- used for decision support system
- high.
- ~~maybe low~~ high flexibility.
- performance → low for complex query.
- updating → periodic & regular.

- * Data Models →
- 1) conceptual models
 - 2) logical model
 - 3) Physical model

Types of Business Intelligence Tools -

- spreadsheets.
- dashboards
- Data mining tools
- Adhoc data analysis
- Online analytical processing
- mobile BI
- Realtime BI
- Operational BI
- Open source BI
- Collaborative BI
- Data visualization software
- Reporting & query software
- Data warehousing tools.

* Data Cube -

allows data to be modelled & viewed in multiple dimensions.

n-D base cube → base cuboid.

0-D cuboid → highest level of summarization → apex cuboid

- Adv →
- ease in aggregating & summarizing data.
 - provide better visualization of data.
 - huge amt of data → simplified way
 - increases overall efficiency of data warehouse.
 - aggregated data in data cube → analysing data fast & thereby reducing access time.

Fact table

- contains measurements along attributes of dimension table.
- contains less attributes & more records.
- Fact table grows vertically
- Fact table \rightarrow primary key which concatenation of primary keys \rightarrow dimension table
- schema contains less no. of fact tables
- fact table can have data in numeric as well textual format.

Dimension table

- contains attributes along which fact table calculates metric
- contains more attributes & less records.
- dimension table grows ~~less~~ horizontally
- each dimensional table contains \rightarrow primary key.
- schema contains more no. of dimension tables.
- always contains attributes in textual format.

Star Schema

- simple & common modelling paradigm where data warehouse comprises of fact table.
- star schema \rightarrow not use normalization.
- contains fact & dimension tables.
- simple to understand & easily.
- high level of data redundancy.
- cube processing is faster.
- uses more space.

Snowflake Schema

- kind of star schema which includes hierarchical form of dimensional tables
- uses normalization to elimination redundancy of data.
- sub-dimension tables \rightarrow including fact & dimensional tables
- Hard to understand & design.
- low level of data redundancy
- cube processing might be slow because of complex join.
- It uses less space.