## **Logistic Regression:-**

Logistic Regression is a **classification algorithm** used when the target variable is **categorical** (Yes/No, Spam/Not Spam, 0/1).

### Why not Linear Regression for Classification?

If we apply Linear Regression, the output can be any real number (e.g.,  $-\infty$  to  $+\infty$ ), but probabilities must lie between **0** and **1**.

So, we need a function that maps any real number to [0,1].

That function is the **Sigmoid function**.

Why is Logistic Regression called 'Regression' if it is used for classification?

- Because it uses a linear equation like Linear Regression (
  z=w0+w1x1+...z = w\_0 + w\_1x\_1 + ...z=w0+w1x1+...) inside,
  but then applies Sigmoid to convert it to probability.
- The name comes from the underlying regression equation.

### What is the decision boundary in Logistic Regression?

- A **threshold** that separates the classes.
- Usually, we use **0.5**:
  - $_{\circ}$  If probability ≥ 0.5  $\rightarrow$  Class 1
  - ∘ If probability  $< 0.5 \rightarrow \text{Class } 0$
- In terms of zzz, the decision boundary is where z=0z=0z=0.

### Why do we use the Sigmoid function in Logistic Regression?

- Because the output of the linear equation  $z=w0+w1x1+...z=w_0$ +  $w_1x_1+...z=w0+w1x1+...$  can be **any real number**.
- We need probabilities between **0** and **1**, so we apply the **Sigmoid** function:

$$\sigma(z)=11+e-z \cdot \sin(z) = \frac{1}{1+e^{-z}} \sigma(z)=1+e-z1$$

This squeezes values into [0, 1].

# Why can't we use Mean Squared Error (MSE) as the cost function in Logistic Regression?

- MSE assumes linear relationship, but Logistic Regression is non-linear because of Sigmoid.
- Using MSE can make the gradient descent very slow and may not converge.
- So, we use **Log Loss**, which works better for probabilities.

# What is the cost function used in Logistic Regression? Explain it.

• Log Loss (Binary Cross-Entropy):

## How do you handle multicollinearity in Logistic Regression?

- Multicollinearity = when features are highly correlated.
  Solutions:
- Remove correlated features.
- Use Principal Component Analysis (PCA).
- Apply regularization (L1 or L2).