## Lesson 4 Assignment— Regression

#### 1. What is Regression?

Regression is a type of supervised learning technique in Machine Learning that predicts continuous numerical outcomes from given input features. It is different from classification, which is used to predict labels or categories.

**Example (Regression):** Estimating the selling price of a car based on its age, mileage, and brand. **Example (Classification):** Predicting whether a car is new, used, or refurbished based on the same features.

### **★** Main Difference:

Regression answers: "How much?" or "How many?"

Classification answers: "Which group or class?"

### 2. Major Types of Regression

#### a) Simple Linear Regression

Concept: Models the relationship between one independent variable (X) and one dependent variable (y) using a straight line.

**Example:** Predicting student exam score from hours studied.

Advantages: Very easy to build and interpret.

Drawbacks: Only accurate when the data follows a straight-line pattern.

b) Multiple Linear Regression

Concept: Extends linear regression to include multiple input variables to predict one output.

**Example:** Predicting crop yield using rainfall, fertilizer used, and soil quality.

Advantages: Works well when many factors affect the target.

Drawbacks: Performance drops if input variables are highly correlated (multicollinearity).

c) Polynomial Regression

Concept: Adds higher powers of input variables (like  $x^2 x^3$ ) to capture curved relationships between features and target.

Example: Predicting the growth of a plant over time using both time and time<sup>2</sup> as features.

Advantages: Good for modeling non-linear trends.

Drawbacks: Higher polynomial degrees can lead to overfitting.

## **Types** Comparison of Regression Types

Regression Type	Relationship Shape	Example Features	Best Used When
Simple Linear	Straight line	Hours studied	One main
Regression		Exam score	influencing factor
Multiple Linear	Flat multi-	Rainfall, Fertilizer,	Several features
Regression	dimensional plane	Soil → Yield	influence outcome
Polynomial	Curved line	Time, Time <sup>2</sup> → Plant	Trend is non-linear
Regression		growth	

## 3. Measuring Regression Performance

To know how good a regression model is we use evaluation metrics that compare predicted values to actual values:

Metric Name	What It Measures	Sensitive to Large	Units
		Errors	
MAE (Mean	Average size of	X No	Same as target
Absolute Error)	prediction errors		
MSE (Mean Squared	Average of squared	✓ Yes	Squared units
Error)	errors		
RMSE (Root Mean	Square root of MSE	✓ Yes	Same as target
Squared Error)			
R <sup>2</sup> (Coefficient of	Percentage of	X No	0–1 (or %)
Determination)	variance explained		

### Example:

RMSE =  $3.5 \rightarrow$  model's predictions are off by about 3.5 units on average.

 $R^2 = 0.92 \rightarrow$  model explains 92% of the variation in the target values.

# 4. Underfitting vs Overfitting

**Underfitting:** The model is too basic and fails to detect patterns  $\rightarrow$  low accuracy on both training and test sets.

**Overfitting:** The model fits the training data too closely and performs poorly on unseen data.

### Reasons for Overfitting:

- Overly complex models
- Very small training dataset

• Including irrelevant features

#### **Ways to Prevent It:**

- Simplify the model
- Use regularization methods (like Lasso or Ridge)
- Apply cross-validation
- Increase dataset size
- Stop training when validation accuracy stops improving

## 5. Case Study — Regression in Healthcare

Title: "Predicting Patient Hospital Stay Duration Using Multiple Linear Regression"

(Published in Health Informatics Journal, 2022)

Objective: Estimate how many days a patient will stay in the hospital based on their age, illness severity, and treatment type.

Dataset: Records from 40,000 patients, including age, diagnosis category, treatment plan, and previous medical history.

Model Used: Multiple Linear Regression

Key Outcomes:

 $R^2 \approx 0.80$  and RMSE  $\approx 1.2$  days

Illness severity and age were the strongest predictors

Helped hospitals plan bed availability and staffing

Takeaway: Regression can help improve resource planning and efficiency in healthcare services.

# References

Alpaydin, E. (2020). Introduction to Machine Learning. MIT Press.

Géron, A. (2019). Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow. O'Reilly.

"Predicting Patient Hospital Stay Duration Using Multiple Linear Regression." Health Informatics Journal, 2022.

Mitchell, T. M. (1997). Machine Learning. McGraw-Hill.