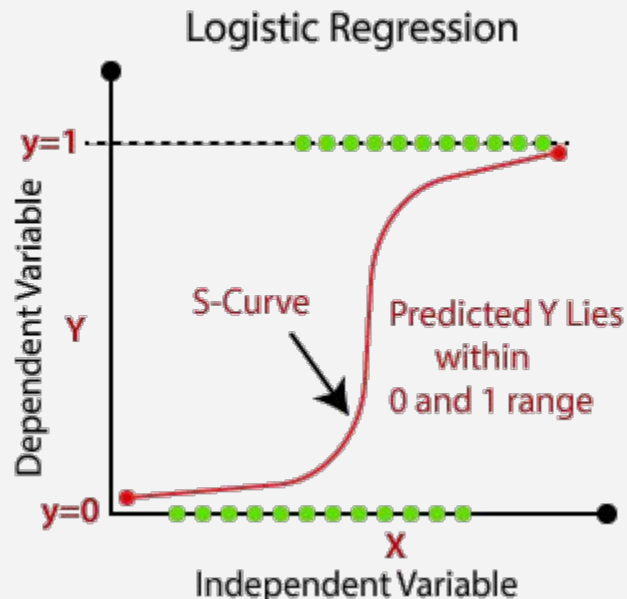
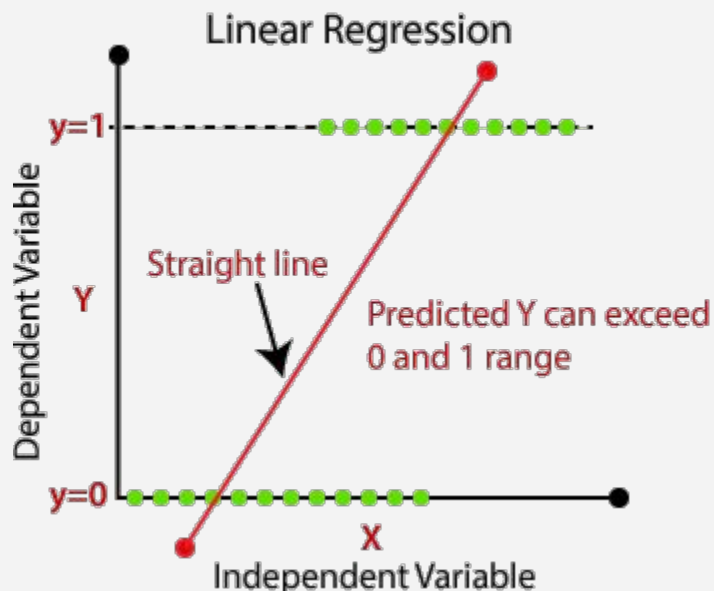


Logistic Regression



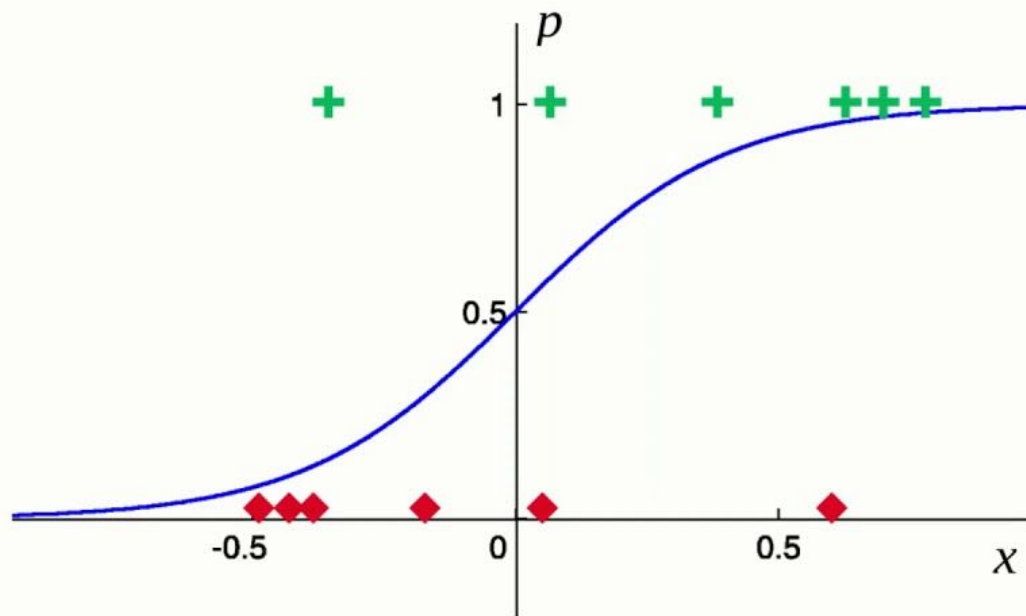
REGRESIÓN LOGÍSTICA

INTERPRETACIÓN

$$\log \frac{p}{1-p} = \beta_0 + \beta_1 x$$

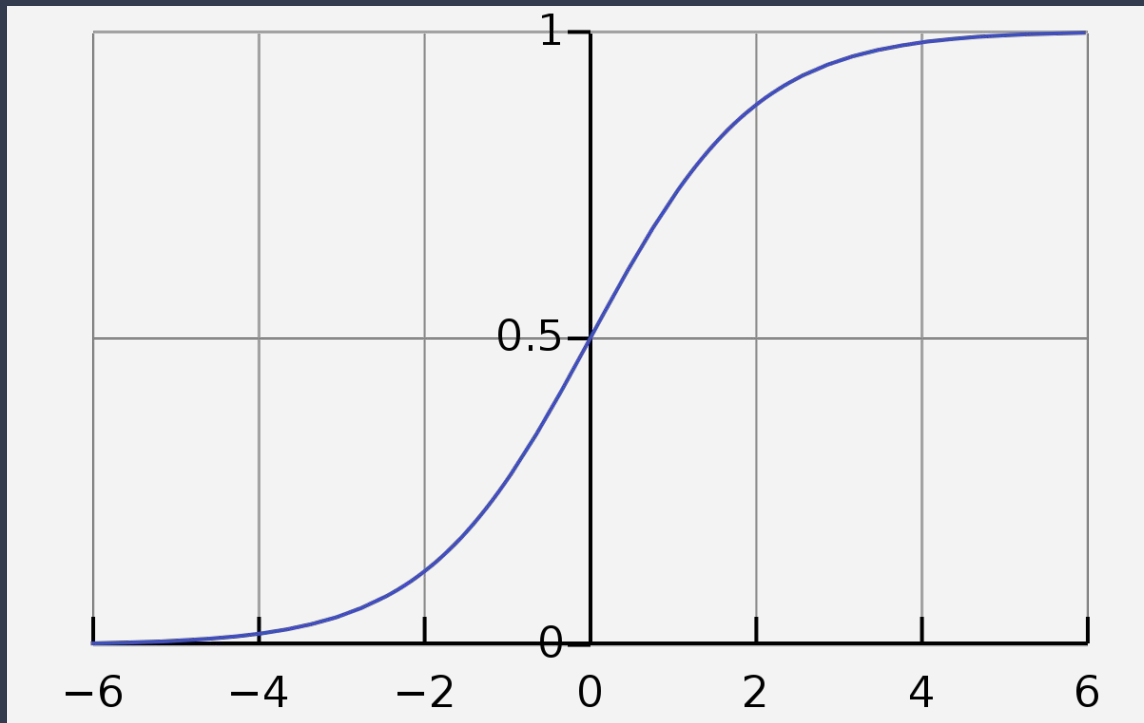
↓

$$p = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x)}}$$



Logística Regresion

[Enlace artículo Logística](#)



$$Q(t) = \frac{1}{1 + e^{-t}}$$

$$h(x) = \frac{1}{1 + e^{-\theta^T x}}$$

Logística Regresion

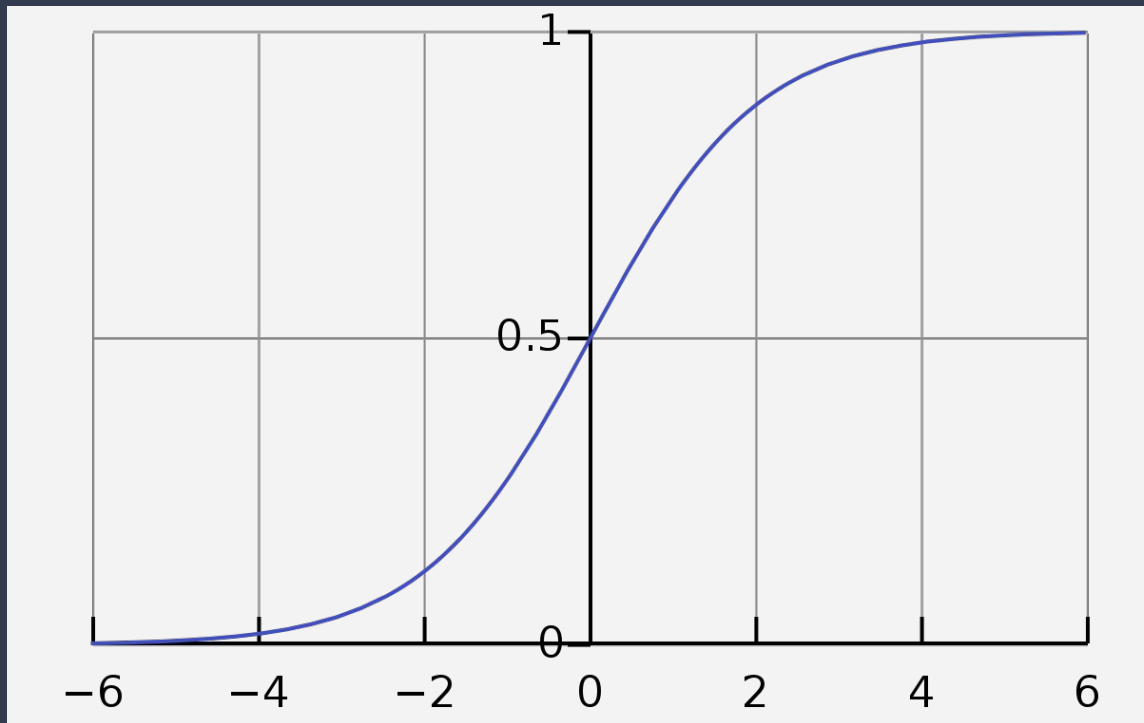
[Enlace artículo Logística](#)

$$\frac{1}{1 + e^{- (\beta_0 + \beta_1 X)}}$$

$$P = \frac{1}{1 + e^{-(-.5596 + 1.2528 X)}}$$

Logística Regresion

[Enlace artículo Logística](#)



$$Q(t) = \frac{1}{1 + e^{-t}}$$

$$h(x) = \frac{1}{1 + e^{-\theta^T x}}$$

Logistic Regression- Cost function

$$h(x) = \frac{1}{1 + e^{-\theta^T x}}$$

$$h(x) = \begin{cases} > 0.5, & \text{if } \theta^T x > 0 \\ < 0.5, & \text{if } \theta^T x < 0 \end{cases}$$

$$\text{cost} = \begin{cases} -\log(h(x)), & \text{if } y = 1 \\ -\log(1 - h(x)), & \text{if } y = 0 \end{cases}$$

$$\text{cost}(h(x), y) = -y \log(h(x)) - (1 - y) \log(1 - h(x))$$

Cost Function

$$J(\theta) = -\frac{1}{m} \sum_{i=1}^m [y^i \log(h(x^i)) + (1 - y^i) \log(1 - h(x^i))]$$

Gradient

$$\frac{\partial J(\theta)}{\partial \theta_j} = \frac{1}{m} \sum_{i=1}^m (h(x^i) - y^i) x_j^i$$

Multi classification problem

- First each class is separated in columns as a binary columns
- 1 if the row belongs to this class
- 0 otherwise

$$\hat{p}_\ell^* = \frac{e^{\hat{y}_\ell}}{\sum_{l=1}^C e^{\hat{y}_l}}$$