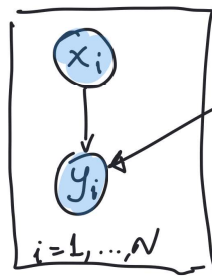


The logistic regression model

Given: $x_{1:N} = (x_1, \dots, x_N)$, $y_{1:N} = (y_1, \dots, y_N)$; $y_i \in \{0, 1\}$

Find: $p(y | x, x_{1:N}, y_{1:N}) = ?$



Likelihood:

$$p(y_i = 1 | x_i, \underline{w}) = f(w_0 + w_1 x_i)$$

$$f(z) = \text{sigm}(z) = \frac{\exp\{z\}}{1 + \exp\{z\}}$$

$$p(y_i = 1 | x_i, \underline{w}) = \text{sigm}(w_0 + w_1 x_i)$$

Important

$$p(y_i = 0 | x_i, \underline{w}) = 1 - p(y_i = 1 | x_i, \underline{w}) = 1 - \text{sigm}(w_0 + w_1 x_i)$$

$$p(y_i | x_i, \underline{w}) = \underbrace{\left[\text{sigm}(w_0 + w_1 x_i) \right]}_{\text{activated when } y_i = 1} \cdot \underbrace{\left[1 - \text{sigm}(w_0 + w_1 x_i) \right]}_{\text{activated when } y_i = 0}^{1-y_i}$$

$$p(y_{1:N} | x_{1:N}, \underline{w}) = \prod_{i=1}^N p(y_i | x_i, \underline{w})$$

$$= \prod_{i=1}^N \left[\text{sigm}(w_0 + w_1 x_i) \right]^{y_i} \cdot \left[1 - \text{sigm}(w_0 + w_1 x_i) \right]^{1-y_i}$$