Logit Loss

$$L(heta) = \ln{(1 + e^{-yf(x, heta)})}, y \in \{-1,1\}, f(x, heta)$$

求导

$$\frac{\partial L_{ll}}{\partial \theta_{i}} = \frac{-ye^{-yf(x,\theta)}}{1 + e^{-yf(x,\theta)}} \frac{\partial f(x,\theta)}{\partial \theta_{i}}
= \frac{-y}{1 + e^{yf(x,\theta)}} \frac{\partial f(x,\theta)}{\partial \theta_{i}}
= -y(1 - \frac{1}{1 + e^{-yf(x,\theta)}}) \frac{\partial f(x,\theta)}{\partial \theta_{i}}$$
(1)

当y = 1,代入(1)

$$\frac{\partial L_{ll}}{\partial \theta_i} = -\left(1 - \frac{1}{1 + e^{-f(x,\theta)}}\right) \frac{\partial f(x,\theta)}{\partial \theta_i} \\
= \frac{-e^{-f(x,\theta)}}{1 + e^{-f(x,\theta)}} \frac{\partial f(x,\theta)}{\partial \theta_i} \tag{2}$$

$$\frac{\partial L_{ll}}{\partial \theta_i} = \left(1 - \frac{1}{1 + e^{f(x,\theta)}}\right) \frac{\partial f(x,\theta)}{\partial \theta_i} \\
= \frac{e^{f(x,\theta)}}{1 + e^{f(x,\theta)}} \frac{\partial f(x,\theta)}{\partial \theta_i} \tag{3}$$

Cross entropy

根据这里的推导,直接给出导数形式, 特别注意 $y \in \{0,1\}$,与Logit Loss不一致。

$$\frac{\partial L_{ce}}{\partial \theta_i} = \left(\frac{1}{1 + e^{-f(x,\theta)}} - y\right) \frac{\partial f(x,\theta)}{\partial \theta_i} \tag{4}$$

y=1代入(4)

$$\frac{\partial L_{ce}}{\partial \theta_i} = \left(\frac{1}{1 + e^{-f(x,\theta)}} - 1\right) \frac{\partial f(x,\theta)}{\partial \theta_i}
= \frac{-e^{-f(x,\theta)}}{1 + e^{-f(x,\theta)}} \frac{\partial f(x,\theta)}{\partial \theta_i}$$
(5)

$$y=0$$
代入 (4)

$$\frac{\partial L_{ce}}{\partial \theta_i} = \frac{1}{1 + e^{-f(x,\theta)}} \frac{\partial f(x,\theta)}{\partial \theta_i}
= \frac{e^{f(x,\theta)}}{1 + e^{f(x,\theta)}} \frac{\partial f(x,\theta)}{\partial \theta_i}$$
(6)

结论

将y代入的目的,是为了理清y的取值范围不同带来的干扰。最后等式(2)与(5)相同,等式(3)与(6)相同,所以Cross Entroy与Logit Loss本质上是相同的损失函数,由于y的取值范围不同,导致形式不一样。