

Logistic Regression: Testing

To test your model, you would run a subset of your data, known as the validation set, on your model to get predictions. The predictions are the outputs of the sigmoid function. If the output is ≥ 0.5 , you would assign it to a positive class. Otherwise, you would assign it to a negative class.

$$\begin{aligned}
 & \bullet X_{val} \quad Y_{val} \quad \theta \\
 & \quad h(X_{val}, \theta) \\
 & \text{pred} = h(X_{val}, \theta) \geq 0.5 \quad \begin{bmatrix} 0.3 \\ 0.8 \\ 0.5 \\ \vdots \\ h_m \end{bmatrix} \geq 0.5 = \begin{bmatrix} \underline{0.3 \geq 0.5} \\ \underline{0.8 \geq 0.5} \\ \underline{0.5 > 0.5} \\ \vdots \\ \text{pred}_m \geq 0.5 \end{bmatrix} = \begin{bmatrix} \underline{0} \\ \underline{1} \\ \underline{1} \\ \vdots \\ \text{pred}_m \end{bmatrix}
 \end{aligned}$$

In the video, I briefly mentioned X validation. In reality, given your X data you would usually split it into three components. $X_{train}, X_{val}, X_{test}$. The distribution usually varies depending on the size of your data set. However, an 80, 10, 10 split usually works fine.

To compute accuracy, you solve the following equation:

$$\text{Accuracy} \longrightarrow \sum_{i=1}^m \frac{(pred^{(i)} == y_{val}^{(i)})}{m}$$

In other words, you go over all your training examples, m of them, and then for every prediction, if it was right you add a one. You then divide by m .