

Q1

$$J(w_1, w_2, b) = \frac{1}{4} \sum (x^{(i)T} w + b - y^{(i)})^2$$

$$= \frac{1}{4} ([ [0 \ 0] \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} + b ] - 0 )^2 + [ [1 \ 1] \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} + b ] - 0 ]^2 + [ [1 \ 0] \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} + b ] - 1 ]^2 + [ [0 \ 1] \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} + b ] - 1 ]^2$$

$$J = \frac{1}{4} ([b]^2 + [w_1 + w_2 + b]^2 + [w_1 + b - 1]^2 + [w_2 + b - 1]^2)$$

$$\frac{dJ}{dw_1} = \frac{1}{4} (2 \cdot 1 [w_1 + w_2 + b] + 2 \cdot 1 [w_1 + b - 1])$$

$$\frac{dJ}{dw_1} = \frac{1}{2} (w_1 + w_2 + b + w_1 + b - 1) = \frac{1}{2} (2w_1 + w_2 + 2b - 1) = 0$$

$$\frac{dJ}{dw_2} = \frac{1}{4} ([w_1 + w_2 + b]^2 + [w_2 + b - 1]^2)$$

$$\frac{dJ}{dw_2} = \frac{1}{4} (2 \cdot 1 [w_1 + w_2 + b] + 2 \cdot 1 [w_2 + b - 1])$$

$$\frac{dJ}{dw_2} = \frac{1}{2} (w_1 + w_2 + b + w_2 + b - 1) = \frac{1}{2} (w_1 + 2w_2 + 2b - 1) = 0$$

$$\frac{dJ}{db} = \frac{1}{4} ([b]^2 + [w_1 + w_2 + b]^2 + [w_1 + b - 1]^2 + [w_2 + b - 1]^2)$$

$$\frac{dJ}{db} = \frac{1}{4} (2 \cdot 1 [b] + 2 \cdot 1 [w_1 + w_2 + b] + 2 \cdot 1 [w_1 + b - 1] + 2 \cdot 1 [w_2 + b - 1])$$

$$\frac{dJ}{db} = \frac{1}{2} (b + w_1 + w_2 + b + w_1 + b - 1 + w_2 + b - 1)$$

$$\frac{dJ}{db} = \frac{1}{2} (2w_1 + 2w_2 + 4b - 2) = 0$$

$$\frac{1}{2} (2w_1 + w_2 + 2b - 1) = \frac{1}{2} (w_1 + 2w_2 + 2b - 1)$$

$$\begin{array}{r} 2w_1 + w_2 + 2b - 1 = w_1 + 2w_2 + 2b - 1 \\ \underline{-w_1 \quad -w_2} \qquad \qquad \underline{-w_1 \quad -w_2} \\ w_1 + 2b - 1 = w_2 + 2b - 1 \\ \underline{-2b + 1} \qquad \qquad \underline{-2b + 1} \\ w_1 = w_2 \end{array}$$

$$\frac{1}{2}(2w_1 + 2w_2 + 4b - 2) = 0$$

$$\frac{1}{2}(4w_1 + 4b - 2) = 0$$

because  $w_1 = w_2$

$$2w_1 + 2b - 1 = 0$$

$$w_1 = \frac{1-2b}{2}$$

$$\frac{1}{2}(2w_2 + w_1 + 2b - 1) = 0 \quad \text{because } w_1 = w_2$$

$$\frac{1}{2}(3w_1 + 2b - 1) = 0$$

$$\frac{1}{2}\left(3\left(\frac{1-2b}{2}\right) + 2b - 1\right) = 0$$

$$\frac{1}{2}\left(\frac{3-6b}{2} + 2b - 1\right)$$

$$\frac{3}{4} - \frac{6b}{4} + b - \frac{1}{2} = 0$$

$$-\frac{6b}{4} + b + \frac{1}{4} = 0$$

$$-\frac{3b}{2} + b = -\frac{1}{4}$$

$$-\frac{1}{2}b = -\frac{1}{4}$$

$$-2 \quad -2$$

$$b = \frac{1}{2}$$

$$2w_1 + 2\left(\frac{1}{2}\right) - 1 = 0$$

$$w_1 = w_2 = 0$$

$$\alpha \mathbf{w}^T \mathbf{w} = \alpha \mathbf{w}^T \mathbf{I} \mathbf{w} = \alpha \mathbf{w}^T \mathbf{S} \mathbf{w} \quad \textcircled{Q3}$$

$$\mathbf{S} = \begin{bmatrix} S_1 & S_2 \\ S_3 & S_4 \end{bmatrix}$$

$$\alpha [\mathbf{w}_1, \mathbf{w}_2] \begin{bmatrix} S_1 & S_2 \\ S_3 & S_4 \end{bmatrix} \begin{bmatrix} \mathbf{w}_1 \\ \mathbf{w}_2 \end{bmatrix}$$

$$\alpha \begin{bmatrix} \mathbf{w}_1 S_1 + \mathbf{w}_2 S_3 & \mathbf{w}_1 S_2 + \mathbf{w}_2 S_4 \end{bmatrix} \begin{bmatrix} \mathbf{w}_1 \\ \mathbf{w}_2 \end{bmatrix}$$

$$\mathbf{w}_1 (\mathbf{w}_1 S_1 + \mathbf{w}_2 S_3) + \mathbf{w}_2 (\mathbf{w}_1 S_2 + \mathbf{w}_2 S_4)$$

$$\alpha [\mathbf{w}_1^2 S_1 + \mathbf{w}_1 \mathbf{w}_2 S_3 + \mathbf{w}_1 \mathbf{w}_2 S_2 + \mathbf{w}_2^2 S_4]$$

$$\alpha [\mathbf{w}_1^2 S_1 + \mathbf{w}_1 \mathbf{w}_2 (S_3 + S_2) + \mathbf{w}_2^2 S_4]$$

$$\mathbf{w}_1^2 - 2\mathbf{w}_1 \mathbf{w}_2 + \mathbf{w}_2^2 \quad (\mathbf{w}_1 - \mathbf{w}_2)^2$$

$$S_1 = S_4 = 1$$

$$S_2 = S_3 = -1$$

Conditional probability

Q4

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Student state based  
off observable state

$$P(x_t | y_1, \dots, y_t) = \frac{P(x_t, y_1, \dots, y_t)}{P(y_1, \dots, y_t)}$$

$$P(x_t | y_1, \dots, y_t) \propto P(x_t, y_1, \dots, y_t)$$

$$P(A|B) = \frac{P(A, B)}{P(B)} \propto P(A, B) \quad P(A|B) = P(A, B)$$

not function of A

$$P(a | b, c) \propto P(b | a, c) P(a, c) \propto \text{bayes rule:}$$

$$P(\underbrace{x_t}_{y_1, \dots, y_{t-1}} | \underbrace{y_1, \dots, y_t}_{y_1, \dots, y_{t-1}}) \propto P(y_t | x_t, y_1, \dots, y_{t-1}) P(x_t | y_1, \dots, y_{t-1})$$

based off total probability (example):

$$P(q) = \sum_r P(q, r)$$

Using total probability:

$$P(y_t | x_t) \sum_{x_{t-1}} P(x_t, x_{t-1} | y_1, \dots, y_{t-1})$$

Using conditional probability:

$$P(y_t | x_t) \sum_{x_{t-1}} P(x_t | x_{t-1}, y_1, \dots, y_{t-1}) P(x_{t-1} | y_1, \dots, y_{t-1}) \rightarrow$$

Using Markov Property and Conditional independence:

$$P(y_t | x_t) \sum_{x_{t+1}} \underbrace{P(x_t | x_{t+1})}_{\text{Markov}} P(x_{t+1} | y_1, \dots, y_{t+1})$$

Q5

$$P(y | x) = \mathcal{N}(u = x^T w, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(y - x^T w)^2}{2\sigma^2}\right)$$

$$P(D | u = x^T w, \sigma^2) = \log \prod_{i=1}^n P(y^{(i)} | x^{(i)}, w, \sigma^2) \\ = \sum_{i=1}^n \log P(y^{(i)} | x^{(i)}, w, \sigma^2)$$

$$= \sum_{i=1}^n \log\left(\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x^{(i)T} w - y^{(i)})^2}{2\sigma^2}\right)\right)$$

$$= \sum_{i=1}^n - (x^{(i)T} w - y^{(i)})^2 - \log(\sqrt{2\pi\sigma^2})$$

$$\therefore \frac{1}{2\sigma^2} \sum_{i=1}^n (x^{(i)T} w - y^{(i)})^2 - \log(\sqrt{2\pi\sigma^2}) \\ \frac{1}{\log(\sqrt{2\pi}) + \log(\sigma)}$$

$$= \frac{1}{2\sigma^2} \sum_{i=1}^n \left( (x^{(i)T} w - y^{(i)})^2 - \log(\sigma) \right) + C$$

$$= -\frac{1}{2\sigma^2} \sum_{i=1}^n \left( (x^T w - y)^2 - \log(\sigma) \right) + C$$

$$= -\frac{1}{2\sigma^2} \sum_{i=1}^n (x^T w - y)^2 - n \log(\sigma) + C$$

$$\nabla_w \log P(D | w, \sigma^2) = \nabla_w \left( -\frac{1}{2\sigma^2} \sum_{i=1}^n (x^T w - y)^2 - n \log(\sigma) + C \right)$$

$$= \nabla_w \left( \frac{1}{2\sigma^2} \sum_{i=1}^n (x^T w - y)^2 + n \log(\sigma) - C \right) \leftarrow \text{factor } -1$$

$$0 = \frac{1}{\sigma^2} \sum_{i=1}^n x(x^T w - y) \quad \leftarrow \text{multiply } \sigma^2 \text{ to both sides}$$

$$0 = \sum_{i=1}^n x x^T w - x y$$

$$\sum_{i=1}^n x y = \sum_{i=1}^n x x^T w$$

$$w = \left( \sum_{i=1}^n x x^T \right)^{-1} \left( \sum_{i=1}^n x y \right)$$

→

$$= \nabla_{\sigma} \left( -\frac{1}{2\sigma^2} \sum_{i=1}^n (x^T w - y)^2 + n \log(\sigma) - C \right) \leftarrow \text{factor } -1$$

$$= -\frac{1}{\sigma^3} \sum_{i=1}^n (x^T w - y)^2 + \frac{n}{\sigma}$$

$$-\frac{n}{\sigma^2} = -\frac{1}{\sigma^3} \sum_{i=1}^n (x^T w - y)^2$$

$$\sigma^2 = \frac{1}{n} \sum_{i=1}^n (x^T w - y)^2$$