# **Lesson 8: Linear Regression Model - Part 2**

#### **⋄** Introduction

Welcome back! In Part 1, we introduced linear regression and its core concepts. Now, it's time to get hands-on. In this lesson, you will:

- Learn how to train a linear regression model
- Understand how to evaluate it
- See an example in Python using scikit-learn

# Steps to Build a Linear Regression Model

## 1. Prepare the Data

Ensure your dataset has input features (X) and a target variable (y).

## 2. Split the Dataset

Typically into training and testing sets (e.g., 80/20 split).

#### 3. Train the Model

Fit the linear regression model to the training data.

#### 4. Make Predictions

Use the trained model to predict the target variable on test data.

#### 5. Evaluate Performance

Use metrics like MSE (Mean Squared Error) and R<sup>2</sup> (coefficient of determination).

# Python Implementation

```
import numpy as np
import pandas as pd
from sklearn.linear model import LinearRegression
from sklearn.model selection import train test split
from sklearn.metrics import mean_squared_error, r2_score
import matplotlib.pyplot as plt
# Sample data
data = {
  'Area (sq ft)': [1000, 1500, 2000, 2500, 3000],
  'Price ($)': [200000, 250000, 300000, 350000, 400000]
}
df = pd.DataFrame(data)
# Features and target
X = df[['Area (sq ft)']]
y = df['Price ($)']
# Train-test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Model
model = LinearRegression()
model.fit(X_train, y_train)
# Predictions
y pred = model.predict(X test)
# Evaluation
print("Mean Squared Error (MSE):", mean_squared_error(y_test, y_pred))
print("R<sup>2</sup> Score:", r2_score(y_test, y_pred))
# Visualization
plt.scatter(X, y, color='blue', label='Actual data')
plt.plot(X, model.predict(X), color='red', label='Regression Line')
plt.xlabel("Area (sq ft)")
plt.ylabel("Price ($)")
plt.legend()
plt.title("Linear Regression Example")
plt.show()
```

# **⋄** Evaluation Metrics Explained

Metric	What It Means
MSE	Measures the average squared error between predicted and actual values
R <sup>2</sup> Score	Indicates how well the model explains the variability of the target variable
	$R^2 = 1$ means perfect fit, $R^2 = 0$ means no predictive power

# Mini Exercise (Optional)

Try this yourself:

Modify the dataset to include more houses, or try predicting salaries based on years of experience using a dataset from Kaggle.

### ◇ Outro

Awesome work! In this lesson, you learned:

- How to train and test a linear regression model
- How to evaluate model performance
- ✓ How to use Python and scikit-learn to bring theory into practice

Linear regression is your launchpad for more advanced models like polynomial regression, logistic regression, and beyond.