Question 1.

a.

$$\mathcal{L}(\mu, \sigma | x_1, x_2, \dots, x_n) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(x_i - \mu)^2}$$

$$= \left(\frac{1}{\sqrt{2\pi\sigma^2}}\right)^n e^{-\sum_{i=1}^n \frac{1}{2\sigma^2}(x_i - \mu)^2}$$

$$= \left(\frac{1}{\sqrt{2\pi\sigma^2}}\right)^n e^{-\frac{1}{2\sigma^2}\sum_{i=1}^n (x_i - \mu)^2}$$

Thus

$$l(\mu, \sigma | x_1, x_2, \dots, x_n) = -\frac{n}{2} \log(2\pi) - \frac{n}{2} \log(\sigma^2) - \frac{1}{2\sigma^2} \sum_{i=1}^{n} (x_i - \mu)^2$$

and

$$\frac{dl}{d\mu} = \frac{1}{2\sigma^2} \sum_{i=1}^n 2(x_i - \mu) = \frac{1}{\sigma^2} \left(-n\mu + \sum_{i=1}^n x_i \right) = 0 \implies \hat{\mu} = \frac{1}{n} \sum_{i=1}^n x_i$$

b.

We also have that

$$\frac{dl}{d\sigma^2} = -\frac{n}{2\sigma^2} + \frac{1}{2\sigma^4} \sum_{i=1}^n (x_i - \hat{\mu})^2 = 0 \implies \hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \hat{\mu})^2$$

2.

a.

$$P_X^2 = X(X^T X)^{-1} \underbrace{X^T X(X^T X)^{-1}}_{I_p} X^T = X(X^T X)^{-1} X^T = P_X$$

For any $u = X\beta$,

$$P_X u = P_X \underbrace{X\beta = X(X^T X)^{-1} X^T X}_{I_p} \beta = X\beta = \mu$$

b.

The modified version does not have part b

3.

a.

$$\hat{y} = X\hat{\beta} = X(X^TX)^{-1}X^Ty = P_Xy$$

b.

$$X^{T}y - X^{T}X\hat{\beta} = X^{T}y - X^{T}X(X^{T}X)^{-1}X^{T}y = X^{T}y - X^{T}y = 0$$

c.

Since $X^T y = X^T \hat{y}$, we proved that there is no better solution that can be represented using the column space of X because $X^T \hat{y}$ already has the same value