

Binary Classification

- Given a dataset \mathcal{D} , we like to construct a function

$$f : \mathcal{D} \rightarrow \{0, 1\},$$

where 0 = (category 0) and 1 = (category 1).

- Examples
 - 1 $\mathcal{D} = \{\text{SAT scores}\}$, 0 = rejection, 1 = admission
 - 2 $\mathcal{D} = \{\text{emails}\}$, 0 = non-spam, 1 = spam

- More precisely, we will construct a function

$$f : \mathcal{D} \rightarrow [0, 1]$$

so that $f(d)$, $d \in \mathcal{D}$, represents the **probability**
that d belongs to (category 1).

- In the SAT scores example,

$$f(1350) = 0.732$$

would mean "A student with SAT score 1350 is accepted with probability 0.732".

Q: How can we construct such a function?

- In linear regression, we use a linear model and minimize the mean square error (MSE).
- In [logistic regression](#), our strategy will be similar to that of linear regression, and the method is called **maximum likelihood estimation** (MLE).

A *training set* \mathcal{T} is given with known classification.

- Step 1: Using a **probabilistic model**, write a function out of \mathcal{T} with unknown *parameters* or *weights*.
- Step 2: Determine the parameters so that the known classification may have the **maximum likelihood**.

It is customary to take the negative log of the likelihood function, and the resulting function is called the **cross-entropy**. Then we need to *minimize* the cross-entropy.

The cross-entropy function will look like

$$E(\mathbf{w}) = - \sum_{n=1}^N \{t_n \ln y_n + (1 - t_n) \ln(1 - y_n)\},$$

where

$$y_n = \sigma(\mathbf{w}_1 x_{n1} + \mathbf{w}_2 x_{n2} + \dots + \mathbf{w}_k x_{nk} + \mathbf{w}_{k+1})$$

and

$$\sigma(x) = \frac{1}{1 + e^{-x}}.$$

Goal: Determine $\mathbf{w} = (w_1, w_2, \dots, w_{k+1})$ which minimizes $E(\mathbf{w})$.

- The cross-entropy $E(\mathbf{w})$ is not linear, and it is not possible to calculate a closed-form formula for \mathbf{w}_* which minimizes $E(\mathbf{w})$.
- On the other hand, it can be shown that $E(\mathbf{w})$ is convex, and a global minimum exists.
- We will find an approximate value for \mathbf{w}_* using [gradient descent](#) and [Newton's method](#).
- The method of gradient descent is widely used in many other parts of machine learning.

Things to Do for Binary Classification

- 1 Gradient Descent and Newton's Method
- 2 Probability Theory
- 3 Logistic Regression