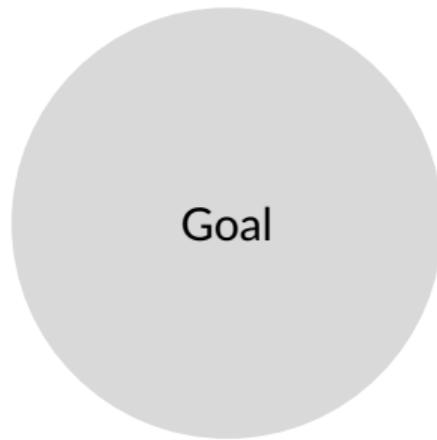


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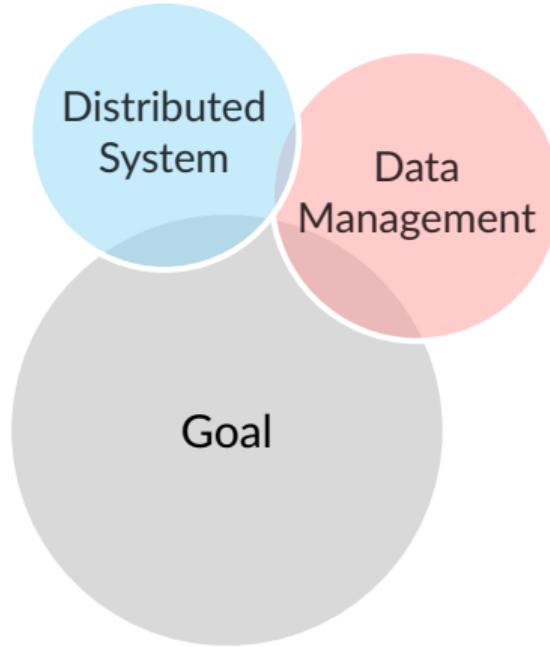
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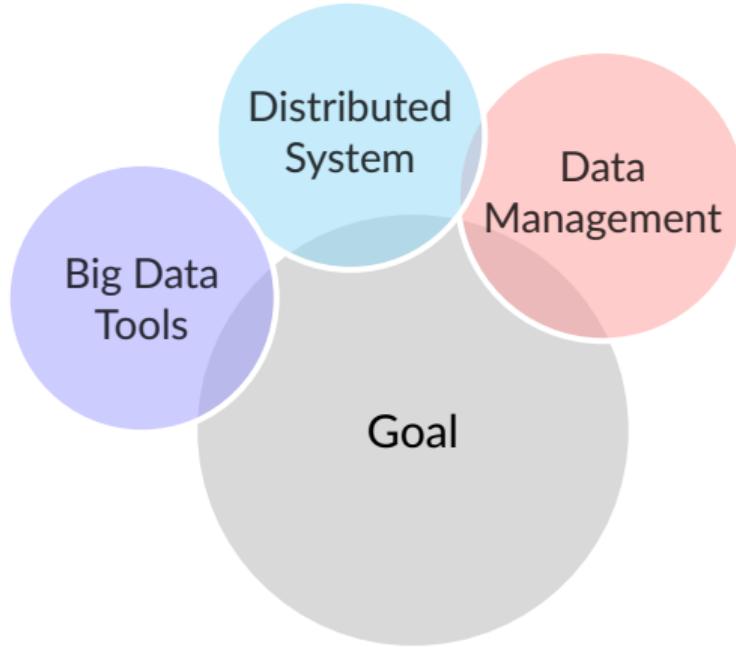
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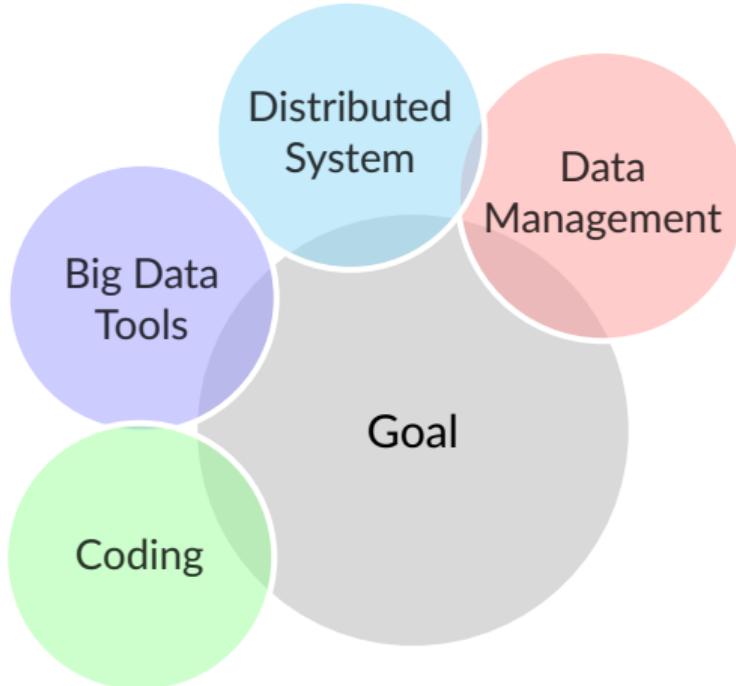
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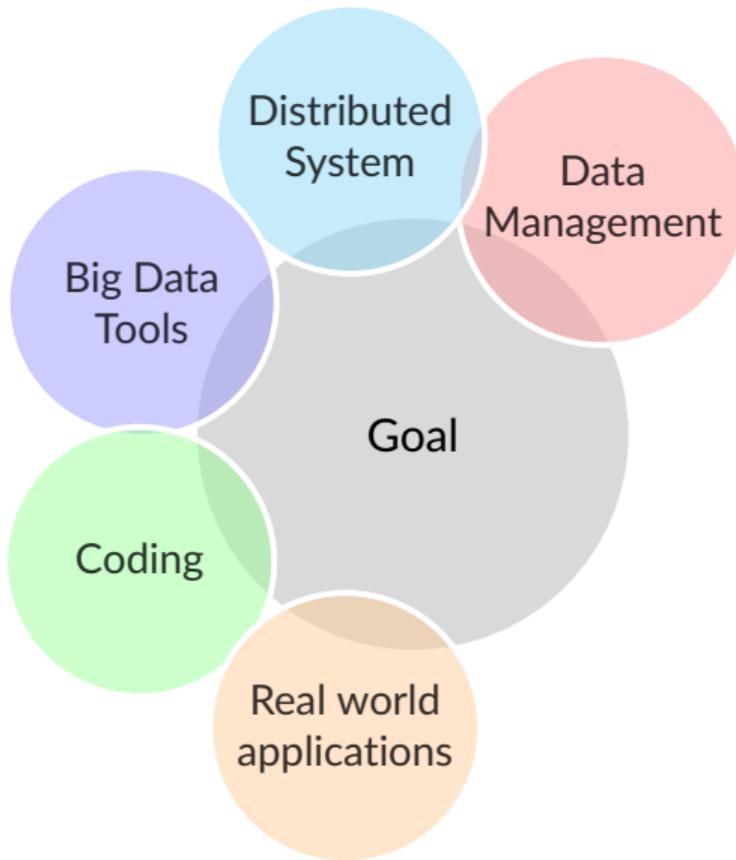
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# Learning Objectives

- Simplify the concepts in data management.



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# Getting max benefit from this course

## Take the course advantage

- Follow the videos order as described.

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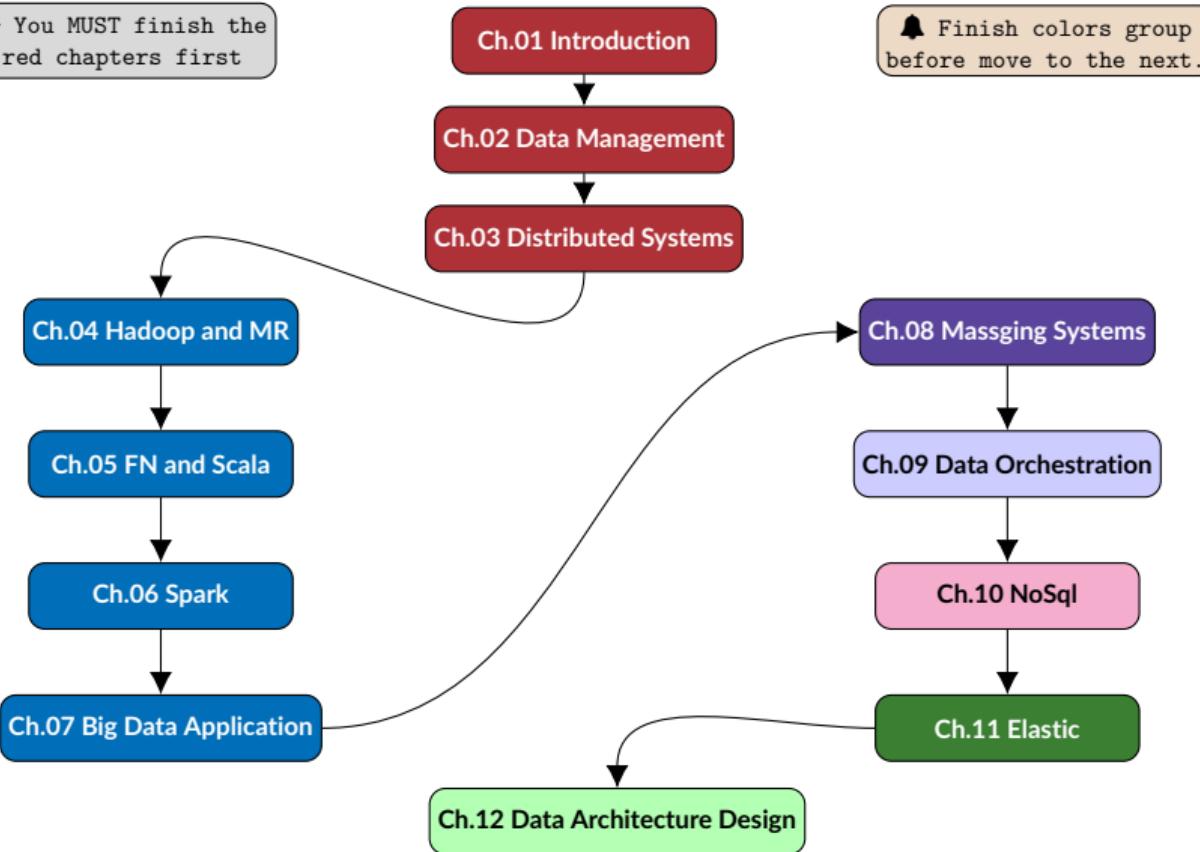
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⚠ You MUST finish the red chapters first



🔔 Finish colors group before move to the next.

# Chapter Dependencies (Jump Out Path)

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Ch.02 Data Management

🔔 Finish colors group before move to the next.

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Ch.04 Hadoop and MR

Ch.05 FN and Scala

Ch.06 Spark

Ch.07 Big Data Application

Ch.08 Messaging Systems

Ch.09 Data Orchestration

Ch.10 NoSQL

Ch.11 Elastic

Ch.12 Data Architecture Design

# Assignments and Labs

## Remark

- Full project code.

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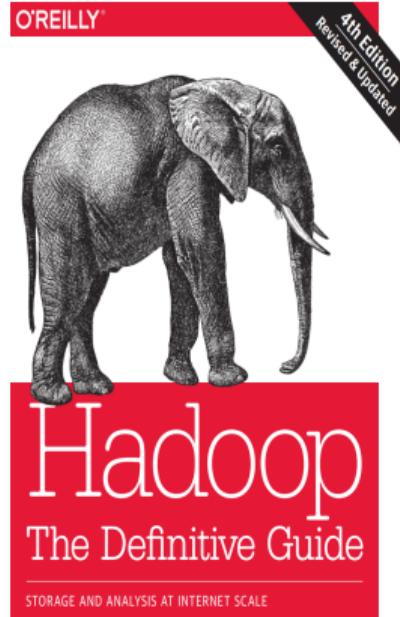
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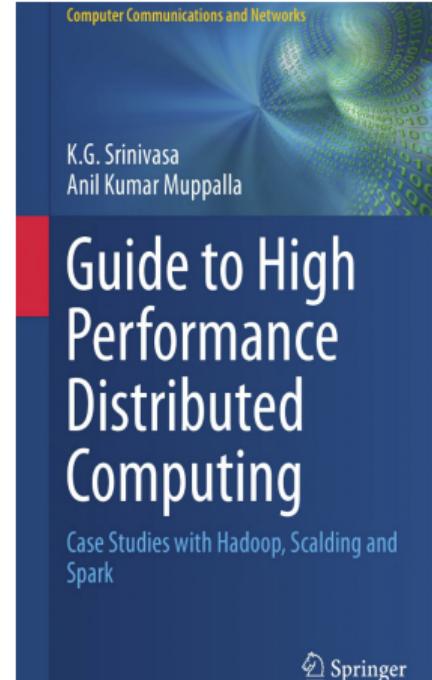
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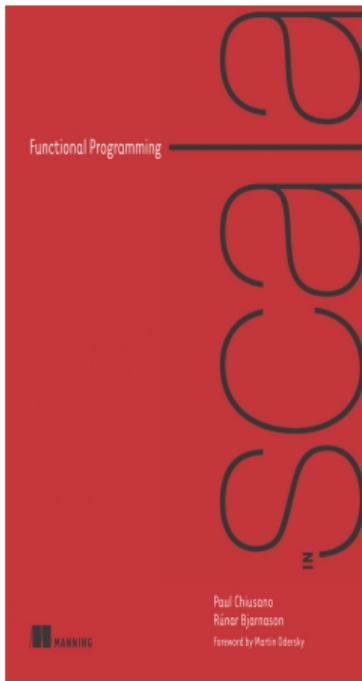
# Textbooks-1



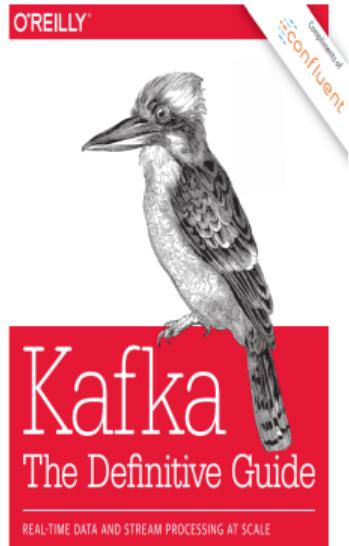
Tom White



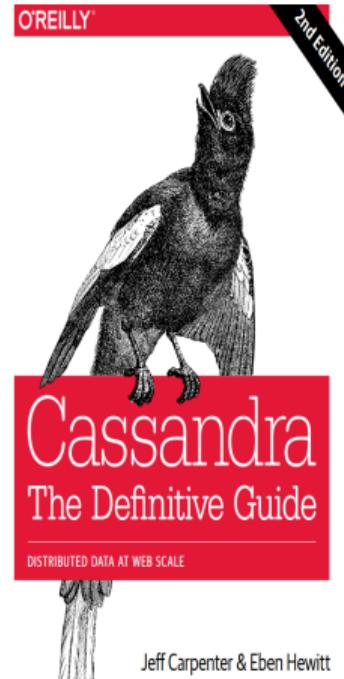
# Textbooks-2



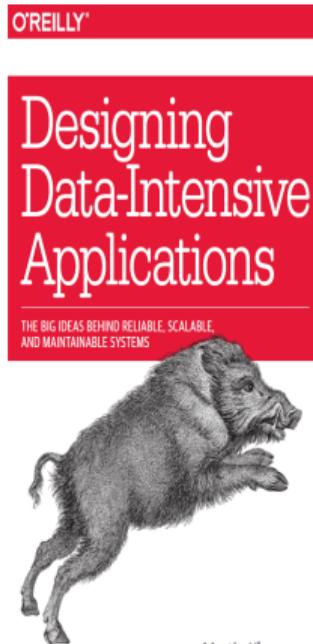
# Textbooks-3



Neha Narkhede,  
Gwen Shapira & Todd Palino



Jeff Carpenter & Eben Hewitt



Martin Kleppmann



# Ugly but important

- User stories or technical discussions are not related to any of my current work or my previous companies.
- I am working at EPAM Systems. My company approved me for doing this online course public but the materials are not reviewed or assessed by my company. It is on my own responsibilities.

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  - Getting max benefit from this course
  - Assignments and Labs
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  - Data Management
  - Data Abstraction
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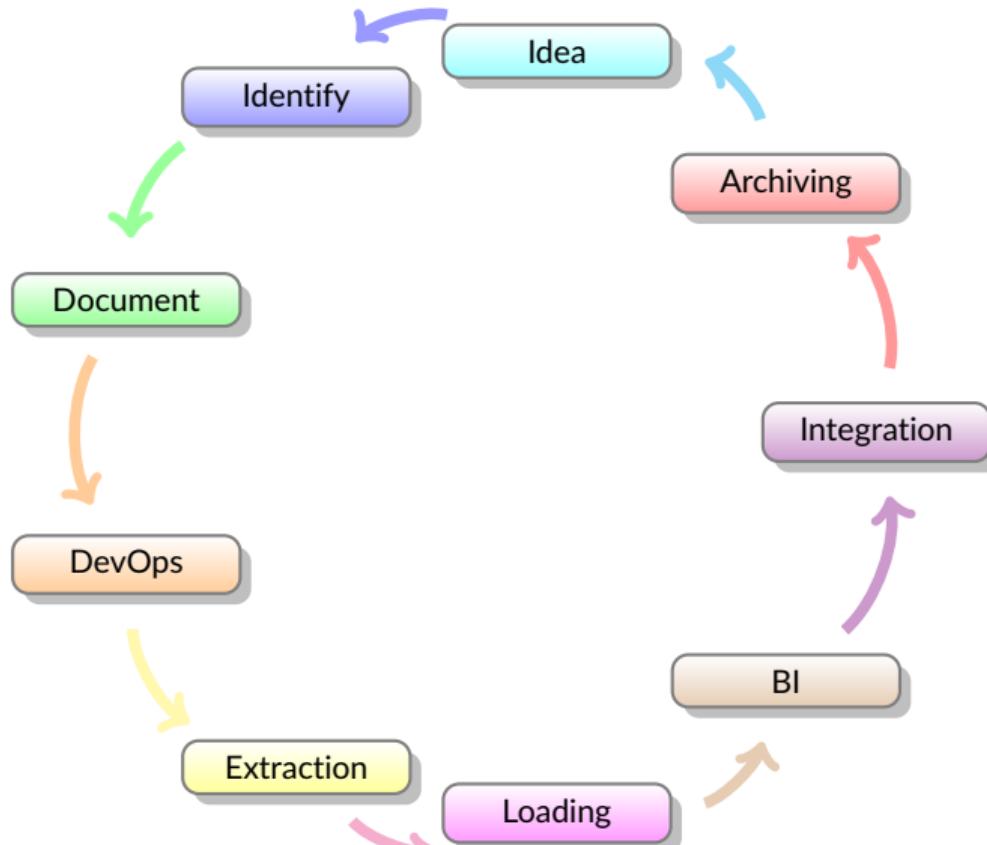
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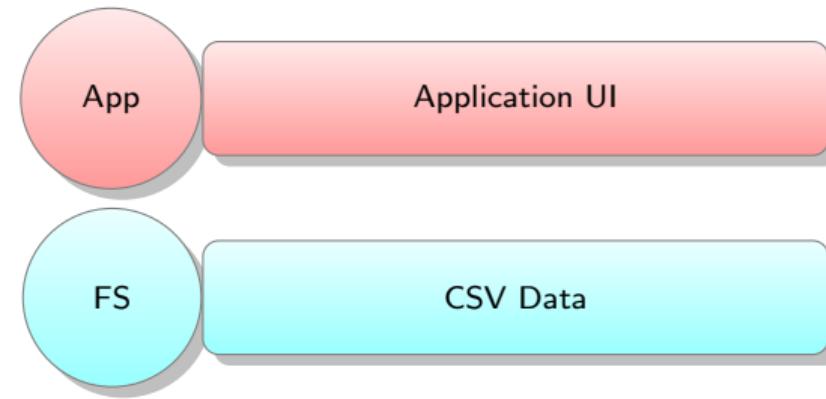
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# Data Management Life-Cycle



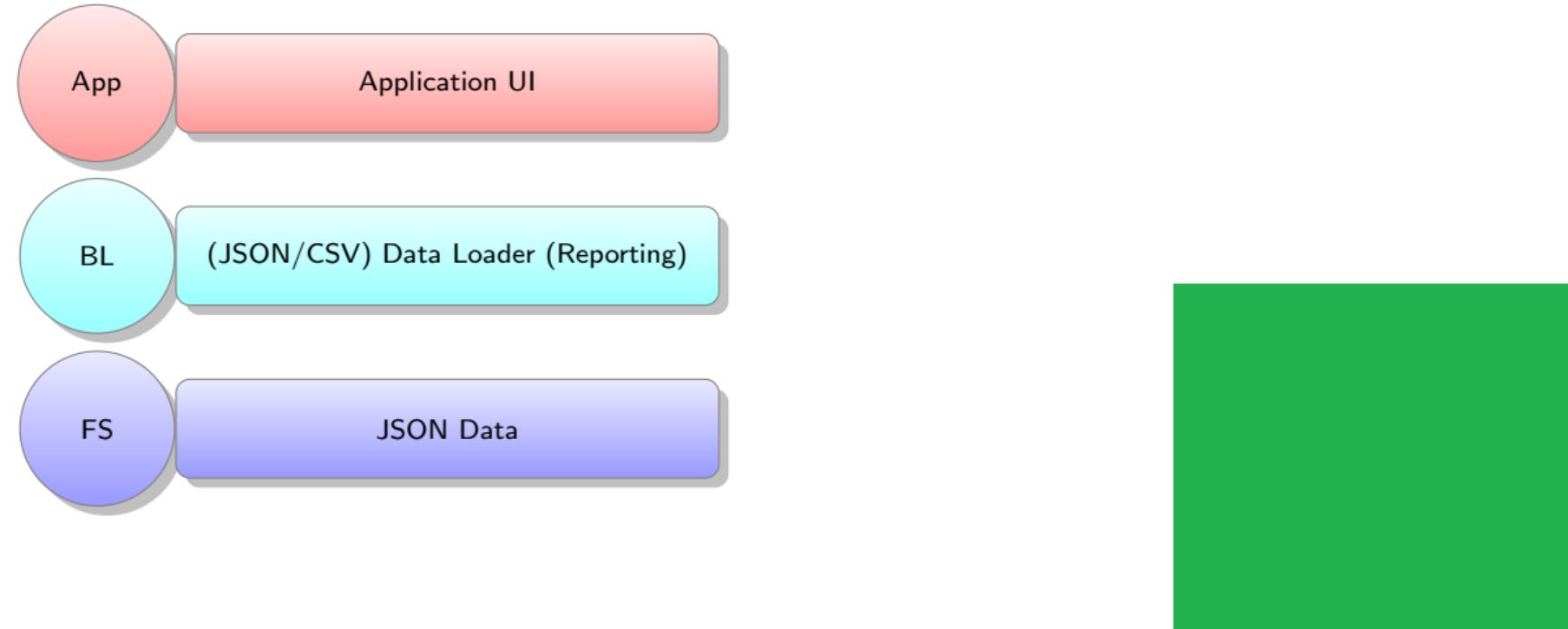
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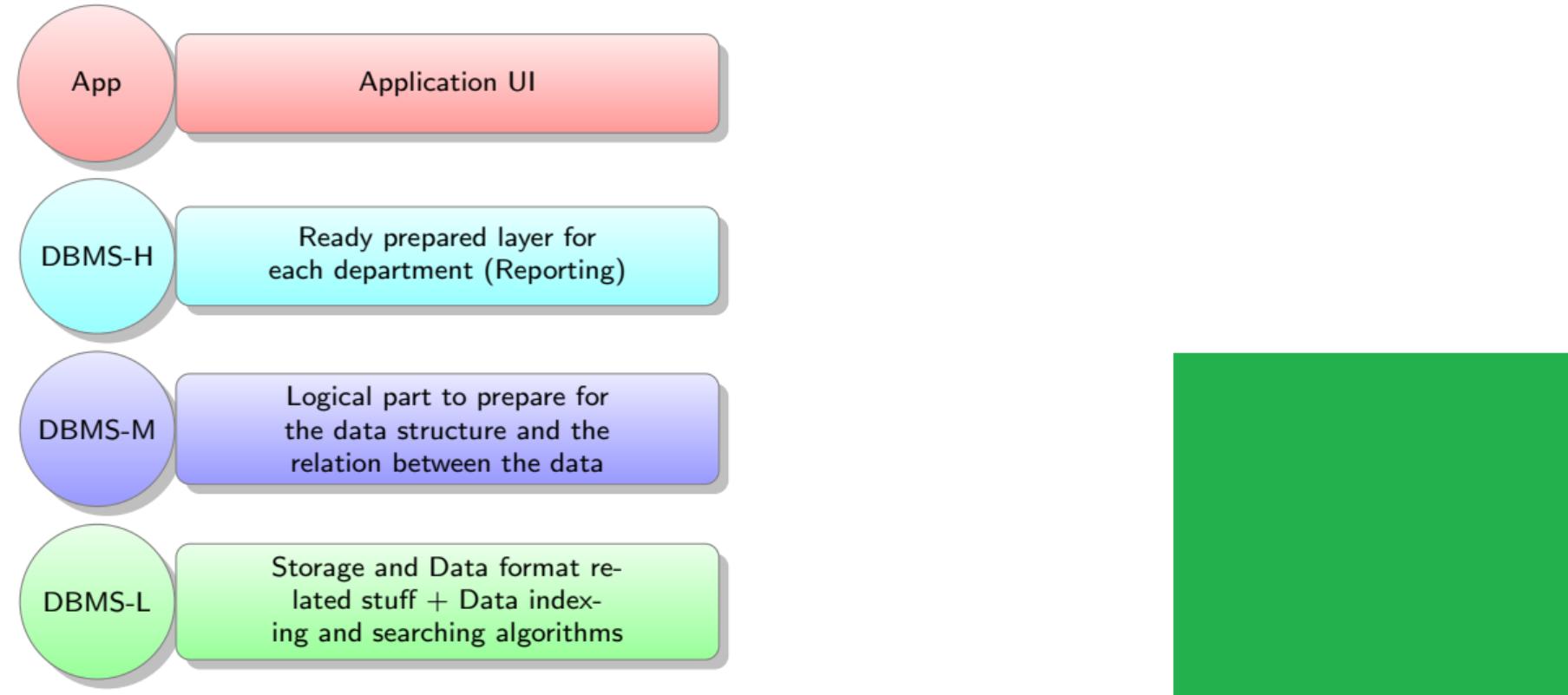
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- To answer these questions you need to understand the data layers.

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- The process of **hiding** irrelevant details from developer (user) is called data **abstraction**.

# Data Layers (Abstraction)

## 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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  - Physical Level
  - Logical/ Conceptual Level.

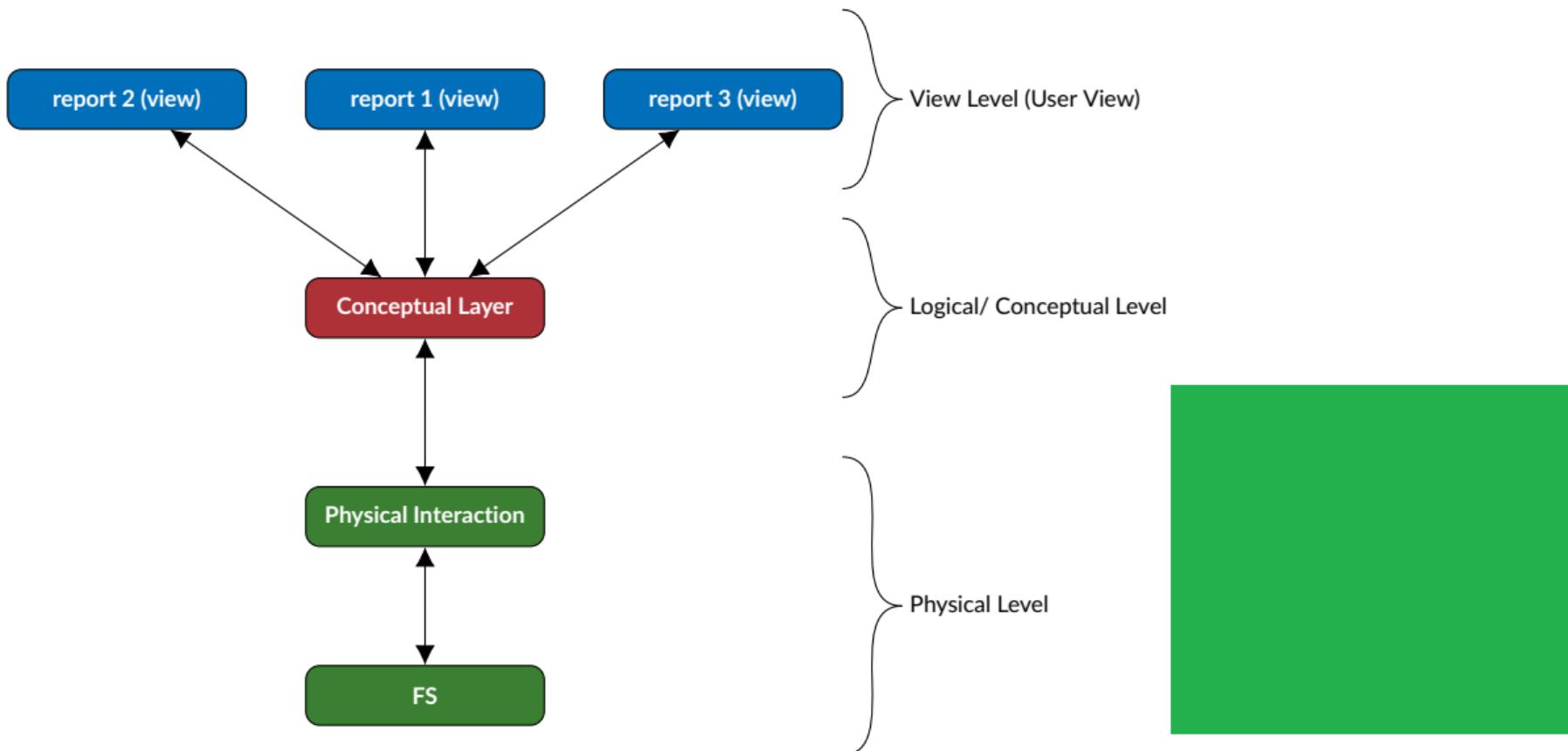
# Data Layers (Abstraction)

## 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.

- There are 3 levels of data abstraction.
  - Physical Level
  - Logical/ Conceptual Level.
  - View Level.

# Data Layers (Abstraction)



# Physical level

- Physical level (Internal):



# Physical level

- **Physical level (Internal):**
  - Lowest level.



# Physical level

- **Physical level (Internal):**
  - Lowest level.
  - Describes how data is stored.



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  - Change the file type or use different storage structure.
  - Changing the access method.
  - Modifying indexes.
  - Change the compression algorithm or hashing technique.

# Physical level

## Example

- Database contains product information.

# Physical level

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  - The amount of memory used.
  - Usually this layer abstracted from the programmers.

# Logical level

- **Logical level (Conceptual):**



# Logical level

- **Logical level (Conceptual):**
  - Intermediate level.



# Logical level

- **Logical level (Conceptual):**
  - Intermediate level.
  - Describes what data is stored.



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

# View level

- View level (External):



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- **View level (External):**

- Highest level.
- View of the data stored?
- Designed for category of users needs.
- It is the final interface for the user.
- It could be extended or hidden based on user's role.
- Not all the views are extended to all users and there is an authentication based on the category.

## Example

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## Example

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## Example

- Database contains product information.
- 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.

## Data solution thinking (Summary)

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  - 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.

# Motivation to Data Warehouse (DWH)

## 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.

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- The DWH is the core of the BI system which is built for data analysis and reporting.

# Motivation to Data Warehouse

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.

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

# DWH vs Operational databases

Metric	Transactions DB	DWH
Volume	GB/TB	TB/PB
Historical rows	Short-term ;1000M	Long-Term 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	Centralized storage (360°)

# Motivation to Data Warehouse

## 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

# Transnational DB Use cases



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# DWH Use cases



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# DWH Use cases



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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 DWH 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.

## Use case (DWH)

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?



# Use case (ODS)

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



## Use case (ODS)

**XTec** has a call center system which handles the customer inquiries. This system requires 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 inquiries.

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# Use case (ODS)

- 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.
  - Its 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).

## Use case (ODS)

- ODS is based on change data capture (CDC). This approach used to determine the data change and apply action based on this change.



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

# DWH Architecture Overview

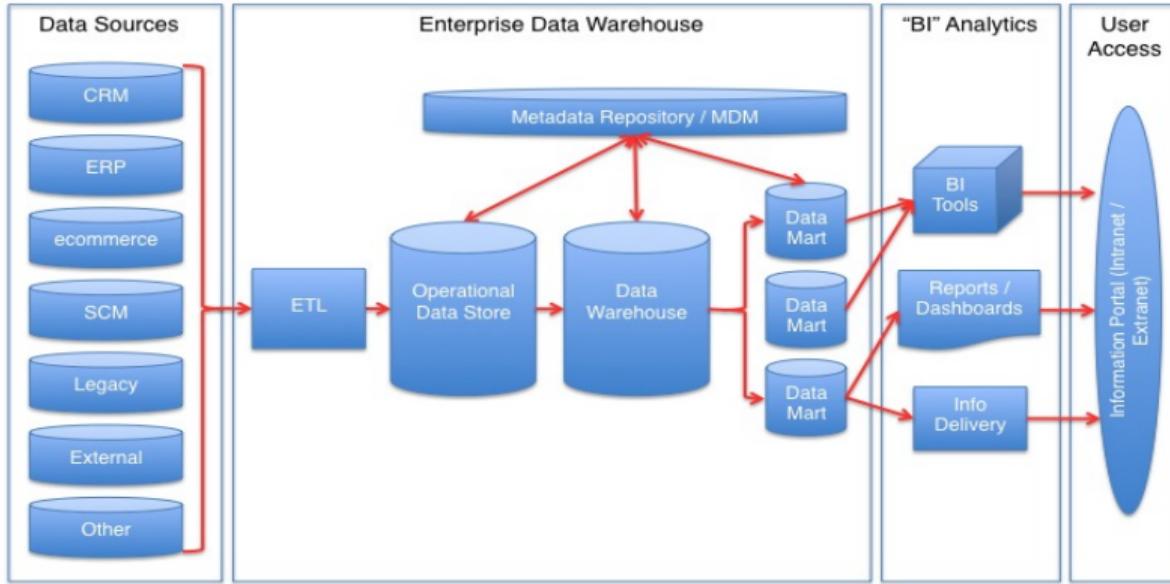


Figure: taken from XXXX

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- It describes how application (report) API data manipulation.
- 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.

# What is data model?

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 vs Implementation

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- What do we do for the implementation?
  - Hire a contractor to build (implement the design) the home.

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- You need to build a home. So, how do we design this home?
  - Determine if the home is one level or multi-level and decide man bedrooms and bathrooms for each floor. (User needs)
  - Hire an architect to put the architecture in more detailed way for example, the size for each room, the distribution of the wireds, where the plumbing fixtures will be placed, etc. (Architecture phase)
  - Decide the decorations, colors for each room, carpets, etc.
- What do we do for the implementation?
  - 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 Characteristics

some details about hot vs cold storage,



# Cold storage vs Hot storage

some details about hot vs cold storage,



# Data Models

- Any Big Data solution working based distributed systems.



# Data Models

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



# Introduction To Distributed Systems

# Chapter Objectives

- Understand the distributed systems concepts.



# Chapter Objectives

- Understand the distributed systems concepts.
- Replication and its usage in distributed systems.



# Chapter Objectives

- Understand the distributed systems concepts.
- Replication and its usage in distributed systems.
- Partitioning and its usage in distributed systems .

# Distributed Systems Concepts

- Any Big Data solution working based distributed systems.



# Distributed Systems Concepts

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- What is distributed systems in brief?



# Distributed Systems Architecture

- Any Big Data solution working based distributed systems.



# Distributed Systems Architecture

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- What is distributed systems in brief?



# Distributed Systems Challenges

- Any Big Data solution working based distributed systems.



# Distributed Systems Challenges

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# Design Simple Distributed System

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# Design Simple Distributed System

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# Introduction to Hadoop and Map-Reduce

# Chapter Objectives

- Introduction to Hadoop and its echo-systems.



# Chapter Objectives

- Introduction to Hadoop and its echo-systems.
- Why we need Hadoop?



# Chapter Objectives

- Introduction to Hadoop and its echo-systems.
- Why we need Hadoop?
- Understand the concept of HDFS and Map-Reduce.



# Chapter Objectives

- Introduction to Hadoop and its echo-systems.
- Why we need Hadoop?
- Understand the concept of HDFS and Map-Reduce.
- Developing Map-Reduce applications.

# Chapter Objectives

- Introduction to Hadoop and its echo-systems.
- Why we need Hadoop?
- Understand the concept of HDFS and Map-Reduce.
- Developing Map-Reduce applications.
- Using HiveQL over Map-Reduce.

# Chapter Objectives

- Introduction to Hadoop and its echo-systems.
- Why we need Hadoop?
- Understand the concept of HDFS and Map-Reduce.
- Developing Map-Reduce applications.
- Using HiveQL over Map-Reduce.
- Hadoop advantages and disadvantages with use cases?

# Hadoop Architecture

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# Hadoop Architecture

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- What is distributed systems in brief?



# Storage

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# Storage

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# Hadoop I/O

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# Hadoop I/O

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# Processing

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# Map-Reduce

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# Map-Reduce

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# Map-Reduce Components

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# Map-Reduce Components

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# Word-Count Example

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# Word-Count Example

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# ZooKeeper

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# Functional Programming

# Spark Framework



# Spark Framework: Spark Philosophy towards the Engine and the Programming languages

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# Spark Framework: Spark Philosophy towards the Engine and the Programming languages

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# Spark Framework: Spark Basics

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# Spark Basics

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# Spark Programming using RDDs

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# Spark Datasets/Dataframe

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# Spark on Production

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# Spark For Batch Processing

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# Spark For Batch Processing

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# Building custom input and output connector using Spark

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# Building custom input and output connector using Spark

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# Spark Streaming

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# Spark using other Programming Languages

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# Spark For Data Scientist

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# Spark Graph Dataframe/Graphx

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# Spark Graph Dataframe/Graphx

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# Spark Graph Dataframe/Graphx

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# Tuning your Spark Jobs

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## Real World Applications

# Massaging Systems

# Data Orchestration

# NOSQL

# Elastic



# Data Architecture Design

# Appendix

# Appendix A- Shell Programming

- Any Big Data solution working based distributed systems.



# Appendix A- Shell Programming

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## Appendix B- Java Programming

- Any Big Data solution working based distributed systems.



## Appendix B- Java Programming

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## Appendix C- Scala Programming

- Any Big Data solution working based distributed systems.



## Appendix C- Scala Programming

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## Appendix D- SQL Programming

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## Appendix D- SQL Programming

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## Appendix E- Oozie Orchestration

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## Appendix E- Oozie Orchestration

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## Appendix F- DWH Concepts and Data Modeling Design

- Any Big Data solution working based distributed systems.



## Appendix F- DWH Concepts and Data Modeling Design

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## Appendix G- Machine Learning Concepts Data Engineers

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## Appendix H- Docker for Data Engineers

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