

Data Science & Machine Learning Project Roadmap (3 Months)

Month 1 – Fraud Detection System for Financial Transactions

• Problem Statement:

Financial institutions need to detect fraudulent transactions in real-time. Your goal is to build a robust fraud detection model using historical transactional data and machine learning classification techniques.

• Tech Stack:

Python, Pandas, NumPy, Scikit-learn, XGBoost, Matplotlib, Seaborn, Jupyter Notebook, SQL

• Development Timeline:

- Week 1:
 - Ingest transactional data (from Kaggle or simulated).
 - Perform EDA, identify patterns in fraud vs. non-fraud cases.
 - Address class imbalance using techniques like SMOTE or undersampling.

Week 2:

- Feature engineering: create time-based, amount-based, and behavior-based features.
- Train models (Logistic Regression, Decision Trees, XGBoost).
- Evaluate using F1 score, Precision-Recall, AUC-ROC.

Week 3:

- Hyperparameter tuning using GridSearchCV.
- Create a pipeline for real-time input testing.

Week 4:

- Final deployment-ready model.
- Create a dashboard (basic Streamlit) to simulate real-time prediction.
- Document findings, risks, and improvements.

Month 2 – AI-based Resume Screening System

• Problem Statement:

HR departments struggle to screen thousands of resumes for job roles. This project will develop a model that reads resume text and classifies them into suitable job categories.



• Tech Stack:

Python, NLTK, Spacy, TF-IDF, Scikit-learn, FastText / BERT (optional), Streamlit

• Development Timeline:

- o Week 1:
 - Gather text-based resume data.
 - Preprocess text (stop words, stemming, tokenization).
 - Apply TF-IDF to convert text to features.

Week 2:

- Use classification models like Naive Bayes, SVM, or Logistic Regression.
- Evaluate precision, recall, confusion matrix for job category classification.

Week 3:

- Upgrade model with pre-trained embeddings (FastText or BERT).
- Fine-tune embeddings on resume data if required.

Week 4:

- Create a resume uploader interface using Streamlit.
- Model reads resume and suggests the best-fit job roles.
- Create documentation and demo deck for HR demo.

Month 3 – Predictive Maintenance for Manufacturing Equipment

Problem Statement:

Predicting machine failure before it happens is crucial in manufacturing. This project aims to analyze sensor data from machines to predict which machines are likely to fail soon.

• Tech Stack:

Python, Pandas, NumPy, Seaborn, Matplotlib, Scikit-learn, Keras/TensorFlow, Jupyter Notebook

• Development Timeline:

- Week 1:
 - Load and explore time-series sensor data.
 - Handle missing values, normalize readings, and extract time-based features.

Week 2:

- EDA on failure vs. non-failure scenarios.
- Train classification model (Random Forest, XGBoost, or LSTM for time series).

Week 3:

- Use survival analysis or binary classification to predict Remaining Useful Life (RUL).
- Evaluate model performance using precision and recall.

Week 4:

- Deploy model with a web-based interface showing real-time predictions.
- Visualize with time-to-failure and alert mechanism.