



AI COURSE PLAN FOR IT PROFESSIONALS (CODERS)

Weeks 1-2 Foundations of AI & Data in Business & IT

Topics:

- ☑ Overview of AI's applications in business and IT.
- Key roles in Business Analytics, Data Analysis, Quality Assurance, & IT Support.
- ☑ Data types and structures: structured, semi-structured, unstructured.
- ☑ Data storage and retrieval basics.
- ☑ Data quality fundamentals: accuracy, consistency, completeness.
- ☑ Using Excel and Power BI for data organization and visualization.
- Creating basic charts and dashboards in Tableau.
- ☑ Understanding non-coding data handling techniques.
- 🗹 Organizing data for analysis.
- ☑ Introduction to using data insights for business decisions.

Tools:

- ☑ Microsoft Excel: Data organization, basic analysis.
- Power BI: Data visualization and dashboard creation.
- ☑ Tableau: Interactive dashboards and data exploration.

Practical Exercises:

- ☑ Structuring and analyzing data in Excel.
- ☑ Building a dashboard in Power BI or Tableau.

Resources:

Harvard Online - Data Science for Business: Data Science for Business

☑ YouTube:

Google Cloud AI in Business

Power BI Tutorial for Beginners

🗹 Tableau Free Training: Tableau Learning

Weeks 3-4 Python Fundamentals for Data Analysis

Topics:

- Python basics: variables, data types, operators.
- ☑ Data structures: lists, dictionaries, tuples.
- ☑ Control flow: if-else, loops.
- ☑ Functions and modules in Python.
- ☑ Introduction to Pandas and NumPy for data handling.
- ☑ Loading and handling data from files (CSV, Excel).
- ☑ Using Jupyter Notebook for data analysis.
- ☑ Google Colab for cloud-based coding.
- Data manipulation basics with Pandas.
- ☑ Simple data transformations and summarizations.

Tools:

- ☑ Jupyter Notebook: Interactive coding environment.
- ☑ Google Colab: Cloud-based Python platform.
- ☑ VS Code: Python development environment.

Practical Exercises:

- ☑ Writing basic Python scripts for data manipulation.
- Exploring data structures and control flow in Jupyter Notebook.

- ☑ YouTube:
- Corey Schafer's Python Tutorials Jupyter Notebook Basics by Tech with Tim
- Real Python Pandas Guide: Real Python
- Google Colab Tutorials: Google Colab

Weeks 5-6 Data Cleaning & Transformation Techniques

Topics:

- ☑ Data cleaning techniques for accuracy and completeness.
- ☑ Handling missing data (imputation, deletion).
- Removing duplicates and outliers.
- ☑ Data normalization and standardization.
- ☑ Converting data types (e.g., strings to numbers, dates).
- Incoding categorical variables (one-hot, label encoding).
- ☑ Text data cleaning with regular expressions.
- ☑ Filtering and sorting data in Pandas.
- ☑ Aggregating data with groupby functions.
- ☑ Introduction to OpenRefine for advanced cleaning.

Tools:

- ☑ Pandas: Data cleaning and manipulation.
- ☑ OpenRefine: Advanced cleaning for complex datasets.

Practical Exercises:

- Cleaning and transforming datasets using Pandas.
- Performing advanced data cleaning with OpenRefine.

- 🗹 Kaggle Data Cleaning Course: Kaggle
- ☑ YouTube:
 - Data Cleaning with Pandas by Data School OpenRefine Basics

Weeks 7-8 Data Visualization Basics

Topics:

- Data visualization principles and chart types.
- Creating bar charts, line charts, histograms in Matplotlib.
- Generating scatter plots and box plots.
- ☑ Advanced statistical visualizations with Seaborn.
- Choosing appropriate chart types for data.
- ☑ Customizing visualizations (labels, titles, legends).
- ☑ Working with subplots for multi-chart visualization.
- ☑ Visualizing data relationships and distributions.
- ☑ Basics of color theory in data visualization.
- ☑ Exporting visualizations to files (PNG, PDF).

Tools:

- Matplotlib: Fundamental plotting.
- Seaborn: Advanced visualizations.
- ☑ Plotly: Interactive charts.

Practical Exercises:

- ☑ Creating and customizing visualizations in Matplotlib and Seaborn.
- ☑ Building an interactive chart with Plotly.

Resources:

☑ YouTube:

StatQuest Data Visualization

Data Visualization with Matplotlib by Sentdex

DataCamp - Data Visualization with Python: DataCamp

Weeks 9-10 Introduction to Cloud Computing for AI and Data

Topics:

- ☑ Basics of cloud computing: IaaS, PaaS, SaaS models.
- ☑ Benefits of cloud computing for AI and data science.
- ☑ Key cloud providers: AWS, Azure, Google Cloud Platform (GCP).
- ☑ Understanding virtual machines and containers.
- ☑ Introduction to serverless computing and its use in AI.
- ☑ Overview of compute services in the cloud (EC2, Azure Vms, GCP Compute Engine).
- ☑ Overview of storage services (S3, Azure Blob, GCP Storage).
- Basics of cloud networking and security.
- Cloud data storage solutions: databases, data lakes.
- ☑ Introduction to cloud-based development environments.

Tools:

☑ AWS, Azure, GCP Free Tiers: Cloud services exploration.

☑ Docker: Containerization.

Practical Exercises:

- ☑ Setting up a VM on AWS or Azure.
- Deploying a Docker container for AI models.

Resources:

AWS Training - Cloud Practitioner Essentials: AWS Training

☑ YouTube:

Intro to Cloud Computing - Google Cloud

Microsoft Azure Fundamentals

🗹 Docker Get Started Guide: Docker

Weeks 11–12 AI/ML Services on Azure, AWS, & GCP

Topics:

- ☑ Introduction to Azure Cognitive Services (text, vision, speech).
- AWS SageMaker for model training and deployment.
- GCP Vertex AI overview for AI/ML workflows.
- ☑ Text analytics and language processing with Azure.
- ☑ Image recognition with AWS Rekognition.
- AutoML tools in cloud platforms for automated model building.
- ☑ Comparing AI/ML service offerings across cloud platforms.
- ☑ Overview of deployment options (model endpoints, batch predictions).
- ☑ Data storage and retrieval for ML on each platform.
- ☑ Using cloud-based notebooks (SageMaker, Azure Notebooks).

Tools:

- ☑ Azure Cognitive Services: Text, vision, and language processing.
- ☑ AWS SageMaker: ML lifecycle management.
- ☑ GCP Vertex AI: Model building and deployment.

Practical Exercises:

- ☑ Deploying an AI model on AWS SageMaker.
- ☑ Using Azure's Text Analytics API for sentiment analysis.

- AWS Machine Learning University: AWS ML University
- ☑ YouTube:
 - Azure Cognitive Services Overview
 - GCP's Vertex Al Tutorial

Weeks 13-14

Data Science Workflow & Advanced Data Wrangling

Topics:

- ☑ Overview of the data science workflow.
- ☑ Data gathering techniques and sources (APIs, scraping).
- ☑ Data cleaning in a structured workflow.
- Exploratory data analysis and visualization for insights.
- Advanced data merging and joining techniques.
- ☑ Working with timestamps and datetime objects.
- ☑ Filtering and slicing data in Pandas.
- Data transformation with pivot and melt functions.
- Exporting clean data for model building.
- Documenting data workflows for collaboration.

Tools:

- Pandas: Advanced data manipulation.
- ☑ BeautifulSoup/Requests: Web scraping.
- ☑ SQL: Querying databases.

Practical Exercises:

- Combining data from multiple sources with Pandas.
- ☑ Writing SQL queries to clean and organize data.

Resources:

DataCamp - Data Manipulation with Pandas: DataCamp

- ☑ YouTube:
 - Advanced Pandas Techniques by Keith Galli Corey Schafer - Web Scraping

Weeks 15-16

Introduction to Large Language Models (LLMs) & Prompt Engineering

Topics:

- ☑ Fundamentals of Transformer models (GPT, BERT).
- ☑ Understanding tokenization and embeddings in LLMs.
- Applications of LLMs in NLP tasks (text generation, Q&A, summarization).
- ☑ Introduction to APIs for LLMs (OpenAI, Hugging Face).
- Basics of prompt engineering: crafting effective prompts.
- ☑ Strategies for prompt optimization (clarity, brevity).
- ☑ Introduction to fine-tuning LLMs for specific tasks.
- ☑ Understanding context and coherence in LLM responses.
- Limitations and ethical considerations of LLMs.
- ☑ Using LLMs for summarization, translation, and Q&A.

Tools:

- ☑ OpenAl API: Accessing GPT models.
- ☑ Hugging Face Transformers: Pre-trained models for NLP.
- ☑ GPT-3 Playground: Experimenting with prompts.

Practical Exercises:

- ☑ Crafting prompts to perform NLP tasks.
- ☑ Using Hugging Face to fine-tune a language model.

- OpenAI API Documentation: OpenAI
- ☑ YouTube:
 - Two Minute Papers GPT-3 Overview
 - Hugging Face Transformers Course
- ☑ Hugging Face Transformers Course: Hugging Face

Weeks 17-18

Exploratory Data Analysis (EDA) & Feature Engineering

Topics:

- Importance of exploratory data analysis in data science.
- Analyzing data distributions and patterns.
- ☑ Detecting and handling outliers.
- ☑ Creating custom features (feature engineering).
- ☑ Feature scaling techniques (normalization, standardization).
- ☑ Encoding categorical variables.
- ☑ Feature selection methods.
- ☑ Dimensionality reduction techniques (PCA, t-SNE).
- ☑ EDA with visualizations for insights.
- ☑ Reporting EDA findings effectively.

Tools:

- ☑ Pandas Profiling: Automated EDA reports.
- ☑ Scikit-Learn: Feature engineering tools.
- ☑ Plotly: Visualizing feature distributions.

Practical Exercises:

- Performing EDA on a dataset.
- ☑ Feature engineering on a dataset.

Resources:

Kaggle - Exploratory Data Analysis: Kaggle

☑ YouTube:

StatQuest - EDA

Data School - Feature Engineering

Weeks 19-20 Supervised Learning Techniques

Topics:

- ☑ Introduction to supervised learning: labeled data and predictions.
- ☑ Linear regression and logistic regression.
- ☑ Support Vector Machines (SVM).
- ☑ Decision Trees and Random Forests.
- ☑ Model evaluation metrics (accuracy, precision, recall, F1).
- ☑ Cross-validation techniques and train-test split.
- Avoiding overfitting and underfitting.
- ☑ Regularization techniques (L1, L2).
- ☑ Hyperparameter tuning with grid search.
- ☑ Model interpretability basics.

Tools:

- ☑ Scikit-Learn: Supervised ML algorithms.
- ☑ MLFlow: Experiment tracking.
- ☑ SHAP: Model interpretability.

Practical Exercises:

- ☑ Training and evaluating a supervised model.
- ☑ Experiment tracking with MLFlow.

Resources:

- ☑ Coursera Supervised ML by Andrew Ng: Coursera
- ☑ YouTube:

StatQuest - Supervised Learning

Scikit-Learn Documentation: Scikit-Learn

Weeks 21-22

Unsupervised Learning & Dimensionality Reduction

Topics:

- ☑ Introduction to unsupervised learning (no labeled data).
- ☑ Clustering techniques (K-means, hierarchical).
- ☑ Association rule learning for pattern recognition.
- ☑ Dimensionality reduction methods (PCA, t-SNE).
- ☑ Visualization of high-dimensional data.
- ☑ Evaluating clustering quality (silhouette score).
- Feature extraction for complex datasets.
- ☑Anomaly detection with clustering.
- Comparing clustering algorithms.
- Applications of unsupervised learning in business.

Tools:

- ☑ Scikit-Learn: Unsupervised learning and dimensionality reduction.
- ☑ TSNEViewer: Visualizing t-SNE results.
- ☑ ELKI: Advanced clustering and outlier detection.

Practical Exercises:

- ☑ Applying K-means clustering on a dataset.
- Reducing dimensions with PCA and visualizing results.

- 🗹 Fast.ai Practical ML Course: Fast.ai
- ☑ YouTube:
 - StatQuest Clustering and PCA 3Blue1Brown - PCA Visual Guide

Weeks 23-24 Model Deployment & Evaluation in the Cloud

Topics:

- ☑ Basics of model deployment for production.
- ☑ Building REST APIs for ML models.
- ☑ Introduction to Flask and FastAPI for model deployment.
- ☑ Monitoring model performance over time.
- ☑ Setting up automated model retraining.
- ☑ Model performance metrics in production (latency, accuracy).
- Security best practices for deployed models.
- \blacksquare Introduction to Prometheus for monitoring.
- Using Grafana for visualization.
- ☑ Scaling models in production with cloud solutions.

Tools:

- ☑ Flask/FastAPI: Deploying APIs for models.
- ☑ Prometheus and Grafana: Monitoring tools.
- Docker: Containerizing models for deployment.

Practical Exercises:

- ☑ Deploying a model with Flask/FastAPI as a REST API.
- Setting up Prometheus and Grafana to monitor a deployed model.

- ☑ AWS SageMaker Model Deployment: AWS
- ☑ YouTube:
 - Flask API for ML Model Deployment Prometheus and Grafana Monitoring

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- ☑ YouTube:
 - Flask API for ML Model Deployment Prometheus and Grafana Monitoring

Practical Exercises/Labs

Design an Advanced Prompt Engineering Application Using an LLM.

Task: Develop a sophisticated application leveraging an LLM, focusing on dynamic prompt engineering.

Objective: Apply advanced prompt strategies like multi-step prompting and chain-of-thought for tasks such as complex Q&A, summarization with context, or content generation.

Expected Output: Deploy a working model with a user interface (UI) or command-line interface (CLI) that can handle diverse input scenarios using crafted prompts.

2.Complete a Capstone Project Involving Data Processing, Modeling, Deployment, and Visualization.

Task: Create a full-stack AI application, including data ingestion, processing, ML modeling, deployment, and visualization.

Objective: Demonstrate a comprehensive understanding of the entire data science workflow, from data engineering to final user-facing analytics.

Expected Output: A deployed model accessible via an API or web application, supported by a visualization dashboard (e.g., Power BI, Tableau) to present insights.

3.Final Presentation of Capstone Project with Peer Review

Task: Present the capstone project in a structured format to peers and instructors, demonstrating project goals, methodology, results, and challenges encountered.

Objective: Communicate technical details effectively and receive constructive feedback for project enhancement.

Expected Output: A slide deck presentation or live demo showcasing project workflows, final outputs, and potential real-world impact.

Tools/Resources 1.Platforms

Azure: Use for LLM APIs, Azure Cognitive Services, and deployment on Azure Kubernetes Service (AKS).

AWS: Leverage SageMaker for model training, deployment on AWS Lambda or SageMaker Endpoints, and other tools like Comprehend or Rekognition.

GCP: Utilize Vertex AI for model training, AutoML, and BigQuery for data processing, plus Cloud Run for deployment.

2.Learning Materials

Advanced Prompt Engineering Papers: Research papers and tutorials on advanced techniques like context-aware prompting, prompt chaining, and multi-part prompts.

Project Management Documentation: Tutorials on Agile methodologies and project management resources (e.g., Trello, Asana) for effective capstone project planning and execution.

Additional Resources:

Tutorials and official docs on Docker/Kubernetes for model deployment.

Articles on integrating AI services within end-user applications (REST APIs, CI/CD).

Assessment

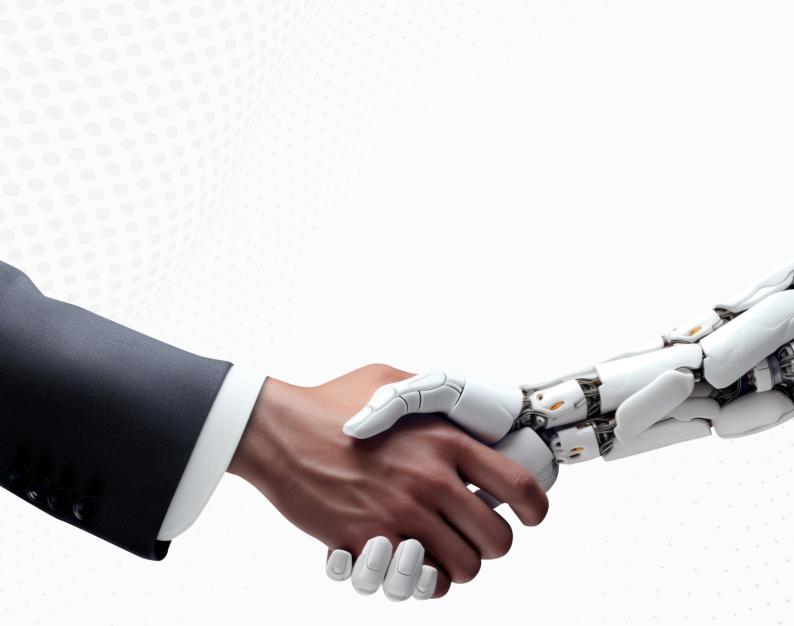
2.Capstone Project Evaluation Criteria:

Design and Architecture: Complexity and scalability of the project's architecture.

Implementation: Effectiveness of data processing, model accuracy, prompt engineering, and deployment.

Presentation: Clarity, technical depth, and alignment with project goals.

Feedback: Peer and instructor feedback focusing on areas of strength, potential improvements, and applications in real-world scenarios.



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