

## Notes DLOI Loss function

## Slide 20 Handling Multiple Classes

$$P_i = \frac{\exp(y_i)}{\sum_{j=1}^C \exp(y_j)}$$

\*  $P_i \in [0, 1]$

$$\sum p_i = 1$$

the bigger the score is, the higher the possibility is.

$y_1, \dots y_c$  are the  $C$  outputs of our network

$q$ : we construct a one-hot encoding  $q$  of the class label  $t$ .

$H(q, p)$ : cross entropy ,

$$H(q, p) = - \sum_{i=1}^c q_i \log p_i$$

$$\begin{aligned}
 l(y_t, t) &= H(q, p) = - \sum_{i=1}^c q_i \log p_i \\
 &= -\log p_t \\
 &= -\log \frac{\exp(y_t)}{\sum_{j=1}^c \exp(y_j)} \\
 &= \log \sum_{j=1}^c \exp(y_j) - y_t
 \end{aligned}$$

### Case B:

$$y = (6, 3, 2)$$

$$P_1 = \frac{e^6}{e^6 + e^3 + e^2} \approx 0.94$$

$$I = -\log(0.94) \approx 0.06$$

- \* When the loss is small, the score  $y_t$  of the correct class must be higher than the score of all other classes.

Example :

assume three classes :  $y = (y_1, y_2, y_3)$

the true label is  $t=1$

Case A: Wrong classes have higher scores.

$$y = (1, 10, 8)$$

$$P_1 = \frac{e^1}{e^1 + e^{10} + e^8} \approx \cancel{0}$$

$$t = -\log p_1 = -\log(0) = +\infty$$

