Problem 1

Logistic regression, likelihood, maximize likelihood, negative log likelihood.

Solution

$$p(C1 \mid \phi) = y(\phi) = \sigma(\mathbf{w}^T \phi)$$
$$p(C2 \mid \phi) = 1 - p(C1 \mid \phi)$$

For a dataset ϕ_n, t_n , where $t_n \in 0, 1$ and $\phi_n = \phi(x_n)$ with n = 1, ..., N the likelihood function can be written as:

$$p(t \mid \mathbf{w}) = \prod_{n=1}^{N} y_n^{t_n} (1 - y_n)^{1 - t_n}$$

where:

$$t_n = (t_1, t_N)^T, y_n = p(C1 \mid \phi_n)$$

Taking negative logarithm we have:

$$L = -logp(t \mid \mathbf{w}) = -\sum_{n=1}^{N} t_n log y_n + ((1 - t_n)log(1 - y_n))$$

Where

$$y_n = \sigma(a_n), a_n = \mathbf{w}^T \phi_n$$

We have:

$$\begin{split} \frac{\partial L}{\partial w} &= \Sigma_{n=1}^N (\frac{\partial L}{\partial y_n} \times \frac{\partial y_n}{\partial a_n} \times \frac{\partial a_n}{\partial w}) \\ \frac{\partial L}{\partial w_n} &= -\frac{t_n}{y_n} + \frac{(1-t_n)}{(1-y_n)} = \frac{y_n - t_n}{y_n (1-y_n)} \\ \frac{\partial y}{\partial a_n} &= \sigma(a_n) (1-\sigma(a_n)) = y_n (1-y_n) \\ \frac{\partial a_n}{\partial w} &= \phi_n \end{split}$$

So:

$$\frac{\partial L}{\partial w} = \sum_{n=1}^{N} \left(\frac{\partial L}{\partial y_n} \times \frac{\partial y_n}{\partial a_n} \times \frac{\partial a_n}{\partial w} \right) = \sum_{n=1}^{N} \frac{y_n - t_n}{y_n (1 - y_n)} \times y_n (1 - y_n) \times \phi_n = \sum_{n=1}^{N} (y_n - t_n) \phi_n$$

Problem 2

Find the function f(x) with f'(x) = f(x)(1-f(x))

Solution

$$f'(x) = f(x)(1 - f(x))$$

$$\iff \frac{d(f(x))}{dx} = f(x)(1 - f(x))$$

$$\implies \frac{d(f(x))}{f(x)(1 - f(x))} = dx$$

$$\implies \int \frac{d(f(x))}{f(x)(1 - f(x))} = \int dx$$

$$\implies \int \frac{1}{f(x)(1 - f(x))} d(f(x)) = \int dx$$

$$\longrightarrow \int \frac{1}{f(x)} + \frac{1}{1 - f(x)} d(f(x)) = \int dx$$

$$\iff \ln(f(x)) - \ln(1 - f(x)) = x$$

$$\iff \frac{f(x)}{1 - f(x)} = e^x$$

$$\longrightarrow f(x) = e^x (1 - f(x))$$

$$\iff f(x) = e^x - e^x f(x)$$

$$\iff 1 = \frac{e^x}{f(x)} - e^x$$

$$\iff f(x) = \frac{e^x}{1 + e^x} = \sigma(x)$$