

# **BIG DATA TECHNOLOGIES**

# ULTIMATE BIG DATA TECHNOLOGIES ROADMAP – THE BEST EVER!

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**Goal: Learn to process, analyze, and manage massive datasets using modern Big Data technologies. Develop expertise in distributed systems, batch/stream processing, and scalable data storage.**

## Why This Roadmap?

- **From Basics to Advanced** – Covers fundamental concepts, hands-on implementation, and real-world projects.
- **Master Distributed Data Processing Frameworks** – Spark, Hadoop, Flink, and more.
- **Learn Scalable Storage and NoSQL Databases** – HDFS, MongoDB, Cassandra, etc.
- **Work on Real-Time Stream Processing** – Kafka, Apache Storm, and others.

## Phase 1: Understanding Big Data Fundamentals

**Goal:** Build a solid foundation in Big Data and Distributed Systems.

### 1. Introduction to Big Data Concepts

**Topics to Cover:**

- What is Big Data?
- Characteristics of Big Data (Volume, Velocity, Variety, Veracity, and Value)
- Data Generation Sources: IoT, Social Media, Sensors, etc.
- Traditional Databases vs. Big Data

**References:**

- [Introduction to Big Data – Coursera](#)
- [Big Data Fundamentals – Udemy](#)
- [Big Data Explained – YouTube](#)

**Hands-On Task:**

- Analyze basic datasets using Python (Pandas & NumPy).
- Explore JSON, CSV, and XML data formats.

### 2. Big Data Ecosystem Overview

**Topics to Cover:**

- Hadoop Ecosystem Overview

- Batch vs. Stream Processing
- Lambda and Kappa Architectures
- Role of ETL, Data Warehouses, and Data Lakes

**References:**

- [Hadoop Ecosystem Explained – Medium](#)
- [Lambda and Kappa Architecture – YouTube](#)

**Hands-On Task:**

- Create a data pipeline using batch and streaming data.
- Build a simple data lake with HDFS.

## **Phase 2: Distributed Storage Systems**

**Goal:** Learn how to store and manage large datasets efficiently.

### **3. Hadoop Distributed File System (HDFS)**

**Topics to Cover:**

- Basics of HDFS Architecture
- NameNode, DataNode, and Block Replication
- Read/Write Operations in HDFS
- HDFS Fault Tolerance and Scalability

**References:**

- [HDFS Overview – Cloudera](#)
- [HDFS Crash Course – YouTube](#)

**Hands-On Task:**

- Set up Hadoop on your local machine.
- Upload and retrieve data from HDFS.

### **4. NoSQL Databases and Key-Value Stores**

**Topics to Cover:**

- Types of NoSQL Databases (Key-Value, Column, Document, Graph)
- MongoDB, Cassandra, Redis, and HBase
- ACID vs. BASE Properties

**References:**

- [MongoDB Tutorial – MongoDB University](#)
- [Cassandra Basics – Datastax](#)

**Hands-On Task:**

- Implement a key-value store using Redis.
- Create and query documents in MongoDB.

## **Phase 3: Batch Processing Frameworks**

**Goal:** Master batch processing frameworks like Hadoop and Spark.

### **5. Apache Hadoop and MapReduce**

**Topics to Cover:**

- Hadoop Cluster Setup and Configuration
- MapReduce Programming Model
- Combiner and Partitioner Concepts
- YARN and Resource Management

**References:**

- [MapReduce Basics – Cloudera](#)
- [Hadoop Hands-On – YouTube](#)

**Hands-On Task:**

- Build a MapReduce job to count word frequencies.
- Execute MapReduce tasks on a Hadoop cluster.

### **6. Apache Spark for Distributed Data Processing**

**Topics to Cover:**

- Spark Architecture (RDDs, DAGs, and Executors)
- Spark vs. Hadoop – Key Differences
- PySpark, Spark SQL, and Spark Streaming
- Performance Tuning in Spark

**References:**

- [Apache Spark Basics – Databricks](#)
- [Spark with Python – YouTube](#)

**Hands-On Task:**

- Process large datasets using PySpark.
- Implement transformations and actions on RDDs.

## **Phase 4: Real-Time Stream Processing**

**Goal:** Master real-time data processing using Kafka, Flink, and Storm.

## 7. Apache Kafka for Real-Time Messaging

### Topics to Cover:

- Kafka Architecture (Broker, Producer, Consumer)
- Topic, Partition, and Offset Management
- Kafka Consumer Group and Load Balancing

### References:

- [Kafka Fundamentals – Confluent](#)
- [Kafka in Action – YouTube](#)

### Hands-On Task:

- Set up a Kafka cluster.
- Publish and consume messages in Kafka.

## 8. Apache Flink and Stream Processing

### Topics to Cover:

- Flink vs. Spark Streaming
- Flink Architecture (Job Manager, Task Manager)
- Windowing and Stateful Processing in Flink

### References:

- [Apache Flink Crash Course – YouTube](#)
- [Flink Documentation – Apache](#)

### Hands-On Task:

- Implement stream processing using Apache Flink.
- Perform windowed operations on streaming data.

## 9. Apache Storm for Real-Time Computation

### Topics to Cover:

- Storm Architecture (Spouts, Bolts, and Topologies)
- Fault Tolerance and Scalability in Storm
- Use Cases of Storm in Real-Time Applications

### References:

- [Storm Basics – Apache Storm](#)
- [Apache Storm Hands-On – YouTube](#)

### Hands-On Task:

- Build a real-time data pipeline using Apache Storm.
- Process real-time logs using Storm topologies.

## **Phase 5: Data Warehousing and ETL**

**Goal:** Learn data warehousing and ETL pipeline development.

### **10. Data Warehousing and ETL Concepts**

**Topics to Cover:**

- ETL (Extract, Transform, Load) Basics
- Data Warehousing Concepts (Star vs. Snowflake Schema)
- Apache Hive and Impala for Querying Big Data

**References:**

- [ETL Fundamentals – Coursera](#)
- [Apache Hive Basics – Cloudera](#)

**Hands-On Task:**

- Design and implement an ETL pipeline.
- Query large datasets using Apache Hive.

### **11. Apache NiFi for Data Flow Automation**

**Topics to Cover:**

- NiFi Architecture (Processor, FlowFile, and Connection)
- Building Data Flow Pipelines with NiFi
- Integrating NiFi with Kafka and Hadoop

**References:**

- [NiFi Documentation – Apache](#)
- [NiFi Crash Course – YouTube](#)

**Hands-On Task:**

- Automate data flow between Kafka and Hadoop using NiFi.
- Create ETL workflows with NiFi processors.

## **Phase 6: Machine Learning and Big Data Analytics**

**Goal:** Build machine learning models on large-scale datasets.

### **12. Machine Learning with Spark MLlib**

**Topics to Cover:**

- Overview of Spark MLlib and Pipelines
- Distributed Model Training and Hyperparameter Tuning
- Feature Engineering with Spark

**References:**

- [Spark MLlib Tutorial – Databricks](#)
- [MLlib Hands-On – YouTube](#)

**Hands-On Task:**

- Build and deploy machine learning models using Spark MLlib.
- Perform feature scaling and selection in distributed datasets.

### **13. Big Data Analytics with Presto and ClickHouse**

**Topics to Cover:**

- Presto and ClickHouse Architecture
- SQL-on-Hadoop Technologies
- Real-Time Analytics on Large Datasets

**References:**

- [Presto Documentation – PrestoSQL](#)
- [ClickHouse Basics – YouTube](#)

**Hands-On Task:**

- Query large datasets using Presto.
- Analyze clickstream data with ClickHouse.

## **Phase 7: Security and Governance in Big Data**

**Goal:** Learn security best practices and data governance in Big Data.

### **14. Data Security and Privacy in Big Data**

**Topics to Cover:**

- Data Encryption and Access Control
- Kerberos and Ranger for Hadoop Security
- Role-Based Access Control (RBAC)

**References:**

- [Hadoop Security – Cloudera](#)
- [Kerberos Authentication Basics – YouTube](#)

**Hands-On Task:**

- Configure Kerberos Authentication in Hadoop.
- Implement role-based security in Apache Ranger.

## **Phase 8: Capstone Projects & Portfolio Building**

**Goal:** Build and showcase industry-level Big Data projects.

## 15. Real-World Capstone Project Ideas

### Project Ideas:

- Real-Time Clickstream Analysis for E-Commerce
- Log Data Processing and Anomaly Detection with Kafka and Flink
- Building a Scalable ETL Pipeline with Apache NiFi
- Fraud Detection in Banking using Spark and MLlib
- IoT Sensor Data Processing with Kafka and HDFS

## Estimated Timeline to Master Big Data:

**Beginner to Intermediate:** 3-4 months

**Intermediate to Advanced:** 5-6 months

**Capstone and Deployment:** 2-3 months

## By Following This Roadmap, You Will:

Master Distributed Data Processing with Spark, Hadoop, and Flink.

Develop Real-Time Data Pipelines Using Kafka and NiFi.

Build and Deploy ML Models on Big Data Platforms.

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