

# Part 1: Multilinear Regression & Polynomial Regression

**Dataset:** Student Performance Dataset:

<https://www.kaggle.com/datasets/spscientist/students-performance-in-exams>

**Objective:** Students will preprocess the data, apply Multilinear and Polynomial Regression models, and compare performance metrics like RMSE and  $R^2$  score.

## Assignment Instructions:

### 1. Data Preprocessing - 4 marks

- Load the dataset.
- Handle missing values (if any).
- Encode categorical variables using one-hot encoding or label encoding.
- Normalize or standardize features (optional but recommended).

### 2. Feature Engineering - 4 marks

- Choose relevant features to predict **math score**.
- Create interaction or polynomial features (only for Polynomial Regression part).

### 3. Model Building - 4 marks

- Split the data into training and test sets (80/20).
- Train two models:
  - Multilinear Regression
  - Polynomial Regression (degree 2 or 3)

### 4. Model Evaluation - 5 marks

- Use metrics: RMSE, MAE, and  $R^2$  score.
- Visualize predicted vs actual values.

- Discuss overfitting or underfitting in Polynomial Regression.

#### 5. Submission Requirements - 3 marks

- Jupyter Notebook with code and markdown explanations.

## Part 2: Naive Bayes, KNN, and Decision Tree

**Dataset:** Social Network Ads Dataset (from Kaggle)

<https://www.kaggle.com/datasets/rakeshrau/social-network-ads>

**Objective:** Students will predict whether a user purchases a product based on their age and estimated salary using Naive Bayes, KNN, and Decision Tree classifiers.

### Assignment Instructions:

#### 1. Data Preprocessing - 4 marks

- Load the dataset (usually named `Social_Network_Ads.csv`).
- Drop the `User ID` column (not a useful feature).
- Encode the `Gender` column using Label Encoding or One-Hot Encoding.
- Encode the target variable `Purchased`.
- Standardize the feature columns (`Age`, `EstimatedSalary`).

#### 2. Model Building - 5 marks

- Split the dataset into training and test sets (75/25).
- Train the following models:
  - Gaussian Naive Bayes
  - K-Nearest Neighbors (test with  $k=3$ , 5, and 7)

- Decision Tree (try both Gini and Entropy as criteria)

### 3. Model Evaluation - 8 marks

- Evaluate all models using:
  - Accuracy
  - Precision
  - Recall
  - F1-Score
  - Confusion Matrix
- Plot decision boundaries (optional but encouraged for 2D features).
- Compare models and identify which one performs best and why.

### 4. Submission Requirements - 3 marks

- Submit a well-documented Jupyter Notebook (.ipynb).
- Include comments, markdown cells, and graphs.
- Include a 1-page summary analyzing the performance of all 3 models

#### **Deliverables:**

Only GitHub repository link will be accepted, containing both [.ipynb](#) notebooks with markdown explanations.