

Question 1

$$L(\mathbf{w}, \lambda) = \mathbf{w}^T(\mathbf{m}_2 - \mathbf{m}_1) + \lambda(\mathbf{w}^T \mathbf{w} - 1)$$

$$\frac{\partial}{\partial \mathbf{w}} L(\mathbf{w}, \lambda) = \mathbf{m}_2 - \mathbf{m}_1 + 2\lambda \mathbf{w} = 0$$

Hence,

$$\mathbf{w} = -\frac{1}{2\lambda}(\mathbf{m}_2 - \mathbf{m}_1) \propto (\mathbf{m}_2 - \mathbf{m}_1)$$

Question 2

$$\begin{aligned} J(\mathbf{w}) &= \frac{(m_2 - m_1)^2}{s_1^2 + s_2^2} \\ &= \frac{(\mathbf{w}^T \mathbf{m}_2 - \mathbf{w}^T \mathbf{m}_1)^2}{\sum_{n \in \mathcal{C}_1} (\mathbf{w}^T \mathbf{x}_1 - \mathbf{w}^T \mathbf{m}_1) + \sum_{n \in \mathcal{C}_2} (\mathbf{w}^T \mathbf{x}_2 - \mathbf{w}^T \mathbf{m}_2)} \\ &= \frac{\mathbf{w}^T (\mathbf{m}_2 - \mathbf{m}_1) (\mathbf{m}_2 - \mathbf{m}_1)^T \mathbf{w}}{\mathbf{w}^T (\sum_{n \in \mathcal{C}_1} (\mathbf{x}_1 - \mathbf{m}_1) + \sum_{n \in \mathcal{C}_2} (\mathbf{x}_2 - \mathbf{m}_2)) \mathbf{w}} \\ &= \frac{\mathbf{w}^T \mathbf{S}_B \mathbf{w}}{\mathbf{w}^T \mathbf{S}_W \mathbf{w}} \end{aligned}$$

Question 3

For a point ϕ_n from \mathcal{C}_k ,

$$p(\phi_n, \mathcal{C}_k) = p(\mathcal{C}_k) p(\phi_n | \mathcal{C}_k) = \pi_k p(\phi_n | \mathcal{C}_k)$$

so the likelihood function is

$$p(\{\phi_n, \mathbf{t}_n\} | \boldsymbol{\pi}, \boldsymbol{\mu}, \Sigma) = \prod_{n=1}^N \prod_{k=1}^K [\pi_k p(\phi_n | \mathcal{C}_k)]^{t_{nk}}$$

Take the log-likelihood

$$\sum_{n=1}^N \sum_{k=1}^K [t_{nk} \ln \pi_k + t_{nk} \ln p(\phi_n | \mathcal{C}_k)]$$

As $\sum_{k=1}^K \pi_k = 1$, we use Lagrange Multiplier to maximize

$$L(\pi_i, \lambda) = \sum_{n=1}^N \sum_{k=1}^K [t_{nk} \ln \pi_k + t_{nk} \ln p(\phi_n | \mathcal{C}_k)] + \lambda (\sum_{k=1}^K \pi_k - 1)$$

$$\frac{\partial}{\partial \lambda} L(\pi_i, \lambda) = \sum_{k=1}^K \pi_k - 1 = 0$$

$$\frac{\partial}{\partial \pi_i} L(\pi_i, \lambda) = \frac{\sum_{n=1}^N t_{ni}}{\pi_i} + \lambda = 0$$

$$\pi_i = -\frac{\sum_{n=1}^N t_{ni}}{\lambda} = -\frac{N_i}{\lambda}$$

Since

$$\sum_{k=1}^K \pi_k = \sum_{k=1}^K -\frac{N_i}{\lambda} = -\frac{N}{\lambda} = 1$$

then $\lambda = -N$

Hence,

$$\pi_i = \frac{N_i}{N}$$

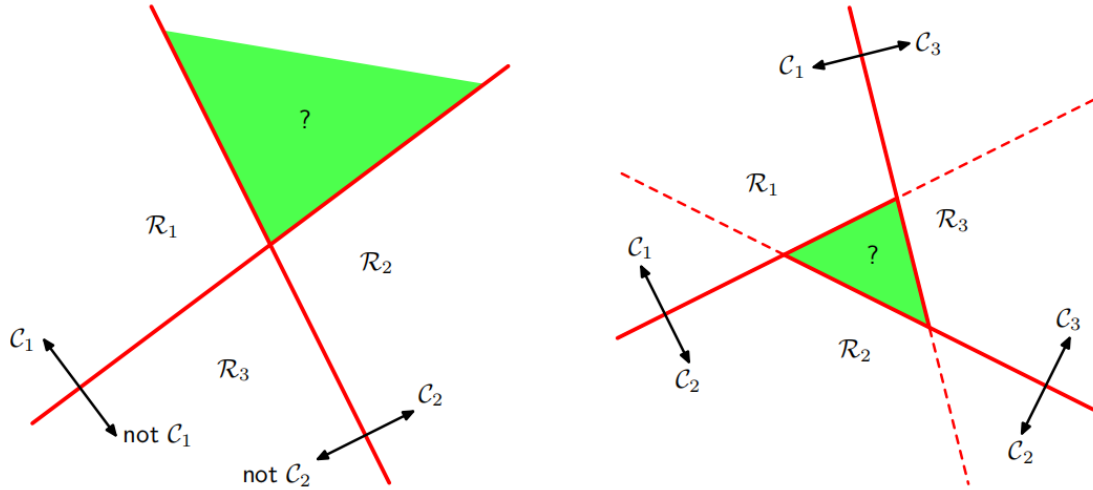
Question 4

$$\begin{aligned} \frac{\partial}{\partial a} \sigma(a) &= \frac{e^{-a}}{(1 + e^{-a})^2} \\ &= \frac{e^{-a}}{1 + e^{-a}} \frac{1}{1 + e^{-a}} \\ &= \frac{e^{-a}}{1 + e^{-a}} \left(1 - \frac{e^{-a}}{1 + e^{-a}}\right) \\ &= \sigma(1 - \sigma) \end{aligned}$$

Question 5

$$\begin{aligned} y_n &= \sigma(a_n) \\ a_n &= \mathbf{w}^T \boldsymbol{\phi}_n \\ \frac{\partial y_n}{\partial \mathbf{w}} &= \frac{\partial y_n}{\partial a_n} \frac{\partial a_n}{\partial \mathbf{w}} = y_n(1 - y_n) \boldsymbol{\phi}_n \\ \nabla \mathbb{E}(\mathbf{w}) &= -\sum_{n=1}^N \left(\frac{t_n}{y_n} \frac{\partial y_n}{\partial \mathbf{w}} - \frac{1 - t_n}{1 - y_n} \frac{\partial y_n}{\partial \mathbf{w}} \right) \\ &= -\sum_{n=1}^N (t_n(1 - y_n) - (1 - t_n)y_n) \boldsymbol{\phi}_n \\ &= \sum_{n=1}^N (y_n - t_n) \boldsymbol{\phi}_n \end{aligned}$$

Question 6



Question 7

Proof by contradiction

If their convex hulls intersect, then

$$\exists \mathbf{y}, \quad s.t. \mathbf{y} = \sum_n \alpha_n \mathbf{x}^n = \sum_m \beta_m \mathbf{z}^m$$

where $\alpha, \beta > 0$ and $\sum_n \alpha_n = \sum_m \beta_m = 1$.

Assume $\{\mathbf{x}^n\}$ and $\{\mathbf{z}^m\}$ are linearly separable, there exists a vector $\hat{\mathbf{w}}$ and a scalar w_0 such that $\hat{\mathbf{w}}^T \mathbf{x}^n + w_0 > 0$ for all \mathbf{x}^n , and $\hat{\mathbf{w}}^T \mathbf{z}^m + w_0 < 0$ for all \mathbf{z}^m .

We figure out

$$\begin{aligned} \sum_n \hat{\mathbf{w}}^T \alpha_n \mathbf{x}^n + \sum_n \alpha_n w_0 &= \sum_n \hat{\mathbf{w}}^T \mathbf{y} + w_0 > 0 \\ \sum_m \hat{\mathbf{w}}^T \beta_m \mathbf{z}^m + \sum_m \beta_m w_0 &= \sum_m \hat{\mathbf{w}}^T \mathbf{y} + w_0 < 0 \end{aligned}$$

It leads a contradiction.

Hence, if their convex hulls intersect, the two sets of points cannot be linearly separable, and conversely that, if they are linearly separable, their convex hulls do not intersect.