

## Solution : Homework 2

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**Due Time:** April 9**Problem 1.****Solution:**

$$\begin{aligned}
\arg \min_f \mathbb{E} \ell_{\alpha, \beta}(f(x), y) &= \arg \min_f \mathbb{E}_{X, Y} [\alpha \mathbf{1}\{f(X) = 1, Y = 0\} + \beta \mathbf{1}\{f(X) = 0, Y = 1\}] \\
&= \arg \min_f \mathbb{E}_X [\mathbb{E}_{Y|X} [\alpha \mathbf{1}\{f(X) = 1, Y = 0\} + \beta \mathbf{1}\{f(X) = 0, Y = 1\}]] \\
&= \arg \min_f \mathbb{E}_X [\int_y \alpha \mathbf{1}\{f(X) = 1, y = 0\} + \beta \mathbf{1}\{f(X) = 0, y = 1\} dP(y|x)] \\
&= \arg \min_f \int_x [\alpha \mathbf{1}\{f(x) = 1\} P(y = 0|x) + \beta \mathbf{1}\{f(x) = 0\} P(y = 1|x)] dP(x)
\end{aligned}$$

Thus, we can minimize the integrand by taking:

$$f(x) = \begin{cases} 1 & \beta P(y = 1|x) \geq \alpha P(y = 0|x) \\ 0 & \alpha P(y = 0|x) > \beta P(y = 1|x) \end{cases}$$

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**Problem 2.****Solution:***Answer for problem (1):*

Likelihood term:

$$\begin{aligned}
P(x_1, \dots, x_N | \mu) &= \prod_{i=1}^N P(x_i | \mu) \\
&= \prod_{i=1}^N \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_i - \mu)^2}{2\sigma^2}} \\
\log(P(x_1, \dots, x_N | \mu)) &= \sum_{i=1}^N -\frac{(x_i - \mu)^2}{2\sigma^2} \log\left(\frac{1}{\sqrt{2\pi\sigma^2}}\right) \\
\frac{d \log(P(x_1, \dots, x_N | \mu))}{d\mu} &= \sum_{i=1}^N \frac{x_i - \mu}{\sigma^2}
\end{aligned}$$

If the left part of equation equals to 0:

$$\begin{aligned}\sum_{i=1}^N \frac{x_i - \mu}{\sigma^2} &= 0 \\ \sum_{i=1}^N (x_i - \mu) &= 0 \\ \sum_{i=1}^N \mu &= \sum_{i=1}^N x_i\end{aligned}$$

MLE estimator for the mean  $\mu$ :

$$\hat{\mu} = \frac{\sum_{i=1}^N x_i}{N}$$

Answer for problem (2):

$$P(\mu|x_1, \dots, x_N) = \frac{P(x_1, \dots, x_N|\mu)P(\mu)}{P(x_1, \dots, x_N)}$$

where

$$P(\mu) = \frac{1}{\sqrt{2\pi\beta^2}} e^{-\frac{(x_i - \mu)^2}{2\sigma^2}}$$

The problem is reduce to find the value of  $\mu$  which maximizes:

$$P(\mu|x_1, \dots, x_N) = \frac{(\prod_{i=1}^N \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_i - \mu)^2}{2\sigma^2}}) \frac{1}{\sqrt{2\pi\beta^2}} e^{-\frac{(\mu - v)^2}{2\beta^2}}}{P(x_1, \dots, x_N)}$$

$$\log P(\mu|x_1, \dots, x_N) = \left(\sum_{i=1}^N \frac{(x_i - \mu)^2}{2\sigma^2}\right) \log(\sqrt{2\pi\sigma^2}) + \frac{(\mu - v)^2}{2\beta^2} \log(\sqrt{2\pi\beta^2})$$

$$\frac{\partial \log P(\mu|x_1, \dots, x_N)}{\partial \mu} = \left(\sum_{i=1}^N \frac{x_i - \mu}{\sigma^2}\right) - \frac{\mu - v}{\beta^2}$$

If the left part of equation equals to 0:

$$\begin{aligned}\left(\sum_{i=1}^N \frac{x_i - \mu}{\sigma^2}\right) - \frac{\mu - v}{\beta^2} &= 0 \\ \left(\sum_{i=1}^N \frac{x_i - \mu}{\sigma^2}\right) &= \frac{\mu - v}{\beta^2} \\ \frac{\mu}{\beta^2} + \frac{N\mu}{\sigma^2} &= \frac{\sum_{i=1}^N x_i}{\sigma^2} + \frac{v}{\beta^2} \\ \frac{(\sigma^2 + N\beta^2)\mu}{\sigma^2\beta^2} &= \frac{\sigma^2 v + \beta^2 \sum_{i=1}^N x_i}{\sigma^2\beta^2}\end{aligned}$$

MLE estimator for the mean  $\mu$ :

$$\hat{\mu} = \frac{\sigma^2 v + \beta^2 \sum_{i=1}^N x_i}{\sigma^2 + N\beta^2}$$

Answer for problem (3):

$$\begin{aligned}
 N &\rightarrow \infty \\
 \frac{\sigma^2}{N\beta^2} &\rightarrow 0 \\
 \frac{\sigma v}{N\beta^2 + \sigma^2} &\rightarrow 0 \\
 \hat{\mu}_{MAP} &\rightarrow \frac{\sum_{i=1}^N x_i}{N} = \hat{\mu}_{MLE}
 \end{aligned}$$

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### Problem 3.

**Solution:**

Answer for problem (1):

$$\begin{aligned}
 P(\text{Class} = X) &= \frac{2}{3}, P(\text{Class} = Y) = \frac{1}{3} \\
 P(A1 = 0|\text{Class} = X) &= \frac{1}{2}, P(A1 = 1|\text{Class} = X) = \frac{1}{4}, P(A1 = 2|\text{Class} = X) = \frac{1}{4} \\
 P(A2 = 0|\text{Class} = X) &= 0, P(A2 = 1|\text{Class} = X) = \frac{3}{4}, P(A2 = 2|\text{Class} = X) = \frac{1}{4} \\
 P(A1 = 0|\text{Class} = Y) &= 0, P(A1 = 1|\text{Class} = Y) = \frac{1}{2}, P(A1 = 2|\text{Class} = Y) = \frac{1}{2} \\
 P(A2 = 0|\text{Class} = Y) &= \frac{1}{2}, P(A2 = 1|\text{Class} = Y) = 0, P(A2 = 2|\text{Class} = Y) = \frac{1}{2} \\
 P(\text{Class} = X|A1 = 2, A2 = 2) &= P(\text{Class} = X)P(A1 = 2|\text{Class} = X)P(A2 = 2|\text{Class} = X) = \frac{1}{24} \\
 P(\text{Class} = Y|A1 = 2, A2 = 2) &= P(\text{Class} = Y)P(A1 = 2|\text{Class} = Y)P(A2 = 2|\text{Class} = Y) = \frac{1}{12}
 \end{aligned}$$

So Class is predicted to Y.

Answer for problem (2):

$$\begin{aligned}
 P(\text{Class} = X) &= \frac{5}{8}, P(\text{Class} = Y) = \frac{3}{8} \\
 P(A1 = 0|\text{Class} = X) &= \frac{3}{7}, P(A1 = 1|\text{Class} = X) = \frac{2}{7}, P(A1 = 2|\text{Class} = X) = \frac{2}{7} \\
 P(A2 = 0|\text{Class} = X) &= \frac{1}{7}, P(A2 = 1|\text{Class} = X) = \frac{4}{7}, P(A2 = 2|\text{Class} = X) = \frac{2}{7} \\
 P(A1 = 0|\text{Class} = Y) &= \frac{1}{5}, P(A1 = 1|\text{Class} = Y) = \frac{2}{5}, P(A1 = 2|\text{Class} = Y) = \frac{2}{5} \\
 P(A2 = 0|\text{Class} = Y) &= \frac{2}{5}, P(A2 = 1|\text{Class} = Y) = \frac{1}{5}, P(A2 = 2|\text{Class} = Y) = \frac{2}{5} \\
 P(\text{Class} = X|A1 = 2, A2 = 2) &= P(\text{Class} = X)P(A1 = 2|\text{Class} = X)P(A2 = 2|\text{Class} = X) = \frac{5}{98} \\
 P(\text{Class} = Y|A1 = 2, A2 = 2) &= P(\text{Class} = Y)P(A1 = 2|\text{Class} = Y)P(A2 = 2|\text{Class} = Y) = \frac{3}{50}
 \end{aligned}$$

Result of previous question do not change.

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### Problem 4.

**Solution:**

The classification accuracy is 83.33%

Using col 7 - col 54 the accuracy raise to 85.00%

