Business Intelligence and Data Warehousing (ANL408)

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Recap from last week....

- Data Warehouse Architecture
- Enterprise Architecture
- Federated Architecture
- Multi-tiered Architecture
- Layers of data warehouse
- Staging Layer
- Temporary Staging Layer
- Persistent Staging Layer
- Data Mart
- Types of Data Mart
- Advantages and Drawbacks of Data Mart
- Data Mart vs Data Warehouse

Traditional Data Warehouse

Centralized warehouse repositories within an organization. They are designed to support:

- Reporting
- Analytics
- Business Intelligence

DECISION – MAKING!!!

Traditional Warehouse - Characteristics

- Structured Data Storage
- ETL Processes
- Dimensional Modelling
- Query and Reporting
- Data Marts
- Historical Data Storage
- Centralized Architecture
- Scalability and Performance

Modern Data Warehouse

Advanced data management platforms that are designed to handle large volumes of structured and unstructured data.

REAL-TIME INSIGHTS!!!

Modern Warehouse - Characteristics

- Cloud Based Architecture
- Massively Parallel Processing (MPP)
- Columnar Storage
- In-Memory Processing
- Integration with Big Data Technologies
- Advanced Analytics and Machine Learning
- Self-Service Analytics

Traditional vs Modern Data Warehouse

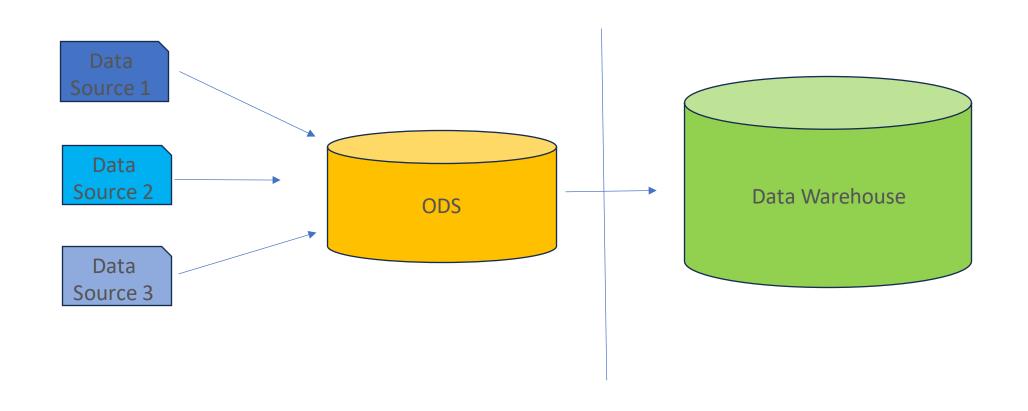
| Parameter | Traditional Data Warehouse | Modern Data Warehouse |
|-------------------------|--|---|
| Architecture | On-premises or Private Cloud | Cloud-based (private, public or hybrid cloud) |
| Scalability | Limited scalability, requires upfront capacity planning | Elastic scalability, can scale resources up or down as needed |
| Processing Model | Single-node architecture with shared resources | Distributed, massively parallel processing (MPP) architecture |
| Data Processing | Row-based storage and processing | Columnar storage and processing for optimized query performance |
| Integration | Limited integration with external data sources | Integrates with big data technologies and external data sources |
| Real-time Analytics | Limited support for real-time analytics | Support for real-time analytics and streaming data processing |
| Cost | Upfront hardware and software costs, ongoing maintenance | Pay-as-you-go pricing model, based on resource usage |
| Maintenance and Updates | Manual maintenance and updates | Automated maintenance and updates through cloud providers |

Operational Data Store (ODS)

"A subject-oriented, integrated, volatile, current valued data store containing detailed data."

"Type of database that serves as a central repository for integrating and storing data from various operational systems within an organization."

Visualizing ODS



Characteristics of ODS

- Real-Time or Near Real-Time Data Integration
- Centralized Data Repository
- Transactional Processing
- No need for long history
- Data Quality and Consistency
- Parallel vs Sequential

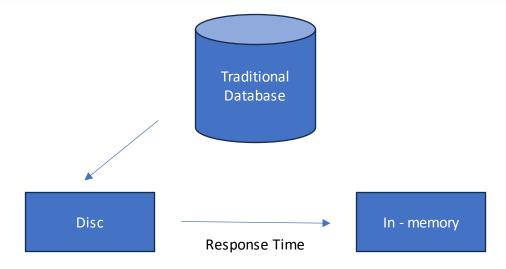
Staging vs ODS

| Parameter | Staging Area | ODS |
|---------------------|--|---|
| Purpose | Temporary storage for raw data before transformation | Central repository for integrated and current operational data |
| Data Transformation | Minimal transformation, raw data format | Integrated, cleansed, and transformed data |
| Data Persistence | Temporary, overwritten or deleted after loading | Persistent, historical data (data retained as per business needs) |
| Performance | Optimized for data ingestion and processing efficiency | Optimized for real-time or near real-time updates and queries |
| Schema Flexibility | Flexible schema to accommodate changes in source data | Structured and stable schema for consistent data model |

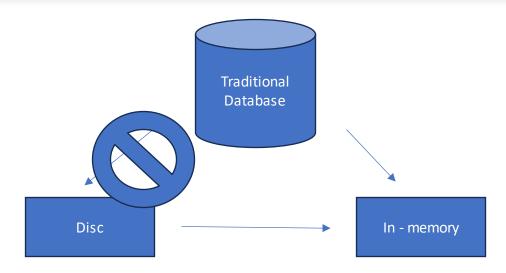


"Type of database management system (DBMS) that primarily stores and manages data in main memory (RAM) rather than on disk storage."

Traditional Databases



In-memory databases



Characteristics: In Memory Databases

- Data Storage in RAM
- High Performance
- Data Compression
- Columnar Storage
- Durability and Persistence (Periodic Snapshots or backups to disc)
- Scalability

Metadata: Overview



"Data about Data E.g. Data Dictionary"



"Provides information about the structure, content, and context of the data, enabling users and administrators to understand, manage, and utilize the data effectively."

Types of Metadata



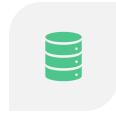
STRUCTURAL METADATA (TABLES, KEYS, RELATIONSHIPS, ETC.)



OPERATIONAL METADATA (ETL, MAINTENANCE, ETC.)



BUSINESS METADATA (BUSINESS CONTEXT AND SEMANTICS)



TECHNICAL METADATA (STORAGE STRUCTURE, INDEXES, ETC.)



DATA LINEAGE METADATA (TRACKS MOVEMENT OF DATA IN DWH)



DATA QUALITY METADATA (PROFILING, VALIDATION RULES, ETC.)



USAGE METADATA (LOGS, QUERY PERFORMANCE, ETC.)

Benefits of Metadata

Improved Data Understanding

Enhanced Data Quality

Increased Data Accessibility

Streamlined Data Governance

Empowered Data Analytics

Metadata Best Practices

- Standardize Metadata Definitions
- Document Metadata
- Centralize Metadata management
- Automate metadata capture
- Implement metadata versioning
- Enable search and discovery
- Monitor metadata usage

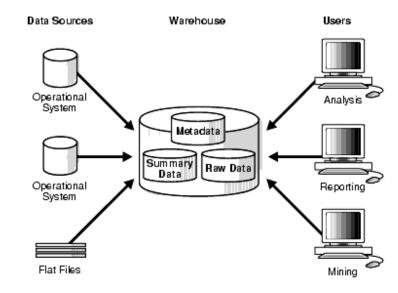
Data Warehousing approach

Methodology or strategy employed in designing, implementing, and managing a data warehouse environment.

- Inmon's approach (Top-Down Approach)
- Kimball's approach (Bottom-Up Approach)

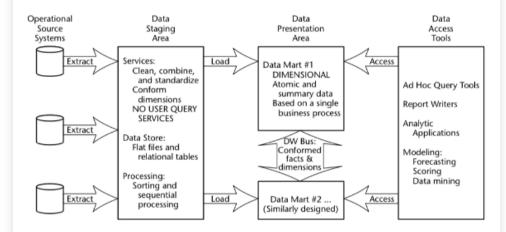
Inmon's Data warehousing Approach

- Enterprise DW
- Normalized Data Model
- Data Integration
- Data Staging
- Data Marts (subset of DW)
- Long Term Strategic Perspective



Kimball's Data warehousing Approach

- Dimensional Data Modelling
- Star and Snowflake Schema
- Data Mart-Centric
- Iterative Development
- Business User Empowerment
- Botton-Up Scalability



Kimball's vs Inmon's Approach

| Parameter | Kimball's Approach | Inmon's Approach |
|-----------------------------|---|---|
| Philosophy | Bottom-Up Approach | Top-Down Approach |
| Centralization | Emphasizes decentralized data marts | Advocates for a centralized Enterprise Data Warehouse (EDW) |
| Data Integration | Data Marts are independently integrated | Extensive data integration process |
| Data Marts | Business specific and tailored | Derived from centralized DW |
| Development Approach | Iterative and Incremental | Long-term strategic |
| User Empowerment | Prioritizes business user empowerment | Focuses on building a robust foundation for EDW |
| Data Modelling | Dimensional | Normalized |
| Scalability | Botton-Up | Top-Down |
| Agility | Agile Approach | Strategic Approach |

Kimball's front and back room analogy

- Front Room (User-Facing)
- Back Room (Technical)

