




# Assignment 1: air quality Pollution Forecasting (Regression)

 **Difficulty:** Advanced |  **Time:** 4–5 hours

 **Tools:** Pandas, NumPy, Scikit-learn, Seaborn, Matplotlib, OOP, Streamlit or MLflow

 **Dataset:** [Air Quality - UCI Machine Learning Repository](#)

## Problem Statement

You are building a predictive model for an environmental agency that can forecast **air pollution concentration** based on weather data. The model will be used in a web application to inform citizens about air quality.

## Assignment Objectives

1. **Build an end-to-end regression pipeline** using object-oriented design.
2. Include **feature engineering, preprocessing, model training**, and **evaluation**.
3. Deploy the model using **Streamlit** or **track experiments** using **MLflow**.

## Task Breakdown

### *Task 1: Data Wrangling & Preprocessing*

- Implement a class `PollutionDataHandler` with:
  - `load_data()` to load the CSV
  - `clean_data()` to handle missing values
  - `engineer_features()` to:
    - Merge date/time columns into one `datetime` column
    - Extract `hour`, `month`, `day_of_week`
  - `scale_features()` for normalization

## Task 2: Model Training & Evaluation

- Create class `PollutionPredictor` with:
  - `train_model()`: Use `RandomForestRegressor` and Linear Regression
  - `evaluate_model()`: Print MAE, RMSE,  $R^2$
  - `cross_validate()` function for robust evaluation
- Visualizations:
  - Feature importance (barplot)
  - Residual plot

## Task 3: Pipeline & Deployment

Choose one of the following:



### ✅ Option A – Streamlit UI


- Allow users to input features via form sliders (temperature, wind, etc.)
- Output PM2.5 prediction
- Add charts: residuals, feature importance

### ✅ Option B – MLflow Integration

- Use MLflow for:
  - Logging hyperparameters
  - Logging metrics
  - Saving models
- Register the best model and use `mlflow.pyfunc.load_model()` for inference

## Assignment 2: Banknote Authentication (Classification)

 **Difficulty:** Advanced |  **Time:** 3–4 hours

 **Tools:** Pandas, NumPy, Scikit-learn, Seaborn, Matplotlib, OOP, MLflow or Streamlit

 **Dataset:** [Banknote Authentication - UCI Machine Learning Repository](#)

## Problem Statement

As a consultant for a security firm, you're tasked with building a classification model that detects **fake banknotes** using statistical image features. This model should be available either as a **web interface** or integrated in a **deployment pipeline**.

## Assignment Objectives

1. **Object-oriented classification system** using scikit-learn
2. Apply **EDA, feature selection**, and **model evaluation**
3. Use **Streamlit or MLflow** for deployment/tracking

## Task Breakdown

### *Task 1: Data Handling & Analysis*

- Build class BanknoteHandler:
  - `load_data()` to read and label columns
  - `preprocess()` to standardize data
  - `plot_distribution()` to visualize fake vs real banknotes

### *Task 2: Model Classifier & Evaluation*

- Build class BanknoteModel:
  - `train_model()` using Logistic Regression, RandomForest
  - `evaluate_model()` to print accuracy, F1-score, confusion matrix
  - `predict_note(features)` to predict a new sample
- Visualize:
  - Confusion matrix
  - ROC-AUC Curve
  - Feature importances

### ***Task 3: Full Pipeline & Deployment***

#### **✔ Option A – Streamlit**

- Create a form with sliders for input features (variance, skewness, etc.)
- Display prediction: Real or Fake
- Add performance charts

#### **✔ Option B – MLflow**

- Log:
  - Models
  - Metrics
  - Parameters
- Use `mlflow models serve` to expose the trained classifier as a REST API