$$Q = \sum_{i=1}^{N} \sum_{k=1}^{K} -\gamma_{k,i} \log \left[p_{X|Z}(x_i|k) \cdot p_Z(k) \right]$$

$$= \sum_{i=1}^{N} \sum_{k=1}^{K} \gamma_{k,i} \left(\frac{1}{2} \log \left[2\pi \det(\Sigma_k) \right] + \frac{1}{2} (x_i - \mu_k)^T \Sigma_k^{-1} (x_i - \mu_k) - \log \left[p_Z(k) \right] \right)$$

Assuming that Q is convex, find the optimal values for μ_k and Σ_k

HINT: You may use the following general matrix/vector gradient equations (refer to the [matrix cook book by Peterson and Pedersen (2012), p.10-11](https://www.math.uwaterloo.ca/ hwolkowi/matrix-cookbook.pdf):

$$\nabla_x (x - y)^T W(x - y) = 2W(x - y) \tag{1}$$

$$\nabla_W(x-y)^T W(x-y) = (x-y)(x-y)^T \tag{2}$$

$$\nabla_{W^{-1}} \log[\det(W)] = -W$$
 if W is symmetric and positive semi-definite (3)

$$\nabla_{\mu_k} \sum_{i=1}^{N} \sum_{k=1}^{K} \gamma_{k,i} \left(\frac{1}{2} \log[2\pi \det(\Sigma_k)] + \frac{1}{2} (x_i - \mu_k)^T \Sigma_k^{-1} (x_i - \mu_k) - \log[p_Z(k)] \right) \stackrel{!}{=} \vec{0}$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} \left(\nabla_{\mu_k} \frac{1}{2} (x_i - \mu_k)^T \Sigma_k^{-1} (x_i - \mu_k) \right) = 0$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} \left(\Sigma_k^{-1} (x_i - \mu_k) \right) = 0$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} \left(\Sigma_k^{-1} x_i - \Sigma_k^{-1} \mu_k \right) = 0$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} \Sigma_k^{-1} x_i = \sum_{i=1}^{N} \gamma_{k,i} \Sigma_k^{-1} \mu_k$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} x_i = \sum_{i=1}^{N} \gamma_{k,i} x_i = \sum_{i=1}^{N} \gamma_{k,i}$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} x_i = \mu_k$$

Lorem ipsum

$$\nabla_{\Sigma_{k}^{-1}} \sum_{i=1}^{N} \sum_{k=1}^{K} \gamma_{k,i} \left(\frac{1}{2} \log[2\pi \det(\Sigma_{k})] + \frac{1}{2} (x_{i} - \mu_{k})^{T} \Sigma_{k}^{-1} (x_{i} - \mu_{k}) - \log[p_{Z}(k)] \right) \stackrel{!}{=} \vec{0}$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} \left(-\frac{1}{2} \Sigma_{k} + \frac{1}{2} (x_{i} - \mu_{k})^{T} (x_{i} - \mu_{k}) \right) = 0$$

$$\Leftrightarrow \sum_{i=1}^{N} \gamma_{k,i} \Sigma_{k} = \sum_{i=1}^{N} \gamma_{k,i} (x_{i} - \mu_{k})^{T} (x_{i} - \mu_{k})$$

$$\Leftrightarrow \Sigma_{k} \sum_{i=1}^{N} \gamma_{k,i} = \sum_{i=1}^{N} \gamma_{k,i} (x_{i} - \mu_{k})^{T} (x_{i} - \mu_{k})$$

$$\Leftrightarrow \Sigma_{k} = \frac{\sum_{i=1}^{N} \gamma_{k,i} (x_{i} - \mu_{k})^{T} (x_{i} - \mu_{k})}{\sum_{i=1}^{N} \gamma_{k,i}}$$