

# Machine Learning Assignment-2

Presented by

**K.Sagar Datta** 

M.Srikar Reddy

M.Deepak

K.Sai Teja

**C.Mohith Reddy** 



## **Supervised Machine Learning**

Task-1

The process of training a model by giving a labeled data where the target value is known

- .labeled data is divided into two parts
- 1) training dataset
- 2) testing dataset

## Advantages of using supervised ML

1)High accuracy when trained with sufficient labeled data 2)The presence of labeled data makes it easier to define a clear relationship between inputs and outputs, enabling the model to learn effectively.

#### example: calculations of weight based on height

there are two types of supervised ml 1)Regression task and 2)Classification tasks

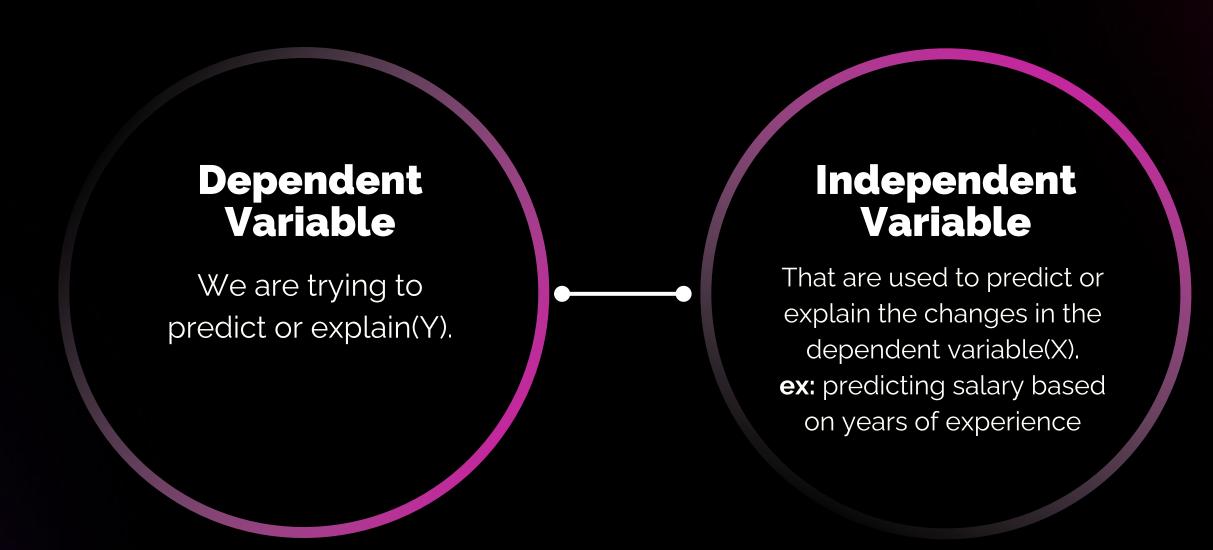


#### **Classification task:**

It is a supervised learning where the goal is to predict a category for a given input based on the training data.

#### **Regression task:**

It is the statistical method that help us to understand and predict the relation btw the varable.



# **Unsupervised Machine Learning**

Unsupervised learning involves analyzing data without labeled outputs to find hidden patterns or structures.

Key Algorithms used are K-Means Clustering, Hierarchical Clustering, DBSCAN, PCA (Principal Component Analysis) and Autoencoders

## Advantages

No need for labeled data, useful for exploratory data analysis. Challenges: Difficult to evaluate, results are not always interpretable, and prone to overfitting



# Supervised Learning

Classification	Regression			
Logistic Regression	Linear Regression			
Naive Bayes	Ridge Regression			
Linear Discriminant Analysis (LDA)	Lasso Regression			
Decision Trees				
Random Forest				
Support Vector Machines (SVM)				
k-Nearest Neighbors (KNN)				
Gradient Boosting algorithms				
Neural Networks				

# Unsupervised Learning

PCA	
K-mean Clustering	
Hierarchical Clustering	
DB Scan Clustering	

### REFERENCE

1..Nasteski, V. (2018). An overview of the supervised machine learning methods. Retrieved from ResearchGate

- 2. Author(s). (2023). Unsupervised machine learning for disease prediction: A comparative study. Fand Technology, 13(1), 45–56
- 3. Lu, H., & Uddin, S. (2024). Unsupervised machine learning for disease prediction: a comparative performance analysis using multiple datasets. Health and Technology, 14(1), 141-154

# Python Code Workflow for Data Manipulation and Regression

#### **Model Evaluation**

Assessing the model's performance

#### **Data Visualization**

Creating visual representations of data

#### **Load Data**

Reading data into the program for processing

#### **Model Training**

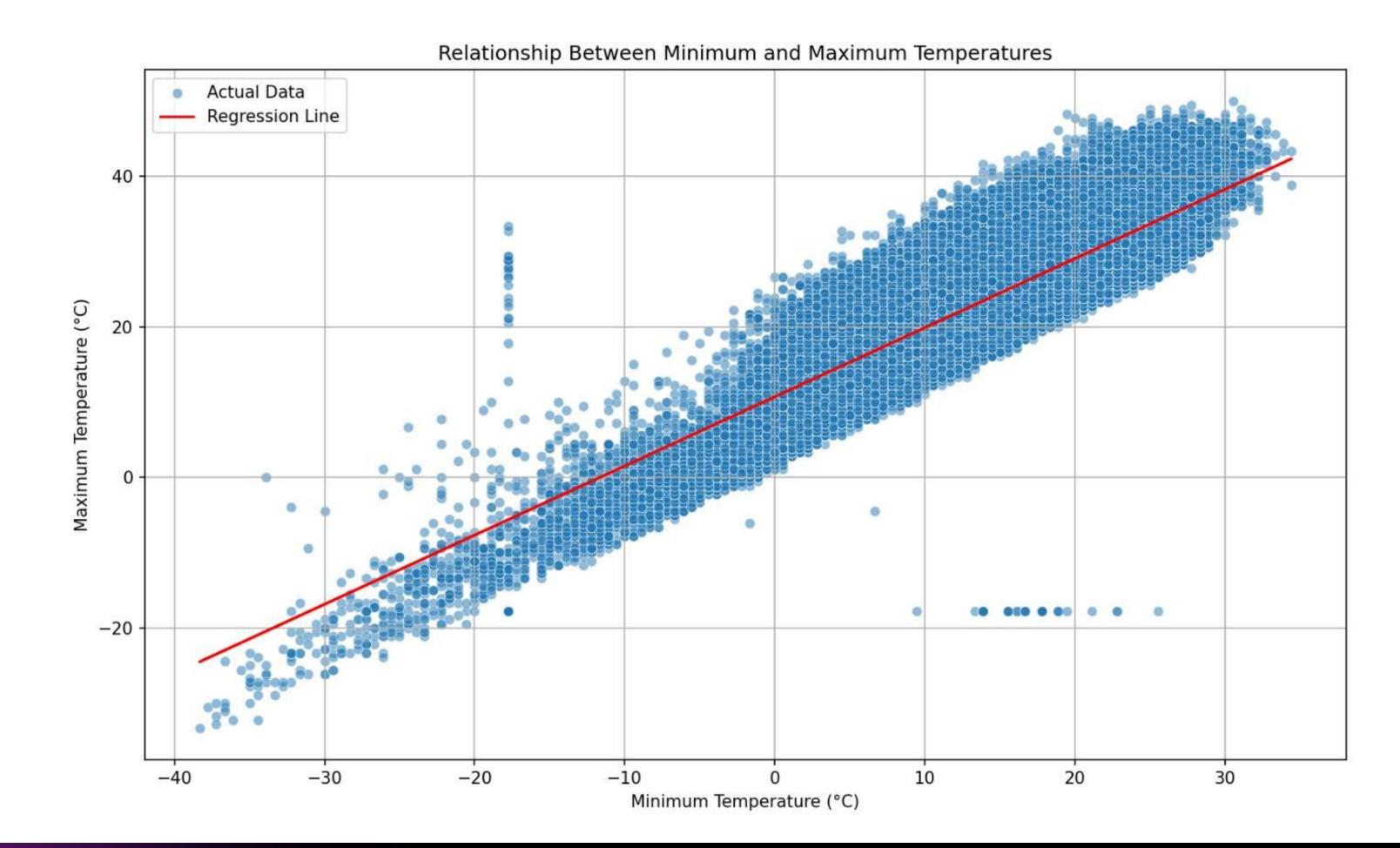
Applying algorithms to train the model

#### **Data Cleaning**

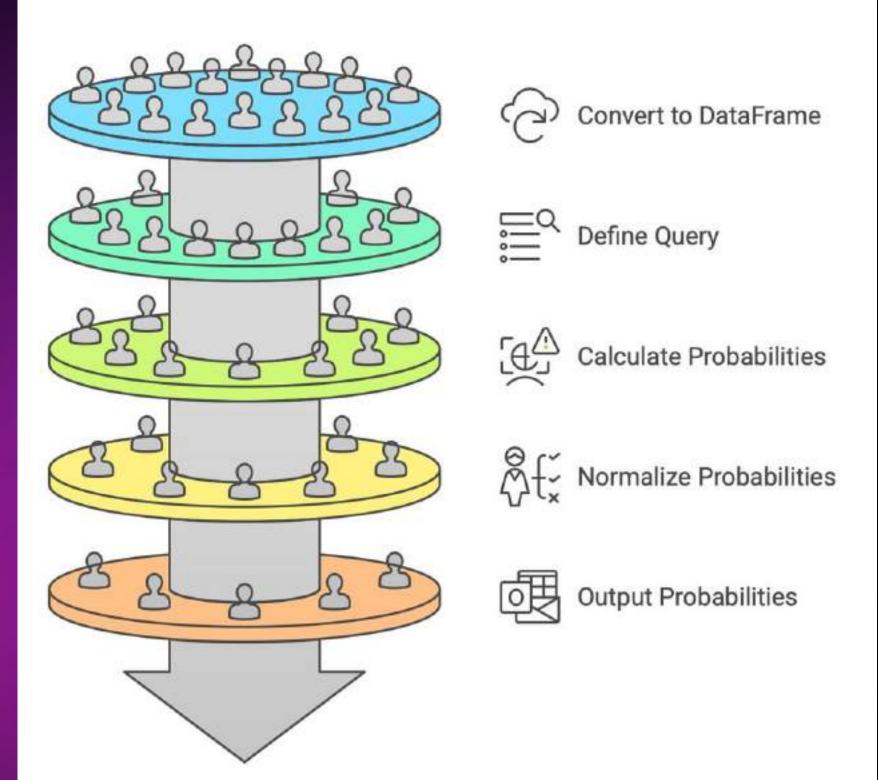
Removing or correcting erroneous data

#### **Import Libraries**

Loading necessary libraries for data manipulation and modeling



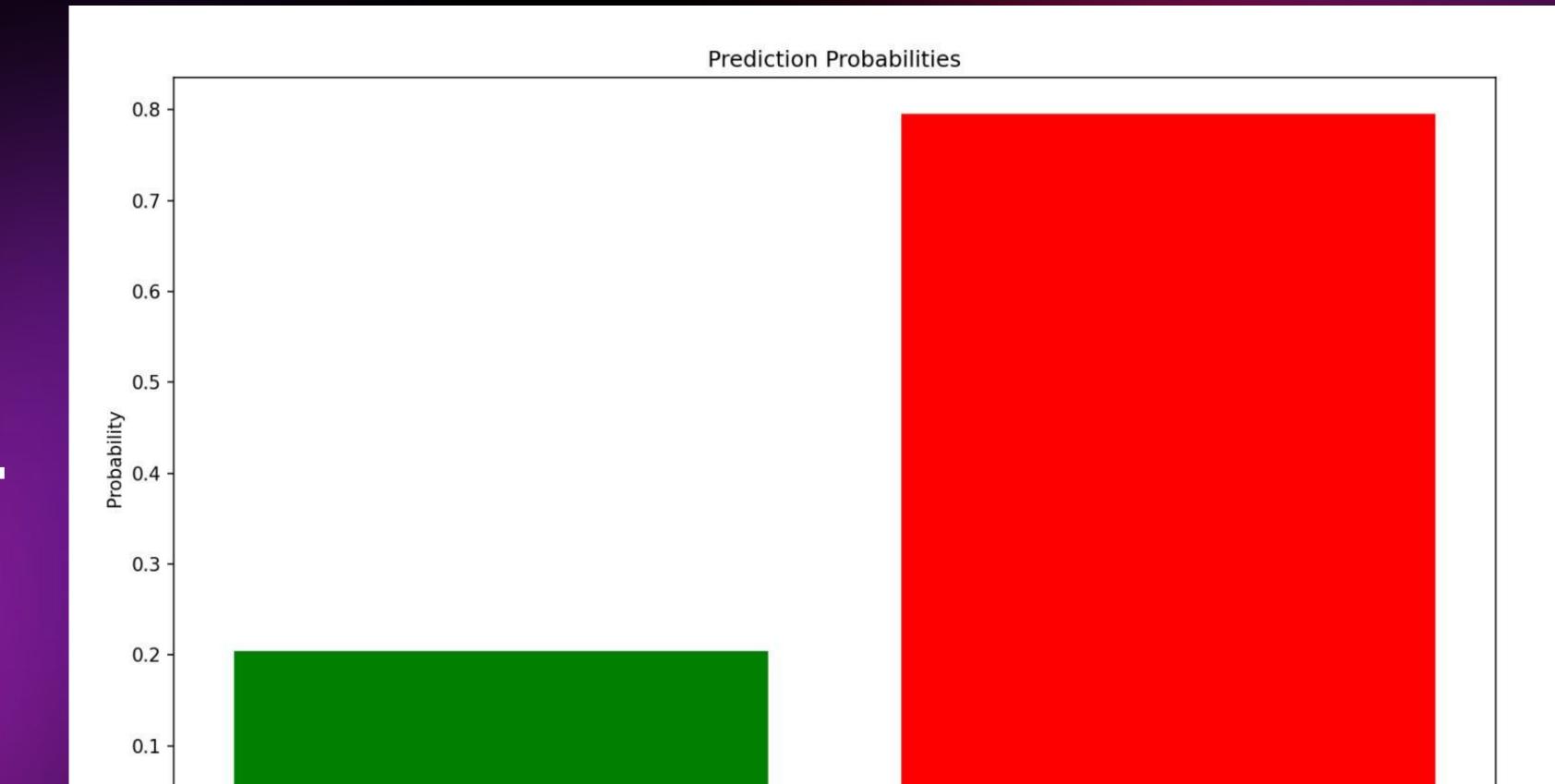
#### **Probability Calculation Funnel**



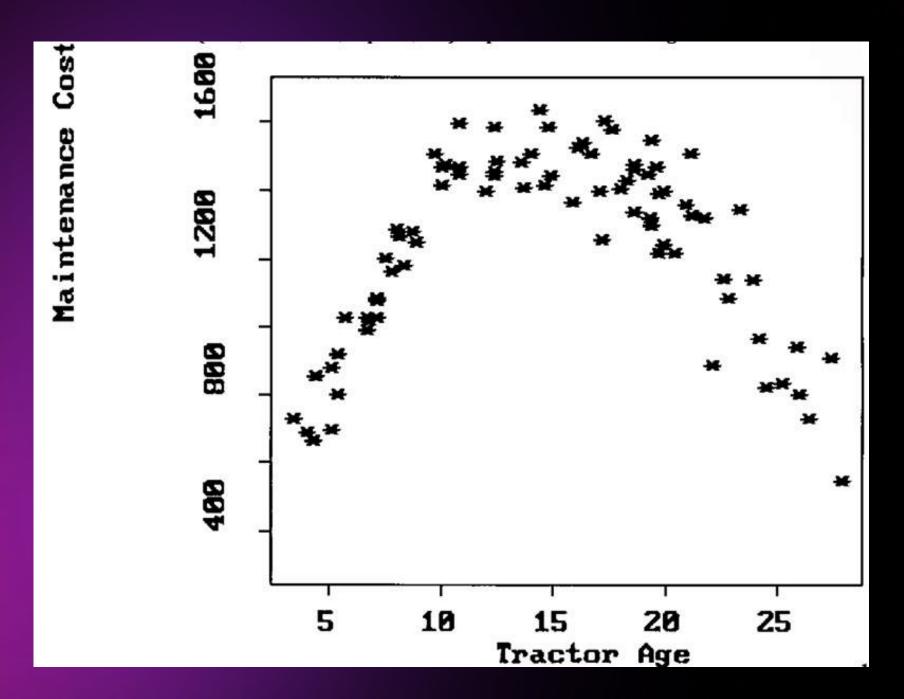
Outlook	Temperature	Humidity	Windy	Class
sunny	hot	high	false	_
sunny	hot	high	true	× —
overcast	hot	high	false	#
rain	mild	high	false	+
rain	cool	normal	false	+
rain	cool	normal	true	-
overcast	cool	normal	true	+
sunny	mild	high	false	_
sunny	cool	normal	false	+
rain	mild	normal	false	+
sunny	mild	normal	true	+
overcast	mild	high	true	+
overcast	hot	normal	false	+
rain	mild	high	true	

0.0

Play Tennis (+)



Not Play Tennis (-)



#### Methodology:

#### 1. Data Preparation

- The dataset consists of two numerical arrays:
- tractor\_age: Represents the age of the tractors in years.
- maintenance\_cost: Represents the corresponding maintenance cost in USD.

```
tractor_age = np.array([5, 8, 10, 12, 15, 18, 20, 23, 25])
maintenance_cost = np.array([800, 1100, 1400, 1550, 1600, 1450, 1300, 900, 700])
```

#### 2. Polynomial Regression (Degree 2)

- Polynomial regression is used because the relationship between tractor age and maintenance cost appears to be non-linear.
- The Polynomial.fit() function is used to fit a second-degree polynomial (quadratic function) to the data.
- convert().coef extracts the coefficients of the polynomial in standard form.

- 3. Generate Regression Line Data
  - We generate a smooth curve using 100 equally spaced points between 5 and 25.
  - Using the polynomial equation, we compute the predicted maintenance cost (y\_fit) for these points.

```
x_fit = np.linspace(5, 25, 100)
y_fit = coefficients[0] + coefficients[1] * x_fit + coefficients[2] * x_fit**2
```

#### 4. Plotting the Results

**Scatter Plot (Data Points)** 

• The original dataset is plotted as black dots to represent real-world data points.

```
plt.scatter(tractor_age, maintenance_cost, color='black', label='Data Points')
```

#### **Plot Regression Curve**

• The polynomial regression line is plotted in blue.

```
plt.plot(x_fit, y_fit, color='blue', label='Regression Line', linewidth=2)
```

#### 5. Customizing the Plot

- Labels & Title: Improves readability.
- Legend: Helps differentiate between data points and the regression curve.
- Grid & Ticks: Enhances clarity.

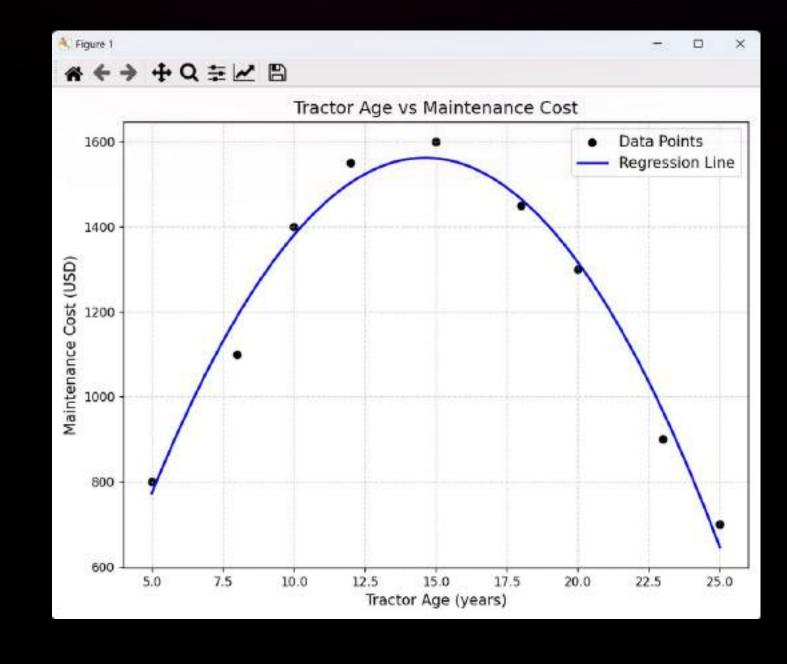
```
# Add labels, title, and legend
plt.xlabel("Tractor Age (years)", fontsize=12)
plt.ylabel("Maintenance Cost (USD)", fontsize=12)
plt.title("Tractor Age vs Maintenance Cost", fontsize=14)
plt.legend(fontsize=12)

# Customize grid and ticks
plt.grid(True, linestyle='--', alpha=0.5)
plt.xticks(fontsize=10)
plt.yticks(fontsize=10)
```

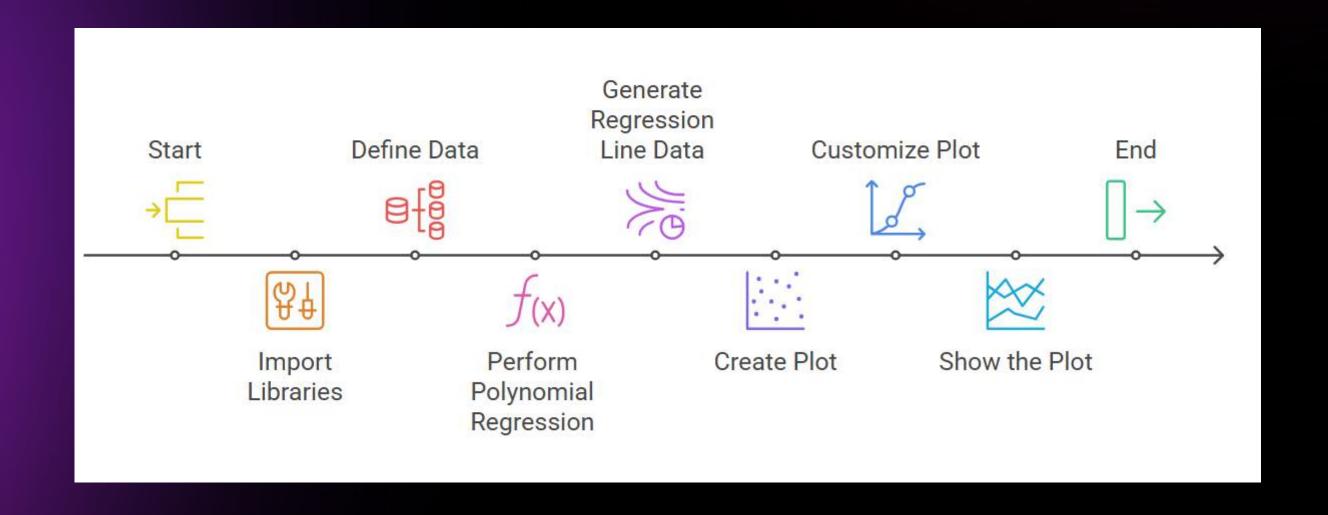
#### 6. Display the Plot

Ensures proper layout and displays the final visualization.

```
plt.tight_layout()
plt.show()
```



# Flowchart





# ThanKyou