

Here's a **detailed course content** for a training program on **Data Engineering, Pipeline Development,** and **ETL**:

## Module 1: Introduction to Data Engineering

## 1. What is Data Engineering?

- o Definition and Role of a Data Engineer
- o Difference Between Data Engineering and Data Science
- o Data Engineering Workflow Overview

# 2. Data Engineering Tools and Technologies

- o Key Tools: Hadoop, Spark, Apache Kafka, Airflow
- Cloud-Based Data Engineering: AWS, GCP, Azure
- o Introduction to NoSQL and SQL Databases

## 3. Data Engineering Ecosystem

- o Data Lakes vs. Data Warehouses
- o Data Streams and Batch Processing
- o Real-Time vs. Batch Processing

# **Module 2: Data Pipeline Basics**

# 1. What is a Data Pipeline?

- Definition and Purpose of Data Pipelines
- Types of Data Pipelines: Batch, Streaming, Hybrid
- Components of a Data Pipeline: Data Collection, Transformation, Storage, and Analysis

# 2. Building Data Pipelines

- o Design Considerations: Scalability, Fault Tolerance, Latency
- Steps to Build a Simple Data Pipeline
- Best Practices for Data Pipeline Development

## 3. Data Flow Management and Scheduling

- o Data Ingestion Methods: APIs, File Systems, Databases
- Workflow Orchestration with Apache Airflow and Luigi
- o Scheduling and Monitoring Data Pipelines

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## Module 3: Extract, Transform, Load (ETL) Fundamentals

#### 1. What is ETL?

- o Definition and Importance of ETL in Data Engineering
- o ETL vs. ELT: Key Differences
- o Common ETL Tools: Apache NiFi, Talend, Informatica

# 2. Extracting Data (E)

- o Data Extraction Methods: APIs, Database Queries, File Transfers
- o Handling Different Data Formats: CSV, JSON, Parquet, Avro
- o Dealing with Structured and Unstructured Data

# 3. Transforming Data (T)

- o Data Cleaning: Handling Missing Values, Duplicates, and Outliers
- o Data Transformation Techniques: Aggregation, Normalization, Parsing
- o Using Python, Spark, and SQL for Data Transformation

## 4. Loading Data (L)

- o Loading Data into Data Warehouses, Databases, or Data Lakes
- o Batch vs. Real-Time Data Loading
- o Optimizing Data Loading for Performance

## **Module 4: Advanced ETL Concepts**

## 1. Data Quality and Validation

- o Importance of Data Quality in ETL Processes
- o Techniques for Data Validation and Integrity Checks
- o Implementing Error Handling and Data Cleansing

# 2. Incremental and Delta Loads

- o What is Incremental Loading?
- o Techniques for Handling Delta Loads
- Using Change Data Capture (CDC) in ETL Pipelines

# 3. Performance Optimization in ETL

- o Parallelism and Distributed Computing with Spark
- o Caching and Partitioning in ETL Pipelines
- o Optimizing SQL Queries and Data Transformations

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## 4. ETL Best Practices and Challenges

- Designing Robust ETL Pipelines for Scalability
- o Managing Dependencies and Scheduling
- o Handling Large-Scale Data Volume

# **Module 5: Data Pipeline Automation and Orchestration**

# 1. Workflow Automation with Apache Airflow

- o Introduction to Apache Airflow: DAGs and Tasks
- o Creating and Managing Data Pipelines in Airflow
- Handling Task Failures and Retries

# 2. Data Pipeline Orchestration

- o Managing Data Workflow Dependencies
- o Orchestrating Complex Data Pipelines with Airflow or Luigi
- o Automating Data Pipelines for Continuous Integration

## 3. Monitoring and Logging Pipelines

- Setting Up Logging and Alerts for Data Pipelines
- o Monitoring ETL Pipeline Performance
- Using Prometheus and Grafana for Real-Time Monitoring

## **Module 6: Cloud-Based Data Engineering**

## 1. Data Engineering in the Cloud

- o Advantages of Cloud-Based Data Pipelines
- o Overview of AWS, Azure, and Google Cloud Data Engineering Services
- o Cloud Storage Solutions: S3, GCS, Blob Storage

## 2. Managed Data Services

- o AWS Glue, Azure Data Factory, Google Cloud Dataflow
- Using Managed Services for ETL and Data Integration
- o Integrating Cloud Services with On-Premise Data Sources

## 3. Cloud Data Warehousing

- o Introduction to Data Warehouses: Snowflake, Redshift, BigQuery
- o Cloud-Based ETL Pipelines for Data Warehouses

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o Scaling Data Pipelines with Cloud Resources

# **Module 7: Real-Time Data Pipelines and Streaming**

## 1. Introduction to Real-Time Data Pipelines

- o Differences Between Batch and Streaming Pipelines
- o Use Cases for Real-Time Data Pipelines
- o Challenges of Real-Time Data Processing

# 2. Streaming with Apache Kafka

- o Overview of Apache Kafka Architecture
- o Setting Up Kafka Producers and Consumers
- o Real-Time Data Processing with Kafka Streams

## 3. Stream Processing Frameworks

- Using Apache Flink for Stream Processing
- Stream Processing with Spark Streaming
- o Building a Real-Time ETL Pipeline

## **Module 8: Data Pipeline Security and Governance**

## 1. Data Security in Pipelines

- Securing Sensitive Data in Transit and at Rest
- o Role-Based Access Control (RBAC) and Authentication
- o Encryption Techniques for Data in Pipelines

# 2. Data Governance and Compliance

- o Importance of Data Governance in Pipelines
- o Data Lineage and Metadata Management
- o Adhering to Data Privacy Regulations: GDPR, HIPAA

## 3. Monitoring Pipeline Security

- o Best Practices for Monitoring and Auditing Data Pipelines
- o Threat Detection and Response in ETL Processes

## **Module 9: Hands-On Projects and Case Studies**

## 1. Project 1: Building an ETL Pipeline Using Apache Airflow

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- 2. Project 2: Designing a Scalable Data Pipeline on AWS
- 3. Project 3: Real-Time Data Streaming with Apache Kafka
- 4. Case Study: Migrating ETL Processes to the Cloud
- 5. Capstone Project: End-to-End Data Pipeline Implementation

# **Module 10: Closing and Certification**

- 1. Final Assessment
- 2. Review and Feedback
- 3. Certification of Completion
- 4. Career Guidance and Further Learning Resources

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