(Big) Data Engineering In Depth From Beginner to Professional

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The Definitive Guide to Big Data Engineering Tasks

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Introduction To Data Management and Data Warehouse

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- Show what is the challenges to build a DWH?
- What is the data modeling and its design?

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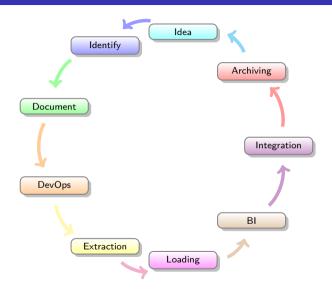
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 - Integration and publishing.
 - Data retention or **archiving** process ex: (Hot or Cold storage).

Data Management Life-Cycle

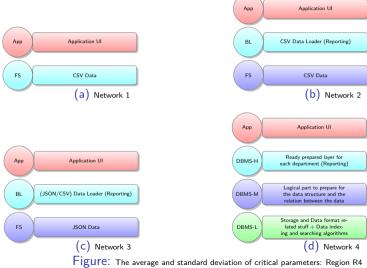




Data Abstraction

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Motivation to Data Layers (Use Case)



• How can we think about a data solution or challenges in the data products?



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Motivation to Data Layers (Solution Thinking)

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 - Ask your self the following questions:
 - Can we solve the problem using the current data structure with adding new features?
 - What if we enhance/change the data structure or modeling?
 - Could it help if we change the backend engine (ex: DBMS system)?
- To answer these questions you need to understand the **data layers**.

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- The developer for each layer hide internal irrelevant details from developer (users).
- The process of <u>hiding</u> irrelevant details from developer (user) is called data <u>abstraction</u>.

Definition

Data Abstraction and Data Independence: DBMS comprise of complex data-structures. In order to make the system efficient in terms of retrieval of data, and reduce complexity in terms of usability of users, developers use abstraction i.e. hide irrelevant details from the users. This approach simplifies database design.

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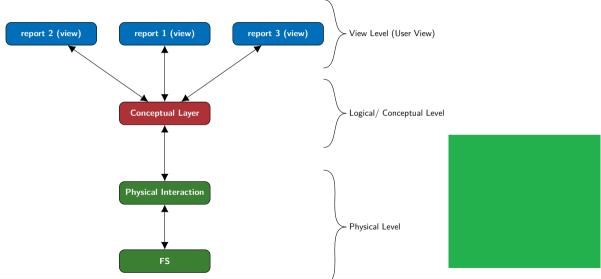
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 - Logical/ Conceptual Level.
 - View Level.



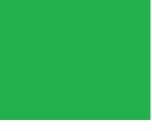
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 - Modifying indexes.
 - Change the compression algorithm or hashing technique.

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 - Usually this layer abstracted from the programmers.

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 - The programmers design this level based on the business knowledge and the requirements.

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- Not all the views is extended to all users and there is an authentication based on the category.

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- It could be designed to show the sales of product in specific region.
- We might hide information about some products based on the teams or users.

Let's answer our previous the question, How can we solve data challenges?



• Let's split the problem based on the data layers.



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 - View layer
 - When we need to add/remove/create new reports it is usually view layer.
 - We don't need to change the logical or physical layer to support the view layer.

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 - When you have missing sources into your logical layer and you need to add this source and its structure.
 - There is a performance issue in the existing reports and you need to change in the model. For example, reduce the join by creating new join table (*materialized view*).
 - Update the data type or the existing relation which could help to fix some data or performance issues.

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 - If we need to change your storage/compression/structure/access technique.
 - If we need to change the data orientation structure from row to column or key-value storage, It is time to change the physical layer.

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Introduction to DWH

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- Reporting the results after passing the data life-cycle will be from storage (Database).
- There are some challenges facing the people who work on data management backend:
 - Performance.
 - Integration.
 - Applying analytical functions.
- Vendors who are working to solve the above challenges creating their own product of DWH and their ultimate work is to optimize the above points.

Definition (What is Data Warehousing?)

A DWH is defined as a technique for collecting and managing data from varied sources to **provide meaningful business insights**. It is a blend of technologies and components which aids the strategic use of data.

The real concept was given by Inmon Bill. He was considered as a father of the DWH. He had written about a variety of topics for building, usage, and maintenance of the warehouse & the Corporate Information Factory

• The DWH is not a product but an environment.



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- The DWH is not a product but an environment.
- It is a process of transforming data into information and make it available to users in a timely manner to make a difference.
- It is an architectural construct of an information system which provides users with current and historical decision support information which is difficult to access or present in the traditional operational data store.

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- It is an architectural construct of an information system which provides users with current and historical decision support information which is difficult to access or present in the traditional operational data store.
- The DWH is the core of the BI system which is built for data analysis and reporting.

Data warehouse system is also known by the following names:

- Decision Support System (DSS).
- Business Intelligence Solution.
- Executive Information System.
- Management Information System.
- Analytic Application.
- Data Warehouse.

Differences Between DWH and Operational DB

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DWH vs Operational databases

Metric	Transactions DB	DWH	
Volume	GB/TB	TB/PB	
Historical	Short-term	Long-Term	
rows	<1000M	1000M>	
Orientation	Product	Subject or multi products	
Business Units	Product team	Multi organizational units	
Normalization	Normalized	Not required (De-normalized in many use cases)	
Data Model	Relational	Star Schema or Multi-dim	
Intelligence	Reporting	Advanced reporting and Machine Learning	
Use cases	Online transactions & operations	Centeralized storage (360°)	

Transnational DB Use cases



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Transnational DB Use cases



DWH Use cases



DWH Use cases



DWH Use cases



Types of DWH

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Types of Data Warehouse

- Enterprise Data Warehouse (EDWH) It provides decision support service across the enterprise. It offers a unified approach for organizing and representing data (DWH Model). It offers data classifications according to the subject with privileges policy.
- Operational Data Store (ODS): is a central database that provides an up-to-date (real-time) data from multiple transnational systems for operational reporting into a single DWH.
- Data Mart: A data mart is a subset of the data warehouse. It specially designed for a particular line of business, such as sales, finance, sales or finance. In an independent data mart, data can collect directly from sources.

DWH vs ODS vs Data Mart

Metric	DWH	ODS	Data Mart
Latency	Day -1	Real-time	Day -1
Data level	Transnational	Transnational	Summary
Historical	Long-term	Snapshot	Aggregated Long-Term
Size	TB/PB	GB	GB/TB
Orientation	Multi sources	Multi sources	Product
Business Units	Multi organizational units	Product team	Business team

Use Cases of Operational DB vs DWH

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• A telecommunication company named **XTec**.



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 - CRM team can report their sales and customer activities from their database.
 - Product owner can take a decision based on their system backend reports.

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 - The decision from the DHW is a global and strategical decision.
 - If the company needs to build a machine learning model which needs data from different sources. They need to load the data from a centralized database rather than read each source alone.

The Full picture required a DWH. However, we still need the other operational databases for product development perspective.



Use case (ODS)

• Why do we need the ODS?



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Use case (ODS)

- Why do we need the ODS?
- How does it fit in our system?



XTec has a call center system which handles the customer inquiries. This system requires the some data related to usage, customer information, billing details to be calculated and accumulated in real-time to be able to give the customer the right answer for his inquires.

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- So, What is the challenge for this system?
 - It needs specific information from different source systems.
 - It requires to track the source system database changes or update in real-time.
 - It's functionality is based on the aggregate data not the transactions for example (It needs the total outgoing calls till time or it needs the total charging amounts from prepaid or the available limits from billing if it is postpaid).

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- ODS uses the real-time aggregations to support the online systems from different source systems.

DWH Characteristics

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- The characteristics of DWH:
 - Integrated: DWH is an integrated environment which allows us to integrate different source systems. Data are modeled (organized) into a unified manner.
 - Time-Variant: Data modeled (organized) based on time periods (hourly, daily, weekly, monthly, quarterly, yearly, etc.)
 - Subject-oriented: *DWH* main target is to support business needs for the whole organization including (decision makers, departments, and specific user requirements).
 - Non-Volatile: It refers to the data will not erased or deleted (It could be archived and retrieved when needed). Data can be accumulated daily the new snapshots (refreshed at based on the source system interval. For example, It could be updated daily, weekly, and monthly).

DWH Architecture

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 - Metadata layer.
 - System operations layer.

DWH Architecture Overview

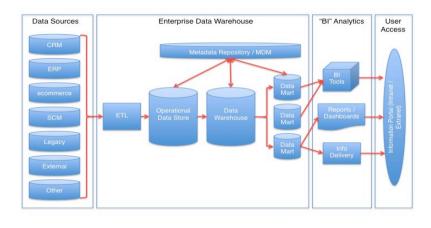


Figure: taken from XXXX

ETL Process

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ETL vs ELT When? Why?

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Data Models

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Data model is

• An abstract model that organizes elements of data.

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- It describes the conceptual design of a business or an application with its flow, logic, semantic information (rules), and how things are done.
- It refers to a set of concepts used in defining such as entities, attributes, relations, or tables.

Data model is not

- a science.
- a static design for each organization.
- a type of database.
- a new invention which needs to be done for each project.

Data model is

- an engineering design practices.
- a general concepts which lead to build full architecture.
- different based on the use case and the database type.
- customizable and we can utilize some of ready built architecture.
- implementing using different ways.
- affecting the information reporting performance and ways.

Why does data models are important?

- Data models are currently affecting software design.
- It decides how engineers will think about the problem they are solving.

Data Model Design

• You need to build a home. So, how do we design this home?



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 - Hire a contractor to build (implement the design) the home.
 - This phase will implement the design but it also include some detail related to the actual way to build the tools and the material. (Physical Design)

DWH Architecture Overview

There are mainly three types of Datawarehouse Architectures: -

- Single-tier architecture.
- Two-tier architecture.
- Three-tier architecture.

File Formats

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• Any Big Data solution working based distributed systems.



- Any Big Data solution working based distributed systems.
- What is distributed systems in brief?



Data Encoding and Formats

• Any Big Data solution working based distributed systems.



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Data Compression Technique

• Any Big Data solution working based distributed systems.



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- What is distributed systems in brief?



Data Archiving and Retention

• some details about hot vs cold storage,



Different Types of Storage

Cold storage vs Hot storage

some details about hot vs cold storage,



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DWH On Cloud

Further Readings and Assignment