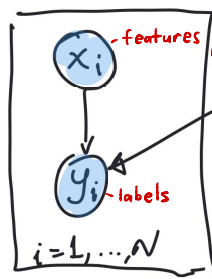


# The logistic regression model

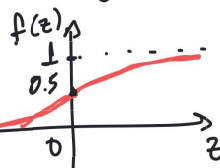
Given:  $x_{1:N} = (x_1, \dots, x_N)$ ,  $y_{1:N} = (y_1, \dots, y_N)$ ;  $y_i \in \{0, 1\}$

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



parameters  $w$  Likelihood:

$$p(y_i = 1 | x_i, w) = f(w_0 + w_1 x_i)$$



$$f(z) = \text{sigmoid}(z) = \frac{\exp\{z\}}{1 + \exp\{z\}} \in [0, 1]$$

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

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

for a single observation

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

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

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