

Question 1: on a $\vec{w} = \arg \min_{\vec{w}} E_D(\vec{w})$

Les meilleurs paramètres sont ceux qui correspondent à $\nabla_{\vec{w}} E_D(\vec{w}) = 0$

$$\begin{aligned} \text{On a } \nabla_{\vec{w}} E_D(\vec{w}) &= \nabla_{\vec{w}} \left(\left(T - W^T \Phi \right)^2 + \lambda W^T W \right) \\ &= \nabla_{\vec{w}} \left(\left(T - W^T \Phi \right)^2 \right) + \lambda \nabla_{\vec{w}} (W^T W) \end{aligned}$$

$$= 2(T - W^T \phi) \nabla_{\vec{w}} (T - W^T \Phi) + \lambda \nabla_{\vec{w}} (W^T W)$$

on sait que T est indépendant de \vec{w} donc $\nabla_{\vec{w}}(T) = 0$

$$\begin{aligned} \nabla_{\vec{w}} E_D(\vec{w}) &= -2(T - W^T \phi) \nabla_{\vec{w}} (W^T \Phi) + \lambda \nabla_{\vec{w}} (W^T W) \\ &\quad (\text{on a } \nabla_{\vec{w}} (W^T \phi) = \phi^T) \end{aligned}$$

$$\nabla_{\vec{w}} E_D(\vec{w}) = 2(W^T \phi - T) \phi^T + 2\lambda W^T$$

$$\text{si on force } \nabla_{\vec{w}} E_D(\vec{w}) = 0$$

$$\text{on a } W^T \phi \cdot \phi^T - T \phi^T + \lambda W^T = 0$$

$$W^T (\phi \phi^T + \lambda I) = T \phi^T$$

$$W^T = T \phi^T (\phi \phi^T + \lambda I)^{-1}$$

Question 2 :

Soit E la fonction de perte

Supposons que E est une entropie croisée

$$E(\vec{w}) = \sum_{i=1}^n t_i \ln(y_i) + (1-t_i) \ln(1-y_i)$$

$$\begin{aligned}\vec{\nabla} E(\vec{w}) &= \sum_{i=1}^n t_i \vec{\nabla}(\ln(y_i)) + (1-t_i) \vec{\nabla}(\ln(1-y_i)) \\ &= \sum_{i=1}^n t_i \frac{\vec{\nabla}(\sigma(\vec{w}^T \phi_i))}{y_i} + (1-t_i) \frac{\vec{\nabla}(1-\sigma(\vec{w}^T \phi_i))}{1-y_i}\end{aligned}$$

$$\begin{aligned}\text{on a } \vec{\nabla}(\sigma(\vec{w}^T \phi_i)) &= \vec{\nabla} \left(\frac{1}{1 + \exp(-\vec{w}^T \phi_i)} \right) \\ &= - \frac{\vec{\nabla}(1 + \exp(-\vec{w}^T \phi_i))}{(1 + \exp(-\vec{w}^T \phi_i))^2} \\ &= + \frac{\vec{\nabla}(\vec{w}^T \phi_i) \exp(-\vec{w}^T \phi_i)}{(1 + \exp(-\vec{w}^T \phi_i))^2} \\ &= \phi_i \cdot \sigma(\vec{w}^T \phi_i) (1 - \sigma(\vec{w}^T \phi_i)) \\ &= \phi_i \cdot y_i (1 - y_i)\end{aligned}$$

$$\vec{\nabla} E(\vec{w}) = \sum_{i=1}^n t_i \frac{\cancel{\phi_i} y_i (1-y_i)}{\cancel{y_i}} - (1-t_i) \frac{\cancel{\phi_i} y_i (1-y_i)}{\cancel{1-y_i}}$$

$$= \sum_{i=1}^n t_i \phi_i (1-y_i) - (1-t_i) \phi_i y_i$$

$$= \sum_{i=1}^n \phi_i (t_i - \cancel{t_i} y_i - y_i + \cancel{t_i} y_i)$$

$$\vec{\nabla} E(\vec{w}) = \sum_{i=1}^n \phi_i (t_i - y_i)$$

Question 3: Définissons d'abord la fonction d'entropie:

$$E = -p_1 \log p_1 - p_2 \log p_2 - p_3 \log p_3$$

Nous avons 2 contraintes à respecter:

$$\begin{cases} p_1 + p_2 + p_3 = 1 \\ p_1 = 2p_2 \end{cases}$$

La fonction lagrangienne est définie comme suit:

$$L = E + \lambda(p_1 + p_2 + p_3 - 1) + \mu(p_1 - 2p_2)$$

Les probabilités qui maximisent l'entropie vérifient:

$$\begin{cases} \frac{\partial L}{\partial p_1} = -\log p_1 - 1 + \lambda + \mu = 0 & (1) \\ \frac{\partial L}{\partial p_2} = -\log p_2 - 1 + \lambda - 2\mu = 0 & (2) \\ \frac{\partial L}{\partial p_3} = -\log p_3 - 1 + \lambda = 0 & (3) \\ \frac{\partial L}{\partial \lambda} = p_1 + p_2 + p_3 - 1 = 0 & (4) \\ \frac{\partial L}{\partial \mu} = p_1 - 2p_2 = 0 & (5) \end{cases}$$

$$(1) - (2) \Rightarrow -\log p_1 + \log p_2 + \mu + 2\mu = 0$$

$$\Rightarrow -\log(2p_2) + \log p_2 + 3\mu = 0 \quad (\text{en utilisant (5)})$$

$$\Rightarrow -\log 2 - \cancel{\log p_2} + \cancel{\log p_2} + 3\mu = 0$$

$$\Rightarrow \mu = \frac{1}{3}$$

les équations deviennent :

$$\left\{ \begin{array}{l} \frac{\partial L}{\partial p_1} = -\log p_1 - \frac{2}{3} + \lambda = 0 \quad (1) \\ \frac{\partial L}{\partial p_2} = -\log p_2 - \frac{5}{3} + \lambda = 0 \quad (2) \\ \frac{\partial L}{\partial p_3} = -\log p_3 - 1 + \lambda = 0 \quad (3) \\ \frac{\partial L}{\partial \lambda} = p_1 + p_2 + p_3 - 1 = 0 \quad (4) \\ \frac{\partial L}{\partial z} = p_1 - 2p_2 = 0 \quad (5) \end{array} \right.$$

$$(3) \Rightarrow -\log p_3 - 1 + \lambda = 0$$

$$\Rightarrow -\log(1 - p_2 - p_1) - 1 + \lambda = 0$$

$$\Rightarrow -\log(1 - 3p_2) - 1 + \lambda = 0$$

$$\Rightarrow \log(1 - 3p_2) = \lambda - 1$$

$$\Rightarrow 1 - 3p_2 = 2^{\lambda-1}$$

$$\Rightarrow 3p_2 = 1 - 2^{\lambda-1}$$

$$\Rightarrow p_2 = \frac{1 - 2^{\lambda-1}}{3} \quad (*)$$

$$(2) \Rightarrow \log p_2 = \lambda - \frac{5}{3}$$

$$\Rightarrow p_2 = 2^{\lambda - \frac{5}{3}} \quad (**)$$

$$\text{donc } 2^{\lambda - \frac{5}{3}} = \frac{1 - 2^{\lambda-1}}{3} \quad (\text{d'après } (*) \text{ et } (**))$$

$$\text{d'où } 3 \cdot \frac{2^\lambda}{2^{5/3}} + \frac{2^\lambda}{2} = 1$$

$$2^\lambda \left(\frac{3}{2^{5/3}} + \frac{1}{2} \right) = 1$$

$$2^\lambda = \frac{1}{\frac{3}{2^{5/3}} + \frac{1}{2}}$$

$$p_1 = 2^{-\frac{2}{3} + \lambda}$$

$$= 2^{-\frac{2}{3}} \cdot 2^\lambda$$

$$= \frac{2^{-\frac{2}{3}}}{\frac{3}{2^{5/3}} + \frac{1}{2}}$$

$$= \frac{1}{\frac{3}{2} + \frac{1}{2^{1/3}}}$$

$$p_1 = \frac{2}{3 + 2^{2/3}}$$

$$p_2 = \frac{1}{3 + 2^{2/3}} \quad (\text{car } p_1 = 2p_2)$$

$$p_3 = 1 - 3p_2 = 1 - \frac{3}{3 + 2^{2/3}}$$

$$p_3 = \frac{2^{2/3}}{3 + 2^{2/3}}$$