

Cloud ML engineer

1.Introduction to Machine Learning

- 1.1 What is machine learning?
- 1.2Types of machine learning (supervised, unsupervised, reinforcement learning)
- 1.3 Machine learning workflow
- 1.4Data preprocessing and feature engineering

2. Python for Machine Learning

- 2.1 Python basics for ML engineers
- 2.2 Popular libraries
- 2.3 Jupyter Notebooks for experimentation

3.Data Collection and Cleaning

- 3.1 Data sources and acquisition
- 3.2 Data cleaning and preprocessing
- 3.3 Handling missing data
- 3.4 Data visualization for analysis

4. Model Building

- 4.1 Supervised learning algorithms
- 4.2 Unsupervised learning algorithms
- 4.3 Neural networks and deep learning
- 4.4 Model evaluation and selection









5.Cloud Computing Fundamentals

- 5.1 Introduction to cloud computing platforms
- 5.2 Setting up cloud accounts and access management
- 5.3 Overview of cloud services (EC2, S3, GCP, Azure ML, etc.)

6.Deploying ML Models on Cloud

- 6.1 Containerization (Docker)
- 6.2 Kubernetes orchestration
- 6.3 Model serving with cloud services (AWS SageMaker, GoogleAI Platform, Azure ML)
- 6.4 Serverless computing for model deployment

7. Scaling and Optimization

- 7.1 AutoML tools and techniques
- 7.2 Hyperparameter tuning
- 7.3 Model optimization and performance improvement
- 7.4 Monitoring and debugging in production

8.Data Pipelines and ETL

- 9.1 Building data pipelines on cloud platforms
- 9.2 Extract, Transform, Load (ETL) processes
- 9.3 Batch and stream processing
- 9.4 Data storage and warehousing

10.Big Data and ML

10.1 Integration of big data tools (Hadoop, Spark) with machinelearning









- 10.2 Distributed computing for large-scale data processing
- 10.3 Training models on large datasets

11.Advanced Topics

- Reinforcement learning and its applications
- 11.2 Natural language processing (NLP) for ML engineers
- 11.3 Computer vision and image recognition
- 11.4 Time series forecasting

12.Ethical and Responsible AI

- 12.1 Bias and fairness in machine learning
- 12.2 Privacy and data security considerations
- 12.3 Regulatory compliance and ethical guidelines

13. Capstone Project

- 13.1 Apply the knowledge and skills learned throughout the course to a real-world project
- 13.2 Develop, deploy, and optimize a machine learning model on acloud platform
- 13.3 Present the project to peers and instructors

14. Final Assessment and Certification

- 14.1 Final exam or assessment to evaluate your understanding of course material
- 14.2 Course completion certificate





