

ECE 421 Homework 3

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Question 1,

$$D = \{(x_i, y_i)\}_{n=1}^N$$

$$x_n \in \mathbb{R}^2$$

$$y_n \in \{+1, -1\}$$

Bias Term Set To Zero

w with same dimension as x

$$h(x) = \frac{e^{w^T x}}{1 + e^{w^T x}} = \frac{1}{1 + e^{-w^T x}}$$

$$x_1 = \begin{bmatrix} 1 & 1 \end{bmatrix}^T \quad y_1 = 1$$

$$x_2 = \begin{bmatrix} 1 & 0 \end{bmatrix}^T \quad y_2 = -1$$

$$E_{in}(w) = - \sum_{n=1}^N \log[P_w(y_n | x_n)] + \lambda \|w\|_2^2, \quad \lambda > 0$$

$$P_w(y | x) = \begin{cases} h(x) & y = +1 \\ 1 - h(x) & y = -1 \end{cases}$$

Part A,

$$\lambda = 0, \text{ Optimal } w^* = [w_1 \ w_2]^T$$

$$E_{in}(w^*) = - \log[P_{w^*}(y_1 | x_1)] - \log[P_{w^*}(y_2 | x_2)]$$

$$P_{n^*}(y_1 | x_1) = h(x_1) = \frac{1}{1 + e^{-w_1^T x_1}} = \frac{1}{1 + e^{-(w_1 + w_2)}}$$

$$P_{n^*}(y_2 | x_2) = 1 - h(x_2) = 1 - \frac{1}{1 + e^{-w_1^T x_2}}$$

$$= 1 - \frac{1}{1 + e^{-w_1}} = \frac{1 + e^{-w_1}}{1 + e^{-w_1}} - \frac{1}{1 + e^{-w_1}}$$

$$= \frac{e^{-w_1}}{1 + e^{-w_1}} = \frac{1}{1 + e^{w_1}}$$

$$E_n(w^*) = -\log\left(\frac{1}{1 + e^{-(w_1 + w_2)}}\right) - \log\left(\frac{1}{1 + e^{w_1}}\right)$$

$$= \log(1 + e^{-(w_1 + w_2)}) + \log(1 + e^{w_1})$$

To Minimize $E_n(w^*)$, we minimize $e^{-(w_1 + w_2)}$ and e^{w_1}

$e^{-\infty}$ is the minimum, thus

$$w_1 + w_2 = \infty$$

$$w_1 = -\infty$$

$$w_2 = 2\infty$$

$$w^* = \begin{bmatrix} -\infty \\ 2\infty \end{bmatrix}$$

Part B,

$$\|w\|_2 \ll 1$$

$$\log(1 + e^{-y_n w^T x_n}) \approx \log(2) - \frac{1}{2} y_n w^T x_n$$

$$F_{in}(w) = -\log\left(\frac{1}{1 + e^{-(w_1 + w_2)}}\right) - \log\left(\frac{1}{1 + e^{w_1}}\right)$$

$$= \log(1 + e^{-(w_1 + w_2)}) + \log(1 + e^{w_1})$$

$$\approx \log(2) - \frac{1}{2} (w_1 + w_2) + \log(2) + \frac{1}{2} w_1 +$$

$$\lambda \|w\|_2^2$$

$$= 2\log(2) - \frac{1}{2} (w_1 + w_2) + \frac{1}{2} w_1 + \lambda (w_1^2 + w_2^2)$$

$$\frac{\partial F_{in}(w)}{\partial w_1} = -\frac{1}{2} + \frac{1}{2} + 2\lambda w_1 = 0$$

$$w_1^* = 0$$

$$\frac{\partial F_{in}(w)}{\partial w_2} = -\frac{1}{2} + 2\lambda w_2 = 0$$

$$-1 + 4\lambda w_2 = 0$$

$$w_2^* = \frac{1}{4\lambda}$$

$$w^* = \begin{bmatrix} 0 \\ -\frac{1}{4\lambda} \end{bmatrix}$$

Question 2,

$$x = (x_1, x_2) \in \mathbb{R}^2$$

$$y \in \{1, 2, 3\}$$

$$D = \{(x_1, y_1), (x_2, y_2), (x_3, y_3), (x_4, y_4), (x_5, y_5)\}$$

$$x_1 = -1, 0^T, y_1 = 1$$

$$x_2 = 1, 0^T, y_2 = 2$$

$$x_3 = 1, 1^T, y_3 = 2$$

$$x_4 = -1, 1^T, y_4 = 1$$

$$x_5 = 0, 3^T, y_5 = 3$$

$$\Omega = \{w(1), w(2), w(3)\}$$

$$w(i) \in \mathbb{R}^3, i \in \{1, 2, 3\}$$

$$\hat{P}_\Omega(i|x) = \frac{e^{\Gamma w(i)^T \tilde{x}}}{\sum_{j=1}^3 e^{\Gamma w(j)^T \tilde{x}}}, i=1, 2, 3$$

$$\tilde{x} = (x_0=1, x_1, x_2)^T \in \mathbb{R}^3, \text{ Augmented Vector}$$

$$E_n(\Omega) = \frac{1}{5} \sum_{n=1}^5 E_n(\Omega)$$

$$E_n(\Omega) = -\log \hat{P}_\Omega(y_n | x_n)$$

$$w(1) = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix}^T$$

$$w(2) = \begin{bmatrix} 0 & 1 & 0 \end{bmatrix}^T$$

$$w(3) = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}^T$$

$$\nabla_{w(1)} \{e_1(\underline{a})\} = \nabla_{w(1)} \left\{ -\log \hat{P}_{\underline{a}}(y_1 | x_1) \right\}$$

$$= \nabla_{w(1)} \left\{ -\log \frac{e^{\underline{w}^{(1)} \hat{x}}}{\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}} \right\}$$

$y_1 = 1$
 $x_1 = [-1, 0]^T$

$$= - \nabla_{w(1)} \left\{ \log \frac{e^{\underline{w}^{(1)} \hat{x}}}{\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}} \right\}$$

$$= - \frac{\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}}{e^{\underline{w}^{(1)} \hat{x}}} \cdot$$

$$\tilde{x} \frac{e^{\underline{w}^{(1)} \hat{x}}}{\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}} - \tilde{x} \frac{e^{\underline{w}^{(1)} \hat{x}}}{\left(\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}\right)^2}$$

$$= - \left(\tilde{x} - \frac{\tilde{x} e^{\underline{w}^{(1)} \hat{x}}}{\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}} \right) = - \tilde{x} \left(1 - \frac{e^{\underline{w}^{(1)} \hat{x}}}{\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}} \right)$$

$$= - \tilde{x} \frac{e^{\underline{w}^{(a)} \hat{x}} + e^{\underline{w}^{(b)} \hat{x}}}{\sum_{j=1}^3 e^{\underline{w}^{(j)} \hat{x}}}$$

$$\nabla_{w(1)} \{e_1(\underline{a})\}$$

$$\nabla_{w(2)} \{e_1(\alpha)\} = \nabla_{w(2)} \left\{ -\log \hat{P}_\alpha(y_1 | x_1) \right\}$$

$$= \nabla_{w(2)} \left\{ -\log \frac{e^{\underline{[w^{(1)}] \hat{x}}}}{\sum_{j=1}^3 e^{\underline{[w^{(j)}] \hat{x}}}} \right\}$$

$y_1 = 1$
 $x_1 = [-1, 0]^T$

$$= - \nabla_{w(2)} \left\{ \log \frac{e^{\underline{[w^{(1)}] \hat{x}}}}{\sum_{j=1}^3 e^{\underline{[w^{(j)}] \hat{x}}}} \right\}$$

$$= \cancel{\frac{\sum_{j=1}^3 e^{\underline{[w^{(j)}] \hat{x}}}}{e^{\underline{[w^{(1)}] \hat{x}}}}} \cdot \cancel{\frac{e^{\underline{[w^{(1)}] \hat{x}}}}{(\sum_{j=1}^3 e^{\underline{[w^{(j)}] \hat{x}}})^2}} \sim \frac{\hat{x} e^{\underline{[w^{(2)}] \hat{x}}}}{\sum_{j=1}^3 e^{\underline{[w^{(j)}] \hat{x}}}}$$

$$= \frac{\hat{x} e^{\underline{[w^{(2)}] \hat{x}}}}{\sum_{j=1}^3 e^{\underline{[w^{(j)}] \hat{x}}}}$$

$$\nabla_{w(2)} \{e_1(\alpha)\}$$

$$\nabla_{\underline{w(3)}} \{e_1(\underline{a})\} = \nabla_{\underline{w(3)}} \left\{ -\log \hat{P}_{\underline{a}}(y_1 | \underline{x}_1) \right\}$$

$$= \nabla_{\underline{w(3)}} \left\{ -\log \frac{e^{\underline{w^{(1)}}^T \hat{\underline{x}}}}{\sum_{j=1}^3 e^{\underline{w^{(j)}}^T \hat{\underline{x}}}} \right\}$$

$y_1 = 1$
 $\underline{x}_1 = [-1, 0]^T$

$$= - \nabla_{\underline{w(3)}} \left\{ \log \frac{e^{\underline{w^{(1)}}^T \hat{\underline{x}}}}{\sum_{j=1}^3 e^{\underline{w^{(j)}}^T \hat{\underline{x}}}} \right\}$$

$$= \cancel{\frac{\sum_{j=1}^3 e^{\underline{w^{(j)}}^T \hat{\underline{x}}}}{e^{\underline{w^{(1)}}^T \hat{\underline{x}}}}} \cdot \frac{e^{\underline{w^{(1)}}^T \hat{\underline{x}}}}{\left(\sum_{j=1}^3 e^{\underline{w^{(j)}}^T \hat{\underline{x}}}\right)^2}} \sim \frac{e^{\underline{w^{(3)}}^T \hat{\underline{x}}}}{\sum_{j=1}^3 e^{\underline{w^{(j)}}^T \hat{\underline{x}}}}$$

$$= \frac{\tilde{\lambda} e^{\underline{w^{(3)}}^T \hat{\underline{x}}}}{\sum_{j=1}^3 e^{\underline{w^{(j)}}^T \hat{\underline{x}}}}$$

$$\nabla_{\underline{w(3)}} \{e_1(\underline{a})\}$$