

Q1.

Proof:

$$\begin{aligned}
p(y = 1|x) &= \frac{p(x|y = 1) * p(y = 1)}{p(x)} = \frac{p(x|y = 1) * p(y = 1)}{p(x|y = 1) * p(y = 1) + p(x|y = 0) * p(y = 0)} \\
&= \frac{1}{1 + \frac{p(x|y = 0) * p(y = 0)}{p(x|y = 1) * p(y = 1)}} = \frac{1}{1 + \exp\left(\ln\left(\frac{p(x|y = 0) * p(y = 0)}{p(x|y = 1) * p(y = 1)}\right)\right)} \\
&= \frac{1}{1 + \exp\left(\ln\left(\frac{p(x|y = 0)}{p(x|y = 1)}\right) + \ln\left(\frac{p(y = 0)}{p(y = 1)}\right)\right)} = \frac{1}{1 + \exp\left(\ln\left(\frac{p(x|y = 0)}{p(x|y = 1)}\right) + \ln\left(\frac{1 - \alpha}{\alpha}\right)\right)} \\
&= \frac{1}{1 + \exp\left(\sum_{i=1}^D \ln\left(\frac{p(x_i|y = 0)}{p(x_i|y = 1)}\right) + \ln\left(\frac{1 - \alpha}{\alpha}\right)\right)} \quad (\text{by conditionally independent}) \\
&= \frac{1}{1 + \exp\left(\sum_{i=1}^D \ln\left(\frac{\frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2}(x_i - \mu_{i0})^2\right)}{\frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2}(x_i - \mu_{i1})^2\right)}\right) + \ln\left(\frac{1 - \alpha}{\alpha}\right)\right)} \\
&= \frac{1}{1 + \exp\left(\sum_{i=1}^D \ln\left(\frac{\exp\left(-\frac{1}{2\sigma^2}(x_i - \mu_{i0})^2\right)}{\exp\left(-\frac{1}{2\sigma^2}(x_i - \mu_{i1})^2\right)}\right) + \ln\left(\frac{1 - \alpha}{\alpha}\right)\right)} \\
&= \frac{1}{1 + \exp\left(\sum_{i=1}^D \ln\left(\exp\left(\frac{1}{2\sigma^2}(x_i - \mu_{i1})^2 - \frac{1}{2\sigma^2}(x_i - \mu_{i0})^2\right) + \ln\left(\frac{1 - \alpha}{\alpha}\right)\right)\right)} \\
&= \frac{1}{1 + \exp\left(\sum_{i=1}^D \frac{(x_i - \mu_{i1})^2 - (x_i - \mu_{i0})^2}{2\sigma^2} + \ln\left(\frac{1 - \alpha}{\alpha}\right)\right)} \\
&= \frac{1}{1 + \exp\left(\sum_{i=1}^D \frac{(\mu_{i1}^2 - \mu_{i0}^2) + 2x_i(\mu_{i0} - \mu_{i1})}{2\sigma^2} + \ln\left(\frac{1 - \alpha}{\alpha}\right)\right)} \\
&= \frac{1}{1 + \exp\left(\sum_{i=1}^D \frac{(\mu_{i0} - \mu_{i1})}{\sigma^2} x_i + \ln\left(\frac{1 - \alpha}{\alpha}\right) + \sum_{i=1}^D \frac{(\mu_{i1}^2 - \mu_{i0}^2)}{2\sigma^2}\right)}
\end{aligned}$$

From the equation above, it shows that $p(y = 1|x)$ takes the form of a logistic function:

$$p(y = 1|x) = \sigma(w^T x + b) = \frac{1}{1 + \exp(-\sum_{i=1}^D w_i x_i - b)}$$

$$\text{where } w_i = \frac{(\mu_{i0} - \mu_{i1})}{\sigma^2} \quad b = -\left(\ln\left(\frac{1-\alpha}{\alpha}\right) + \sum_{i=1}^D \frac{(\mu_{i1}^2 - \mu_{i0}^2)}{2\sigma^2}\right)$$

Q2.

$$\begin{aligned} E(w, b) &= -\ln\left(\prod_{i=1}^N p(y^i = 0|x^i)^{1-y^i} * p(y^i = 1|x^i)^{y^i}\right) \\ &= -\ln\left(\prod_{i=1}^N p(y^i = 0|x^i)^{1-y^i}\right) - \ln\left(\prod_{i=1}^N p(y^i = 1|x^i)^{y^i}\right) \\ &= -(1-y^i) \ln\left(\prod_{i=1}^N p(y^i = 0|x^i)\right) - y^i * \ln\left(\prod_{i=1}^N p(y^i = 1|x^i)\right) \\ &= \sum_{i=0}^N -(1-y^i) \ln(p(y^i = 0|x^i)) - \sum_{i=0}^N y^i \ln(p(y^i = 1|x^i)) \end{aligned}$$

let $z^i = w^T x_i + b$, then

$$\begin{aligned} &= \sum_{i=0}^N (1-y^i) \ln(1 + \exp(-z^i)) + \sum_{i=0}^N (1-y^i) \ln(\exp(z^i)) + \sum_{i=0}^N y^i \ln(1 + \exp(-z^i)) \\ &= \sum_{i=0}^N \ln(1 + \exp(-z^i)) + \sum_{i=0}^N (1-y^i) z^i \end{aligned}$$

Then derive expressions for the derivatives of E with respect to each of the model parameters:

$$\begin{aligned} \frac{E(w, b)}{\partial b} &= -\sum_{i=0}^N \frac{\exp(-(wx^i + b))}{1 + \exp(-(wx^i + b))} + \sum_{i=0}^N (1-y^i) = -\sum_{i=0}^N \frac{\exp(-z^i)}{1 + \exp(-z^i)} + \sum_{i=0}^N (1-y^i) \\ \frac{E(w, b)}{\partial w_j} &= -\sum_{i=0}^N x_j^i \frac{\exp(-(wx^i + b))}{1 + \exp(-(wx^i + b))} + \sum_{i=0}^N (1-y^i) x_j^i \\ &= x_j^i \left(-\sum_{i=0}^N \frac{\exp(-z^i)}{1 + \exp(-z^i)} + \sum_{i=0}^N (1-y^i) \right) \end{aligned}$$

Q3.

$$p(w, b|D) \propto p(D|w, b)p(w, b)$$

$$p(w, b) = \prod_{i=1}^D N(w_i | 0, 1/\lambda) N(b | 0, 1/\lambda)$$

$$p(D|w, b) = \prod_{i=1}^N \frac{1}{(1 + \exp(-z))} \frac{\exp(-z)^{y^i}}{(1 + \exp(-z))^{1-y^i}} \quad \text{let } z = wx + b$$

$$p(w, b|D) \propto \left(\prod_{i=1}^N \frac{1}{(1 + \exp(-z))} \frac{\exp(-z)^{y^i}}{(1 + \exp(-z))^{1-y^i}} \right) \prod_{i=1}^D N(w_i | 0, 1/\lambda) N(b | 0, 1/\lambda)$$

$$L(w, b) = -\ln \left(\left(\prod_{i=1}^N \frac{1}{(1 + \exp(-z))} \frac{\exp(-z)^{y^i}}{(1 + \exp(-z))^{1-y^i}} \right) \left(\prod_{i=1}^D N(w_i | 0, 1/\lambda) N(b | 0, 1/\lambda) \right) \right)$$

$$= -\ln \left(\prod_{i=1}^N \frac{1}{(1 + \exp(-z))} \frac{\exp(-z)^{y^i}}{(1 + \exp(-z))^{1-y^i}} \right) - \ln \left(\prod_{i=1}^D N(w_i | 0, 1/\lambda) N(b | 0, 1/\lambda) \right)$$

$$= E(w, b) - \sum_{i=0}^D \ln \left(N(w_i | 0, 1/\lambda) \right) - \ln \left(N(b | 0, 1/\lambda) \right)$$

$$= E(w, b) - \sum_{i=0}^D -\frac{w_i^2 * \lambda}{2} - D * \ln \left(\frac{\sqrt{\lambda}}{\sqrt{2\pi}} \right) - \left(-\frac{b^2 * \lambda}{2} \right) - \ln \left(\frac{\sqrt{\lambda}}{\sqrt{2\pi}} \right)$$

$$= E(w, b) + \frac{\lambda}{2} \sum_{i=0}^D w_i^2 + \left(\frac{\lambda}{2} \right) b^2 - (D + 1) \ln \left(\frac{\sqrt{\lambda}}{\sqrt{2\pi}} \right)$$

$$= E(w, b) + \frac{\lambda}{2} \sum_{i=0}^D w_i^2 + \frac{\lambda}{2} b^2 + c(\lambda)$$

Then derive expressions for the derivatives of L with respect to each of the model parameters:

$$\frac{\partial L(w, b)}{\partial b} = - \sum_{i=0}^N \frac{\exp(-z^i)}{1 + \exp(-z^i)} + \sum_{i=0}^N (1 - y^i) + \lambda b$$

$$\begin{aligned}
\frac{L(w, b)}{\partial w_j} &= - \sum_{i=0}^N x_j^i \frac{\exp(-(wx^i + b))}{1 + \exp(-(wx^i + b))} + \sum_{i=0}^N (1 - y^i) x_j^i + \lambda \sum_{i=0}^D w_i \\
&= x_j^i \left(- \sum_{i=0}^N \frac{\exp(-z^i)}{1 + \exp(-z^i)} + \sum_{i=0}^N (1 - y^i) \right) + \lambda \sum_{i=0}^D w_i
\end{aligned}$$

Q4.