

For sample i ,

$$P(x^{(i)} \in C_1) = y^{(i)} = f(x^{(i)}; \theta)$$

Thus, our model tells us that the likelihood of observing $x^{(i)}$ is $y^{(i)}$. If $x^{(i)} \in C_0$, then the likelihood of observing $x^{(i)}$ is $1 - y^{(i)}$.

Putting those together into one formula that works for both cases...

$$P(x^{(i)} \in C_{t^{(i)}}) = (y^{(i)})^{t^{(i)}} (1 - y^{(i)})^{1 - t^{(i)}} \text{ for } t^{(i)} \in \{0, 1\}.$$

The likelihood of observing ALL the samples

$$x^{(i)} \in C_{t^{(i)}} \text{ for } i = 1, \dots, N$$

is the product of all their individual probabilities.

$$P(x^{(i)} \in C_{t^{(i)}} \text{ for } i = 1, \dots, N)$$

$$= \prod_{i=1}^N (y^{(i)})^{t^{(i)}} (1 - y^{(i)})^{1 - t^{(i)}}$$

Taking the negative log-likelihood...

$$-\ln P(x^{(i)} \in C_{t^{(i)}} \text{ for } i = 1, \dots, N)$$

$$= - \left[\sum_{i=1}^N \ln (y^{(i)})^{t^{(i)}} + \ln (1 - y^{(i)})^{1 - t^{(i)}} \right]$$

$$= - \sum_{i=1}^N t^{(i)} \ln y^{(i)} + (1 - t^{(i)}) \ln (1 - y^{(i)})$$

as required. \blacksquare

