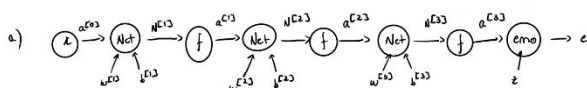


I. Pen-and-paper

1)

a. 1) $\omega^{E13} = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix}$ $\omega^{E23} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$ $\omega^{E33} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$

$$b^{[13]} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad b^{[23]} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad b^{[33]} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$



$$\lambda = [1 \ 1 \ 1 \ 1 \ 1]^T$$

$$z = [1 \ -13]^T$$

$$N^{C^{13}} = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 6 \\ 6 \end{pmatrix} + \begin{pmatrix} 6 \\ 6 \end{pmatrix} = \begin{pmatrix} 12 \\ 12 \end{pmatrix}$$

$$Q^{[2]} = f(N^{[2]}) = \begin{pmatrix} f(2f(6) + f(1) + 1) \\ f(2f(6) + f(1) + 1) \end{pmatrix}$$

$$Q^{[1]} = f(N^{[1]}) = \begin{pmatrix} 1(6) \\ 1(1) \\ 1(6) \end{pmatrix}$$

$$N^{[2]} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \left(\frac{1}{2} (2 \frac{1}{2}(6) + \frac{1}{2}(1) + 1) \right) + \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$N^{(2)} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} \frac{1}{3}(6) \\ \frac{1}{3}(1) \\ \frac{1}{3}(6) \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 2\frac{1}{3}(6) + \frac{1}{3}(1) + 1 \\ 2\frac{1}{3}(6) + \frac{1}{3}(1) + 1 \end{pmatrix}$$

$$a^{[3]} = f(N^{[3]}) = \begin{pmatrix} f(0) \\ f(0) \end{pmatrix}$$

$$E(z, a^{(3)}) = \frac{1}{2} \sum_{i=1}^2 (z_i - a_i^{(3)})^2$$

$$\frac{\partial E}{\partial a_{ij}^{(2)}} = \frac{1}{2} \sum_{k=1}^2 \frac{\partial E}{\partial a_{ij}^{(2)}} (z_k - a_{ij}^{(2)})^2 = \frac{1}{2} (-1) \cdot (z_1 - a_{ij}^{(2)}) = a_{ij}^{(2)} - z_1 ; \quad \nabla_{a^{(2)}} E = \begin{pmatrix} a_{11}^{(2)} - z_1 \\ a_{12}^{(2)} - z_1 \\ a_{21}^{(2)} - z_1 \\ a_{22}^{(2)} - z_1 \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$$

$$\nabla_{\vec{a}} C_3 = \begin{pmatrix} \frac{\partial C_3}{\partial a_1} \\ \frac{\partial C_3}{\partial a_2} \\ \frac{\partial C_3}{\partial a_3} \end{pmatrix} = \begin{pmatrix} \frac{\partial C_3}{\partial a_1} \\ \frac{\partial C_3}{\partial a_2} \\ \frac{\partial C_3}{\partial a_3} \end{pmatrix}$$

$$\nabla_{\alpha}^{[3]} = \begin{pmatrix} \frac{\partial \alpha_1^{[3]}}{\partial N_1^{[3]}} & \frac{\partial \alpha_2^{[3]}}{\partial N_1^{[3]}} \\ \frac{\partial \alpha_1^{[3]}}{\partial N_2^{[3]}} & \frac{\partial \alpha_2^{[3]}}{\partial N_2^{[3]}} \end{pmatrix}$$

$$a_{C3} = \begin{pmatrix} f(0) \\ 1(0) \end{pmatrix}; N_{C3} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$f'(x) = \left(\frac{e^x - e^{-x}}{e^x + e^{-x}} \right)' = \left(\frac{e^{2x} - 1}{e^{2x} + 1} \right)' = \frac{(e^{2x})'(e^{2x} + 1) - (e^{2x} - 1)(e^{2x})'}{(e^{2x} + 1)^2} = \frac{2e^{2x}(e^{2x} + 1) - (e^{2x} - 1)(2e^{2x})}{(e^{2x} + 1)^2}$$

$$\cdot \begin{bmatrix} \frac{\partial \{f(N_1^{CS})\}}{\partial N_1^{CS}} & \frac{\partial \{f(N_2^{CS})\}}{\partial N_1^{CS}} \\ \frac{\partial \{f(N_1^{CS})\}}{\partial N_2^{CS}} & \frac{\partial \{f(N_2^{CS})\}}{\partial N_2^{CS}} \end{bmatrix} = \begin{bmatrix} f'(N_1^{CS}) & 0 \\ 0 & f'(N_2^{CS}) \end{bmatrix} = \begin{bmatrix} \text{sech}^2(0) & 0 \\ 0 & \text{sech}^2(0) \end{bmatrix}$$

$$\frac{\partial f(\theta^{(j)})}{\partial \theta^{(j)}} = \begin{cases} f'(\theta^{(j)}) & , i=j \\ 0 & , i \neq j \end{cases}$$

$$N^{[2]} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \alpha^{[2]} = \begin{pmatrix} \frac{1}{2}(2f(6) + f(1) + 1) \\ \frac{1}{2}(2f(6) + f(1) + 1) \end{pmatrix}$$

$$\frac{\partial N_j^{[23]}}{\partial a_i^{[23]}} = \frac{\partial}{\partial a_i^{[23]}} \sum_{t=1}^L w_{jt}^{[23]} \cdot a_t^{[23]} + b_j^{[23]} = w_{ji}^{[23]}$$

$$\nabla_{\vec{a}} \vec{C} = \begin{bmatrix} \frac{\partial N_1}{\partial a_1} \vec{C} & \frac{\partial N_2}{\partial a_1} \vec{C} \\ \frac{\partial N_1}{\partial a_2} \vec{C} & \frac{\partial N_2}{\partial a_2} \vec{C} \end{bmatrix} = \begin{bmatrix} w_{11} \vec{C} & w_{21} \vec{C} \\ w_{12} \vec{C} & w_{22} \vec{C} \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\frac{\partial J_j}{\partial a_i} = \frac{\partial}{\partial a_i} \sum_{k=1}^L w_{jk} \cdot a_k + b_j = w_{ji}$$

$$\nabla_{N^{[2]}} \theta^{[2]} = \begin{bmatrix} \frac{\partial \theta^{[2]}}{\partial N_1^{[2]}} & \frac{\partial \theta^{[2]}}{\partial N_2^{[2]}} \\ \frac{\partial \theta^{[2]}}{\partial N_1^{[2]}} & \frac{\partial \theta^{[2]}}{\partial N_2^{[2]}} \end{bmatrix} = \begin{bmatrix} f'(N_1^{[2]}) & 0 \\ 0 & f'(N_2^{[2]}) \end{bmatrix} = \begin{bmatrix} \text{sech}^2(z \tanh(\theta) + \tanh(1) + 1) & 0 \\ 0 & \text{sech}^2(z \tanh(\theta) + \tanh(1) + 1) \end{bmatrix}$$

$$\nabla_{\alpha^{(1)}} f^{(2)} = \begin{bmatrix} w_{11}^{(2)} & w_{21}^{(2)} \\ w_{12}^{(2)} & w_{22}^{(2)} \\ w_{13}^{(2)} & w_{23}^{(2)} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\nabla_{NC^{13}} \alpha^{C13} = \begin{bmatrix} \frac{\partial \alpha^{C13}}{\partial N_1^{C13}} & \frac{\partial \alpha^{C13}}{\partial u_1^{C13}} & \frac{\partial \alpha^{C13}}{\partial \theta_1^{C13}} \\ \frac{\partial \alpha^{C13}}{\partial N_2^{C13}} & \frac{\partial \alpha^{C13}}{\partial u_2^{C13}} & \frac{\partial \alpha^{C13}}{\partial \theta_2^{C13}} \\ \frac{\partial \alpha^{C13}}{\partial N_3^{C13}} & \frac{\partial \alpha^{C13}}{\partial u_3^{C13}} & \frac{\partial \alpha^{C13}}{\partial \theta_3^{C13}} \end{bmatrix} = \begin{bmatrix} \text{sech}^4(\tanh(v)) & 0 & 0 \\ 0 & \text{sech}^2(\tanh(v)) & 0 \\ 0 & 0 & \text{sech}^4(\tanh(v)) \end{bmatrix}$$

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$$\begin{aligned}
 \delta^{[3]} &= \nabla_{w^{[3]}}^e = \nabla_{a^{[3]}}^e \cdot \nabla_{z^{[3]}}^e = \begin{pmatrix} \text{sech}^2(0) & 0 \\ 0 & \text{sech}^2(0) \end{pmatrix} \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \begin{pmatrix} -\text{sech}^2(0) \\ \text{sech}^2(0) \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \end{pmatrix} \\
 \nabla_{w^{[3]}}^e &= \delta^{[3]} \cdot a^{[2]T} = \begin{pmatrix} -1 \\ 1 \end{pmatrix} \begin{pmatrix} f(z^{[2]}(6) + f^{[1]}(1)) & f(z^{[2]}(6) + f^{[1]}(1)) \end{pmatrix} = \begin{pmatrix} -f(z^{[2]}(6) + f^{[1]}(1)) & -f(z^{[2]}(6) + f^{[1]}(1)) \\ f(z^{[2]}(6) + f^{[1]}(1)) & f(z^{[2]}(6) + f^{[1]}(1)) \end{pmatrix} \\
 w^{[3]} &= w^{[3]} - \eta \cdot \nabla_{w^{[3]}}^e = w^{[3]} - 0,1 \cdot \begin{pmatrix} -f(z^{[2]}(6) + f^{[1]}(1)) & -f(z^{[2]}(6) + f^{[1]}(1)) \\ f(z^{[2]}(6) + f^{[1]}(1)) & f(z^{[2]}(6) + f^{[1]}(1)) \end{pmatrix} \approx \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} - \begin{pmatrix} -0,09583 & -0,09583 \\ 0,09583 & 0,09583 \end{pmatrix} = \begin{pmatrix} 0,09583 & 0,09583 \\ -0,09583 & -0,09583 \end{pmatrix} \\
 \delta^{[2]} &= \nabla_{w^{[2]}}^e = \nabla_{a^{[2]}}^e \cdot \nabla_{z^{[2]}}^e \cdot \nabla_{w^{[2]}}^e = \begin{pmatrix} \text{sech}^2(z^{[2]}(6) + \tanh(1)) & 0 \\ 0 & \text{sech}^2(z^{[2]}(6) + \tanh(1)) \end{pmatrix} \cdot \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \\
 \nabla_{w^{[2]}}^e &= \delta^{[2]} \cdot a^{[1]T} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \begin{pmatrix} f(6) & f(1) & f(6) \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \\
 w^{[2]} &= w^{[2]} - \eta \cdot \nabla_{w^{[2]}}^e = w^{[2]} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \\
 \delta^{[1]} &= \nabla_{w^{[1]}}^e = \nabla_{a^{[1]}}^e \cdot \nabla_{z^{[1]}}^e \cdot \nabla_{w^{[1]}}^e = \begin{pmatrix} \text{sech}^2(\tanh(0)) & 0 & 0 \\ 0 & \text{sech}^2(\tanh(0)) & 0 \\ 0 & 0 & \text{sech}^2(\tanh(0)) \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \\
 \nabla_{w^{[1]}}^e &= \delta^{[1]} \cdot a^{[0]T} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \\
 w^{[1]} &= w^{[1]} - \eta \cdot \nabla_{w^{[1]}}^e = w^{[1]} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \\
 b^{[3]} &= b^{[3]} - \eta \cdot \delta^{[3]} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} - 0,1 \cdot \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0,1 \\ -0,1 \end{pmatrix} \\
 b^{[2]} &= b^{[2]} - \eta \cdot \delta^{[2]} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - 0,1 \cdot \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \\
 b^{[1]} &= b^{[1]} - \eta \cdot \delta^{[1]} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - 0,1 \cdot \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}
 \end{aligned}$$

b.

b) $F: \mathbb{R}^n \rightarrow \mathbb{R}^n$

$$F(x) = \begin{pmatrix} f(x_1) \\ \vdots \\ f(x_n) \end{pmatrix}$$

$$f(x_k) = \frac{e^{x_k}}{\sum_{i=1}^n e^{x_i}}$$

$$L(x, y) = -\sum_{i=1}^n x_i \log(y_i)$$

Aprendizagem 2021/22 Homework III – Group 17

Forward

$$\lambda = [1 \ 1 \ 1 \ 1 \ 1]^T$$

$$\beta = [1 \ 0 \ 0]^T$$

$$N^{(1)} = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 5 \\ 5 \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 6 \\ 6 \end{pmatrix}$$

$$\alpha^{(1)} = F(N^{(1)}) = \begin{pmatrix} f(2f(6)+f(1)+1) \\ f(2f(6)+f(1)+1) \end{pmatrix} = \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix}$$

$$a^{(1)} = F(N^{(1)}) = \begin{pmatrix} f(6) \\ f(1) \\ f(6) \end{pmatrix} \approx \begin{pmatrix} 0.49933 \\ 0.00066 \\ 0.49933 \end{pmatrix}$$

$$N^{(2)} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} f(6) \\ f(1) \\ f(6) \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 2f(6)+f(1)+1 \\ 2f(6)+f(1)+1 \end{pmatrix}$$

$$N^{(2)} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} f(2f(6)+f(1)+1) \\ f(2f(6)+f(1)+1) \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\alpha^{(2)} = F(N^{(2)}) = \begin{pmatrix} f(0) \\ f(0) \end{pmatrix}$$

$$\nabla_{a^{(2)}} \mathcal{L} = \begin{bmatrix} -2x_1 \cdot \frac{1}{a_1^{(2)}} \\ -2x_2 \cdot \frac{1}{a_2^{(2)}} \end{bmatrix} = \begin{bmatrix} -\frac{1}{f(0)} \\ 0 \end{bmatrix}$$

$$\frac{\partial \mathcal{L}}{\partial a_1^{(2)}} = \frac{\partial}{\partial a_1^{(2)}} \left(-\sum_{i=1}^2 x_i \cdot \log(a_i^{(2)}) \right)$$

$$= -x_1 \cdot \frac{1}{a_1^{(2)}}$$

$$\nabla_{N^{(2)}} \mathcal{L} = \begin{bmatrix} \frac{\partial}{\partial N_1^{(2)}} f(N_1^{(2)}) \cdot \frac{\partial}{\partial N_1^{(2)}} (N_1^{(2)}) & \frac{\partial}{\partial N_2^{(2)}} f(N_2^{(2)}) \cdot \frac{\partial}{\partial N_2^{(2)}} (N_2^{(2)}) \end{bmatrix} = \begin{bmatrix} f(N_1^{(2)})(1-f(N_1^{(2)})) & -f(N_2^{(2)})f(N_1^{(2)}) \\ -f(N_1^{(2)})f(N_2^{(2)}) & f(N_2^{(2)})(1-f(N_2^{(2)})) \end{bmatrix}$$

$$\alpha^{(3)} = F(N^{(3)})$$

$$a^{(3)} = \begin{pmatrix} f(N_1^{(3)}) \\ f(N_2^{(3)}) \end{pmatrix}$$

$$\frac{\partial}{\partial x_i} f(x_i) = \frac{\partial}{\partial x_i} \left(\frac{e^{x_i}}{\sum_{k=1}^n e^{x_k}} \right) = \frac{\left(\frac{\partial e^{x_i}}{\partial x_i} \right) \left(\sum_{k=1}^n e^{x_k} \right) - \left(e^{x_i} \right) \left(\sum_{k=1}^n \frac{\partial e^{x_k}}{\partial x_i} \right)}{\left(\sum_{k=1}^n e^{x_k} \right)^2}, \quad \frac{\partial e^{x_k}}{\partial x_i} = \begin{cases} 0, & i \neq k \\ 1, & i = k \end{cases}$$

$$\frac{\partial}{\partial x_i} f(x_j) = \begin{cases} \frac{e^{x_j} \cdot \sum_{k=1}^n e^{x_k} - e^{x_j}}{\left(\sum_{k=1}^n e^{x_k} \right)^2}, & i=j \rightarrow f(x_j)(1-f(x_j)) \\ -\frac{e^{x_j} \cdot e^{x_i}}{\left(\sum_{k=1}^n e^{x_k} \right)^2}, & i \neq j \\ \hookrightarrow -f(x_j) \cdot f(x_i) \end{cases}$$

$$\frac{\partial}{\partial x_i} f(x_j) = \begin{cases} f(x_j)(1-f(x_j)), & i=j \\ -f(x_j) \cdot f(x_i), & i \neq j \end{cases}$$

$$\nabla_{a^{(2)}} N^{(2)} = \begin{bmatrix} \frac{\partial N_1^{(2)}}{\partial a_1^{(2)}} & \frac{\partial N_2^{(2)}}{\partial a_1^{(2)}} \\ \frac{\partial N_1^{(2)}}{\partial a_2^{(2)}} & \frac{\partial N_2^{(2)}}{\partial a_2^{(2)}} \end{bmatrix} = \begin{bmatrix} w_{11}^{(2)} & w_{21}^{(2)} \\ w_{12}^{(2)} & w_{22}^{(2)} \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\frac{\partial N_j^{(2)}}{\partial a_i^{(2)}} = w_{ji}^{(2)}$$

$$\nabla_{N^{(2)}} \mathcal{L} = \begin{bmatrix} \frac{\partial \mathcal{L}}{\partial N_1^{(2)}} & \frac{\partial \mathcal{L}}{\partial N_2^{(2)}} \end{bmatrix} = \begin{bmatrix} f(N_1^{(2)})(1-f(N_1^{(2)})) & -f(N_2^{(2)})f(N_1^{(2)}) \\ -f(N_1^{(2)})f(N_2^{(2)}) & f(N_2^{(2)})(1-f(N_2^{(2)})) \end{bmatrix}$$

$$\nabla_{a^{(2)}} N^{(2)} = \begin{bmatrix} w_{11}^{(2)} & w_{21}^{(2)} \\ w_{12}^{(2)} & w_{22}^{(2)} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\nabla_{N^{(3)}} \mathcal{L} = \begin{bmatrix} \frac{\partial \mathcal{L}}{\partial N_1^{(3)}} & \frac{\partial \mathcal{L}}{\partial N_2^{(3)}} & \frac{\partial \mathcal{L}}{\partial N_3^{(3)}} \end{bmatrix} = \begin{bmatrix} f(N_1^{(3)})(1-f(N_1^{(3)})) & -f(N_2^{(3)})f(N_1^{(3)}) & -f(N_3^{(3)})f(N_1^{(3)}) \\ -f(N_1^{(3)})f(N_2^{(3)}) & f(N_2^{(3)})(1-f(N_2^{(3)})) & -f(N_3^{(3)})f(N_2^{(3)}) \\ -f(N_1^{(3)})f(N_3^{(3)}) & -f(N_2^{(3)})f(N_3^{(3)}) & f(N_3^{(3)})(1-f(N_3^{(3)})) \end{bmatrix}$$

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$$\begin{aligned} \delta^{[3]} &= \nabla_{N^{[3]}} \epsilon = \nabla_{N^{[3]}}^{\alpha^{[3]}} \cdot \nabla_{\alpha^{[3]}} \epsilon = \begin{bmatrix} f(N_1^{[3]})(1-f(N_1^{[3]})) & -f(N_2^{[3]})f(N_1^{[3]}) \\ -f(N_1^{[3]})f(N_2^{[3]}) & f(N_2^{[3]})(1-f(N_2^{[3]})) \end{bmatrix} \begin{bmatrix} -\frac{1}{f(0)} \\ 0 \end{bmatrix} \\ &= f(0)(1-f(0)) \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} -\frac{1}{f(0)} \\ 0 \end{bmatrix} = \cancel{f(0)(1-f(0))}(-1) \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} f(0)-1 \\ 1-f(0) \end{bmatrix} \\ &= \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \end{aligned}$$

$$\nabla_{w^{[3]}} \epsilon = \delta^{[3]} \cdot \alpha^{[2]T} = \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \begin{bmatrix} f(2f(6)+f(1)+1) & f(2f(6)+f(1)+1) \end{bmatrix} = \begin{pmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

$$w^{[3]} = w^{[3]} - \eta \cdot \nabla_{w^{[3]}} \epsilon = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} - \begin{pmatrix} -0,025 & -0,025 \\ 0,025 & 0,025 \end{pmatrix} = \begin{pmatrix} 0,025 & 0,025 \\ -0,025 & -0,025 \end{pmatrix} = \begin{bmatrix} -\frac{1}{4} & -\frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} \end{bmatrix}$$

$$\begin{aligned} \delta^{[2]} &= \nabla_{N^{[2]}} \epsilon = \nabla_{N^{[2]}}^{\alpha^{[2]}} \cdot \nabla_{\alpha^{[2]}}^{\beta^{[2]}} \cdot \nabla_{\beta^{[2]}} \epsilon = \begin{bmatrix} f(N_1^{[2]})(1-f(N_1^{[2]})) & -f(N_2^{[2]})f(N_1^{[2]}) \\ -f(N_1^{[2]})f(N_2^{[2]}) & f(N_2^{[2]})(1-f(N_2^{[2]})) \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{2}(1-\frac{1}{2}) & (-\frac{1}{2})\frac{1}{2} \\ (-\frac{1}{2})\frac{1}{2} & \frac{1}{2}(1-\frac{1}{2}) \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \\ &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} \end{aligned}$$

$$\nabla_{w^{[2]}} \epsilon = \delta^{[2]} \cdot \alpha^{[1]T} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \begin{pmatrix} f(6) & f(1) & f(6) \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$w^{[2]} = w^{[2]} - \eta \cdot \nabla_{w^{[2]}} \epsilon = w^{[2]} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

$$\delta^{[1]} = \nabla_{N^{[1]}} \epsilon = \nabla_{N^{[1]}}^{\alpha^{[1]}} \cdot \nabla_{\alpha^{[1]}}^{\beta^{[1]}} \cdot \nabla_{\beta^{[1]}} \epsilon = \begin{bmatrix} f(6)(1-f(6)) & -f(1) \cdot f(6) & -f(6)^2 \\ -f(6)f(1) & f(1)(1-f(1)) & -f(6)f(1) \\ -f(6)^2 & -f(1)f(6) & f(6)(1-f(6)) \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\nabla_{w^{[1]}} \epsilon = \delta^{[1]} \cdot \alpha^{[0]T} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \end{bmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$w^{[1]} = w^{[1]} - \eta \cdot \nabla_{w^{[1]}} \epsilon = w^{[1]} = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

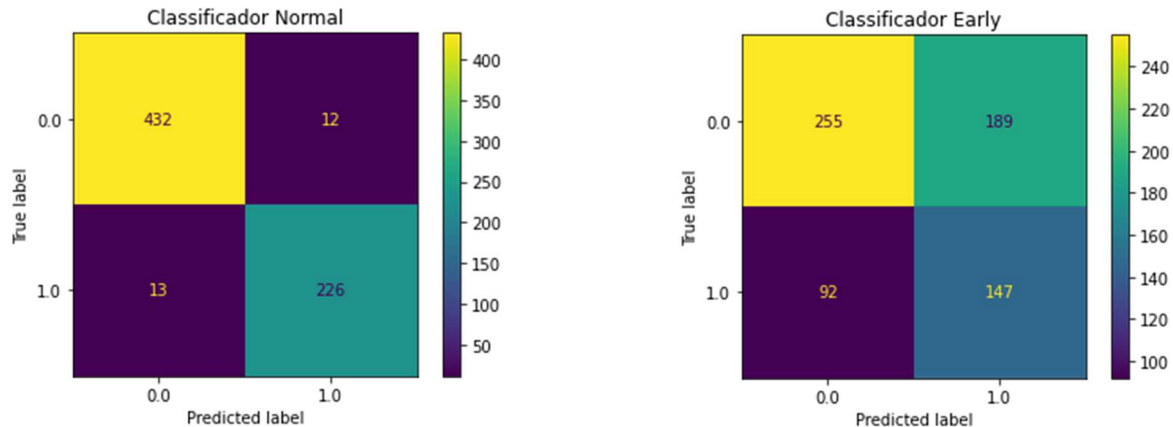
$$b^{[3]} = b^{[3]} - \eta \cdot \delta^{[3]} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} - 0,1 \begin{pmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{pmatrix} = \begin{pmatrix} 0,05 \\ -0,05 \end{pmatrix}$$

$$b^{[2]} = b^{[2]} - \eta \cdot \delta^{[2]} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - 0,1 \begin{pmatrix} 0 \\ 0 \end{pmatrix} = b^{[2]} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$b^{[1]} = b^{[1]} - \eta \cdot \delta^{[1]} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - 0,1 \cdot \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

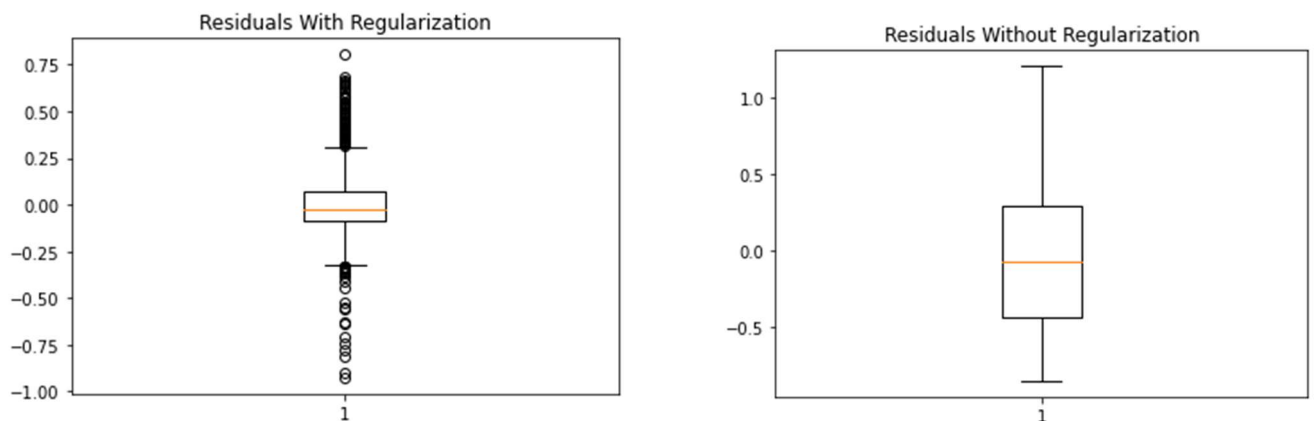
II. Programming and critical analysis

2)



A estratégia de early stopping é geralmente utilizada para evitar overfitting, no entanto, devido à reduzida dimensão da amostra, neste caso prejudica a performance do classificador causando underfitting; para além disso, o early stopping pode levar à paragem num mínimo local, podendo levar assim a classificações erradas.

3)



Utilizar regularização para evitar overfit aos dados de treino; diminuir a learning rate para fazer com que o regressor converja mais lentamente; utilizar gradient descent em vez de stochastic gradient descent, o que permite uma melhor aproximação ao mínimo da função de erro; reduzir o número de features, o que ajuda a diminuir o ruído e evitar overfitting.

III. APPENDIX

```
import numpy as np
import scipy.io.arff
from sklearn.neural_network import MLPClassifier, MLPRegressor
from sklearn.model_selection import KFold
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay #cross entropy
import matplotlib.pyplot as plt

arff = scipy.io.arff

breast_data, breast_meta = arff.loadarff("breast.w.new.arff")
kin_data, kin_meta = arff.loadarff("kin8nm.arff")

breast_targets = np.array(list(el["Class"] for el in breast_data))
breast_X = np.array(np.array(list(list(el[i] for i in range(9)) for el in breast_data)))

kin_targets = np.array(list(el["y"] for el in kin_data))
kin_X = np.array(np.array(list(list(el[i] for i in range(8)) for el in kin_data)))

#2
kfold = KFold(n_splits=5, random_state=0, shuffle=True)

classifier = MLPClassifier(hidden_layer_sizes=(3,2), activation="relu", early_stopping=False,
max_iter=2000)
classifier_early = MLPClassifier(hidden_layer_sizes=(3,2), activation="relu", early_stopping=True,
max_iter=2000)
predictions = []
predictions_early = []
truth = []

for train_i, test_i in kfold.split(breast_X):
    train_data, train_target = breast_X[train_i], breast_targets[train_i]
    test_data, test_target = breast_X[test_i], breast_targets[test_i]
    classifier.fit(train_data, train_target)
    classifier_early.fit(train_data, train_target)
    predictions += list(classifier.predict(test_data))
    predictions_early += list(classifier_early.predict(test_data))
    truth += list(test_target)

conf_m = confusion_matrix(np.array(truth), np.array(predictions), labels=classifier.classes_)
conf_m_early = confusion_matrix(np.array(truth), np.array(predictions_early),
labels=classifier_early.classes_)

disp = ConfusionMatrixDisplay(conf_m, display_labels=classifier.classes_)

disp_e = ConfusionMatrixDisplay(conf_m_early, display_labels=classifier_early.classes_)
```

```
disp.plot()
plt.title("Classificador Normal")
plt.show()

disp_e.plot()
plt.title("Classificador Early")
plt.show()

#3
kfold = KFold(n_splits=5, random_state=0, shuffle=True)

regressor = MLPRegressor(hidden_layer_sizes=(3,2), activation="relu", alpha=10, max_iter=2000)
regressor_no_reg = MLPRegressor(hidden_layer_sizes=(3,2), activation="relu", alpha=0, max_iter=2000)
residuals = []
residuals_no_reg = []

for train_i, test_i in kfold.split(breast_X):
    train_data, train_target = breast_X[train_i], breast_targets[train_i]
    test_data, test_target = breast_X[test_i], breast_targets[test_i]
    regressor.fit(train_data, train_target)
    regressor_no_reg.fit(train_data, train_target)
    residuals += list(test_target - regressor.predict(test_data))
    residuals_no_reg += list(test_target - regressor_no_reg.predict(test_data))

plt.boxplot(np.array(residuals))
plt.title("Residuals With Regularization")
plt.show()

plt.boxplot(np.array(residuals_no_reg))
plt.title("Residuals Without Regularization")
plt.show()
```

END