

Homework 4 Solutions

HW4 p1

Problem 1: We did this in the lecture, see 4p1.

Problem 2:
$$\mathcal{L}(\theta) = \sum_{n=1}^N \sum_{c=1}^6 \mathbb{I}(y=c) \theta_c$$

note $\theta_6 = 1 - \theta_1 - \dots - \theta_5$
so only five unknowns

$$\begin{aligned} \text{NLL}(\theta) &= -\ln \mathcal{L}(\theta) \\ &= -\sum_{n=1}^N \sum_{c=1}^6 \mathbb{I}(y=c) \cdot \ln(\theta_c) \\ &= -\sum_{n: y_n=1} \ln(\theta_1) - \dots - \sum_{n: y_n=6} \ln(\theta_6) \\ &= -N_1 \ln(\theta_1) - \dots - N_5 \ln(\theta_5) - N_6 \ln(1 - \theta_1 - \dots - \theta_5) \end{aligned}$$

$$\frac{\partial}{\partial \theta_k} \text{NLL}(\theta) = -\frac{N_k}{\theta_k} + \frac{N_6}{1 - \sum_{\ell=1}^5 \theta_\ell} = 0 \quad k=1..5$$

$$\Rightarrow \frac{\theta_k}{N_k} = \frac{1 - \sum_{\ell=1}^5 \theta_\ell}{N_6} \Rightarrow \theta_k + N_k \sum_{\ell=1}^5 \theta_\ell = \frac{N_k}{N_6}$$

this is a linear system. The matrix form is

$$\left(\mathbf{I} + \underbrace{\frac{1}{N_6} \begin{bmatrix} N_1 \\ \vdots \\ N_5 \end{bmatrix}}_{\underline{u}} \cdot \underbrace{\begin{bmatrix} 1, 1, 1, 1, 1 \end{bmatrix}}_{\underline{v}^T} \right) \begin{bmatrix} \theta_1 \\ \vdots \\ \theta_5 \end{bmatrix} = \frac{1}{N_6} \underbrace{\begin{bmatrix} N_1 \\ \vdots \\ N_5 \end{bmatrix}}_{\underline{u}}$$

By the Sherman-Morrison formula $(\mathbf{I} + \underline{u} \underline{v}^T)^{-1} = \mathbf{I} - \gamma \underline{u} \underline{v}^T$ where $\gamma = \frac{1}{1 + \underline{v}^T \underline{u}}$

It follows that

$$\begin{bmatrix} \theta_1 \\ \vdots \\ \theta_5 \end{bmatrix} = \left(\mathbf{I} - \frac{N_6}{N} \frac{1}{N_6} \begin{bmatrix} N_1 \\ \vdots \\ N_5 \end{bmatrix} \cdot \begin{bmatrix} 1, \dots, 1 \end{bmatrix} \right) \cdot \frac{1}{N_6} \begin{bmatrix} N_1 \\ \vdots \\ N_5 \end{bmatrix} \quad \gamma = \frac{1}{1 + \frac{N_1 + \dots + N_5}{N_6}} = \frac{N_6}{N}$$

$$\begin{aligned} \Rightarrow \theta_k &= \frac{N_k}{N_6} - \frac{1}{N} N_k \cdot \sum_{\ell=1}^5 N_\ell \cdot \frac{1}{N_6} \\ &= \frac{N_k}{N_6} \cdot \left(1 - \frac{1}{N} \sum_{\ell=1}^5 N_\ell \right) = \frac{N_k}{N_6} \cdot \frac{1}{N} \underbrace{\left(N - \sum_{\ell=1}^5 N_\ell \right)}_{= N_6} = \frac{N_k}{N} \end{aligned}$$

Note: In sec. 4.2.4 the method of Lagrange multipliers is used to find the optimal $\theta_1, \dots, \theta_6$ under the constraint $\sum \theta_k = 1$. Both methods are good ways to solve this problem.

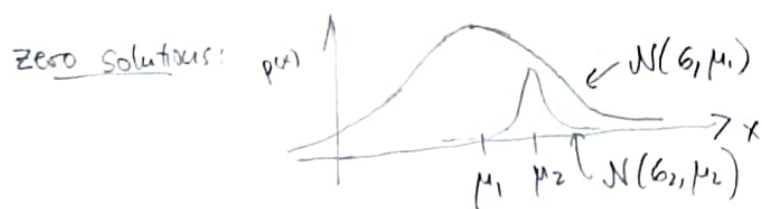
Problem 3 $p(X=x | Y=1) \pi_1 = p(X=x | Y=2) \pi_2$

$$\Rightarrow \frac{1}{\sqrt{2\pi} \sigma_1} \exp\left(-\frac{1}{2}\left(\frac{x-\mu_1}{\sigma_1}\right)^2\right) \pi_1 = \frac{1}{\sqrt{2\pi} \sigma_2} \exp\left(-\frac{1}{2}\left(\frac{x-\mu_2}{\sigma_2}\right)^2\right) \pi_2$$

$$\Rightarrow \ln \frac{\pi_1}{\sigma_1} - \left(\frac{x-\mu_1}{\sigma_1}\right)^2 = \ln \frac{\pi_2}{\sigma_2} - \left(\frac{x-\mu_2}{\sigma_2}\right)^2$$

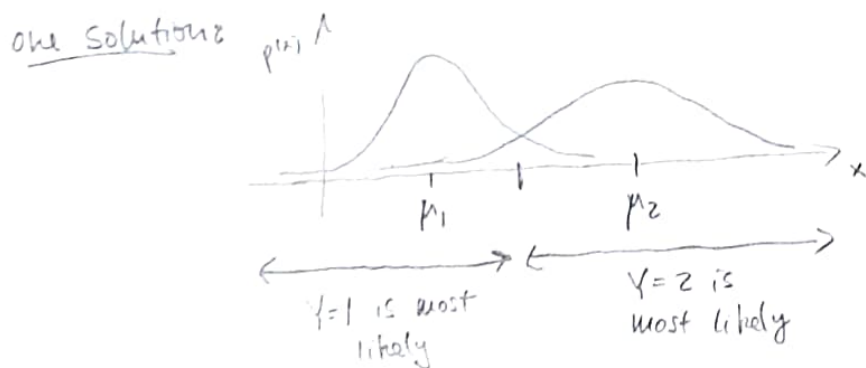
$$\Rightarrow \frac{1}{2}\left(\frac{x-\mu_2}{\sigma_2}\right)^2 - \frac{1}{2}\left(\frac{x-\mu_1}{\sigma_1}\right)^2 - \ln \frac{\pi_2}{\sigma_2} + \ln \frac{\pi_1}{\sigma_1} = 0$$

$$\Rightarrow \frac{1}{2}\left(\frac{1}{\sigma_2^2} - \frac{1}{\sigma_1^2}\right) x^2 + bx + c = 0 \quad \text{can have zero one or two solutions}$$



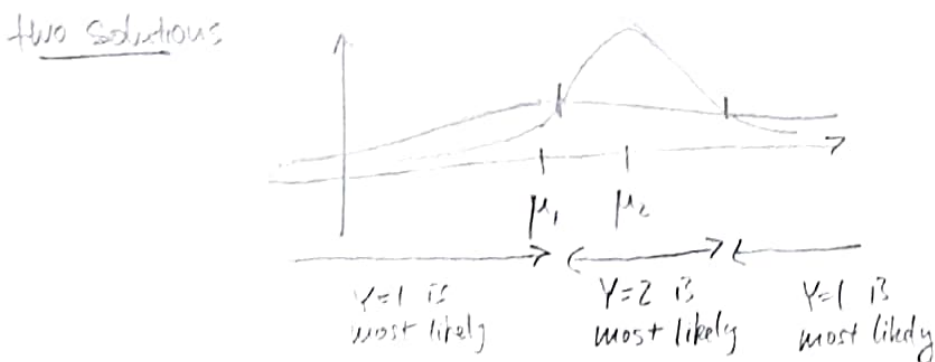
μ_1 & μ_2 are close
 $\pi_2 N(\sigma_2, \mu_2)$ has a small amplitude
 $\pi_1 N(\sigma_1, \mu_1)$ " " larger amplitude

In this case the most likely class is always $Y=1$.



μ_1 and μ_2 are separated

or: $\sigma_1 = \sigma_2$



$\sigma_1 < \sigma_2$

μ_1 & μ_2 are close

Problem 4

HW 4 p3

$$p(Y=1) = \sigma(b + w_1 x_1 + w_2 x_2)$$

$$x_1 = \text{hrs}$$

$$w_1 = 0.05$$

$$x_2 = \text{gpa}$$

$$w_2 = 1$$

$$b = -6$$

$$a.) \quad \sigma(-6 + \underbrace{0.05}_{\frac{1}{20}}(40) + 1 \cdot 3.5) = \sigma(-0.5) = \frac{1}{1+e^{0.5}} \approx 37.6\%$$

$$b.) \quad \text{Note: } \sigma(0) = \frac{1}{2} \Rightarrow \text{increase } x_1 \text{ such that}$$

$$-6 + \frac{1}{20} \cdot x_1 + 3.5 \geq 0$$

$$\Rightarrow x_1 \geq 20 \cdot 2.5 = 50.$$

Problem 5 $Y \in \{0, 1\}$ No / Yes $X = \text{last year's profits}$

$$p(Y=0 | X=x) = \frac{1}{\sqrt{2\pi} \cdot 6} \cdot \exp\left(-\frac{x^2}{2 \cdot 36}\right) \quad \pi_0 = 0.2$$

$$p(Y=1 | X=x) = \frac{1}{\sqrt{2\pi} \cdot 6} \cdot \exp\left(-\frac{(x-10)^2}{2 \cdot 36}\right) \quad \pi_1 = 0.8$$

\uparrow
cancels

Bayes' Law

$$p(X=x | Y=1) = \frac{\exp\left(-\frac{(x-10)^2}{72}\right) \cdot 0.8}{\exp\left(-\frac{x^2}{72}\right) \cdot 0.2 + \exp\left(-\frac{(x-10)^2}{72}\right) \cdot 0.8}$$

$$p(X=4 | Y=1) = \frac{\exp\left(-\frac{36}{72}\right)}{\exp\left(-\frac{16}{72}\right) \cdot 0.2 + \exp\left(-\frac{36}{72}\right) \cdot 0.8} = \frac{\exp\left(-\frac{1}{2}\right)}{\exp\left(-\frac{2}{9}\right) \cdot 0.2 + \exp\left(-\frac{1}{2}\right) \cdot 0.8}$$

$$\approx 94.0\%$$