

$$\begin{aligned}
 a) \quad \sigma'(x) &= \frac{d}{dx} \left(\frac{1}{1+e^{-x}} \right) = \frac{d}{dx} (1+e^{-x})^{-1} \\
 &= e^{-x} (1+e^{-x})^{-2} \\
 &= \left(\frac{1}{1+e^{-x}} \right) \left(\frac{e^{-x}}{1+e^{-x}} \right) \\
 &= \sigma(x) \left(\frac{1+e^{-x}-1}{1+e^{-x}} \right) \\
 &= \sigma(x) \left(1 - \frac{1}{1+e^{-x}} \right) \\
 &= \sigma(x) [1 - \sigma(x)]
 \end{aligned}$$

b) Negative log likelihood for regression is

$$\begin{aligned}
 n \ell(\theta) &= - \sum_i y_i \sigma(\theta^T x_i) \sigma'(\theta^T x_i) + (1 - y_i) \frac{1}{1 - \sigma(\theta^T x_i)} (-\sigma'(\theta^T x_i)) \\
 &= \sum_i (\mu_i - y_i) x_i = X^T (\mu - y)
 \end{aligned}$$

c) Using this, we can find the Hessian matrix.

$$\begin{aligned}
 H_\theta &= \nabla_\theta (\nabla_\theta n \ell(\theta))^T = \nabla_\theta [X^T (\mu - y)]^T = X^T \text{diag}(\mu(1-\mu)) X \\
 &= X^T S X
 \end{aligned}$$

$$Z = \int_{\mathbb{R}} P(x; \sigma^2) dx = \int_{\mathbb{R}} \frac{1}{Z} \exp\left(-\frac{x^2}{2\sigma^2}\right) dx = \frac{1}{Z} \int_{\mathbb{R}} \exp\left(-\frac{x^2}{2\sigma^2}\right) dx = 1$$

$$\begin{aligned}
 Z^2 &= \int_{\mathbb{R}} \exp\left(-\frac{x^2}{2\sigma^2}\right) dx \int_{\mathbb{R}} \exp\left(-\frac{y^2}{2\sigma^2}\right) dy = \iint_{\mathbb{R}^2} \exp\left(-\frac{x^2+y^2}{2\sigma^2}\right) dx dy \\
 &= -2\pi\sigma^2(0-1) = 2\pi\sigma^2
 \end{aligned}$$

$$\text{Hence } Z^2 = 2\pi\sigma^2 \Rightarrow Z = \sqrt{2\pi\sigma^2} = \sqrt{2\pi}\sigma$$