

Loss - Function - Forward

softmax function

- to apply ~~is~~ softmax on i th entry

$$q_i = \frac{e^{\hat{y}_i}}{\sum_j e^{\hat{y}_j}} \quad \hat{y} : \text{network output}$$

subtract by largest entry

$$q_i = \frac{e^{\hat{y}_i - \max_j \hat{y}_j}}{\sum_k e^{\hat{y}_k - \max_j \hat{y}_j}} = \frac{e^{(-\max_j \hat{y}_j) \cdot \hat{y}_i}}{\sum_k e^{(-\max_j \hat{y}_j) \cdot \hat{y}_k}} = \frac{e^{-\max_j \hat{y}_j} \cdot e^{\hat{y}_i}}{e^{-\max_j \hat{y}_j} \cdot \sum_k e^{\hat{y}_k}} = \frac{e^{\hat{y}_i}}{\sum_k e^{\hat{y}_k}}$$

Loss Function

Cross Entry Loss : measures the differences between two probability distributions.

The Loss for a single sample :

$$L(p, q) = - \sum_{k=1}^K p_k \log(q_k) = - \sum_{k=1}^K \log(q_k) \quad \text{where } p_k = 1$$

p_k : the k^{th} element of p

because we use one-hot coding,

p_k is either 0 or 1

p : the true label with p_k

q_k : the k^{th} element of q

q : the network output with the softmax applied

K : the number of elements in p and q

Apply the loss function to multiple samples

$$\text{Loss} = - \frac{1}{N} \sum_{n=1}^N \sum_{k=1}^K p_{n,k} \log(q_{n,k})$$

concret example :

$K=3$, $p = [0, 1, 0]$, network output = $[1.2, 2.3, 0.8]$

Step 1 : calculate q_i

Step : calculate cross entropy loss

$$q_k = \frac{e^{\hat{y}_k}}{\sum_j e^{\hat{y}_j}}$$

$$L(p, q) = - p_1 \log(q_1) + [-p_2 \log(q_2)] + [-p_3 \log(q_3)]$$

$$= -p_2 \log(q_2) = -\log(0.642) = 0.443$$

$$* p[0] = p_1 = 0, p[2] = p_3 = 0$$

$$\sum_j e^{\hat{y}_j} = e^{1.2} + e^{2.3} + e^{0.8} = 15.52$$

$$q = \left[\frac{e^{1.2}}{15.52}, \frac{e^{2.3}}{15.52}, \frac{e^{0.8}}{15.52} \right] = [0.214, 0.642, 0.144]$$

because $q_i = \frac{e^{\hat{y}_i - \max_j \hat{y}_j}}{\sum_k e^{\hat{y}_k - \max_j \hat{y}_j}}$

$$L(p, q) = - \sum_{k=1}^K p_k \log(q_k) = - \sum_{k=1}^K \log(q_k), \quad \text{where } p_k = 1$$

$$\Rightarrow L(p, q) = - \log(q_i), \quad \text{because we use one hot coding.}$$

$$\Rightarrow L(p, q) = - \log \left(\frac{e^{\hat{y}_i - \max_j \hat{y}_j}}{\sum_k e^{\hat{y}_k - \max_j \hat{y}_j}} \right)$$

$$= - (\hat{y}_i - \max_j \hat{y}_j) + \log \left(\sum_{k=1}^K e^{\hat{y}_k - \max_j \hat{y}_j} \right)$$

$$\text{Loss} = - \frac{1}{N} \sum_{n=1}^N \log(q_{n, \kappa})$$

$$= - \frac{1}{N} \sum_{n=1}^N \left(- (\hat{y}_i - \max_j \hat{y}_j) + \log \left(\sum_{k=1}^K e^{\hat{y}_k - \max_j \hat{y}_j} \right) \right)$$