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State Finished

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Time taken 5 days 2 hours

Grade 9.00 out of 10.00 (90%)

Question 1

Correct

Mark 1.00 out of 1.00

The PDF of the Gaussian mixture is given as

- ☒ $\sum_{i=1}^K p_i \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2}\|\bar{\mathbf{x}}-\bar{\boldsymbol{\mu}}_i\|^2}$
- ☐ $\sum_{i=1}^K \left(\left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2}\|\bar{\mathbf{x}}-\bar{\boldsymbol{\mu}}_i\|^2}\right)^{p_i}$
- ☐ $\prod_{i=1}^K p_i \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2}\|\bar{\mathbf{x}}-\bar{\boldsymbol{\mu}}_i\|^2}$
- ☐ $\prod_{i=1}^K \left(\left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2}\|\bar{\mathbf{x}}-\bar{\boldsymbol{\mu}}_i\|^2}\right)^{p_i}$



Your answer is correct.

The correct answer is:

$$\sum_{i=1}^K p_i \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2}\|\bar{\mathbf{x}}-\bar{\boldsymbol{\mu}}_i\|^2}$$

Question 2

Correct

Mark 1.00 out of 1.00

The likelihood of the complete data is

- ☐ $\prod_{j=1}^M \prod_{i=1}^K \left(\alpha_i(j) p_i \times \left(\frac{1}{2\pi\sigma^2} \right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2} \right)$
- ☒ $\prod_{j=1}^M \prod_{i=1}^K \left(p_i \times \left(\frac{1}{2\pi\sigma^2} \right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2} \right)^{\alpha_i(j)}$ ✓
- ☐ $\prod_{j=1}^M \sum_{i=1}^K p_i \times \left(\frac{1}{2\pi\sigma^2} \right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}} - \bar{\boldsymbol{\mu}}_i\|^2}$
- ☐ $\sum_{j=1}^M \sum_{i=1}^K \alpha_i(j) p_i \times \left(\frac{1}{2\pi\sigma^2} \right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}} - \bar{\boldsymbol{\mu}}_i\|^2}$

Your answer is correct.

The correct answer is:

$$\prod_{j=1}^M \prod_{i=1}^K \left(p_i \times \left(\frac{1}{2\pi\sigma^2} \right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2} \right)^{\alpha_i(j)}$$

Question 3

Correct

Mark 1.00 out of 1.00

The expected value of the log-likelihood in iteration l is

- ☐ $\prod_{j=1}^M \sum_{i=1}^N \left(\alpha_i^{(l)}(j) \ln p_i - \frac{N}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2 \right)$
- ☒ $\sum_{j=1}^M \sum_{i=1}^N \alpha_i^{(l)}(j) \left(\ln p_i - \frac{N}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2 \right)$ ✓
- ☐ $\prod_{j=1}^M \prod_{i=1}^N \left(\alpha_i^{(l)}(j) \ln p_i - \frac{N}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2 \right)$
- ☐ $\sum_{j=1}^M \prod_{i=1}^N \left(\ln p_i - \frac{N}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2 \right)^{\alpha_i^{(l)}(j)}$

Your answer is correct.

The correct answer is:

$$\sum_{j=1}^M \sum_{i=1}^N \alpha_i^{(l)}(j) \left(\ln p_i - \frac{N}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i\|^2 \right)$$

Question **4**

Correct

Mark 1.00 out of 1.00

The quantity $\alpha_i^{(l)}(j) = \Pr(\mathcal{C}_i | \bar{\mathbf{x}}(j))$ is given as

- ☒
$$\frac{p_i \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i^{(l-1)}\|^2}}{\sum_{k=1}^K p_k \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_k^{(l-1)}\|^2}}$$
- ☐
$$\frac{p_i \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i^{(l-1)}\|^2}}{\prod_{k=1}^K p_k \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_k^{(l-1)}\|^2}}$$
- ☐
$$\frac{\left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i^{(l-1)}\|^2}}{\sum_{k=1}^K \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_k^{(l-1)}\|^2}}$$
- ☐
$$\frac{p_i}{\sum_{k=1}^K p_k}$$



Your answer is correct.

The correct answer is:

$$\frac{p_i \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_i^{(l-1)}\|^2}}{\sum_{k=1}^K p_k \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{\mathbf{x}}(j) - \bar{\boldsymbol{\mu}}_k^{(l-1)}\|^2}}$$

Question 5

Correct

Mark 1.00 out of 1.00

The quantity $\bar{\mu}_i^{(l)}$ is given as

- ☐ $\frac{\sum_{j=1}^M \alpha_i^{(l)}(j) e^{-\frac{1}{2\sigma^2} \|\bar{x}(j) - \bar{\mu}_i^{(l-1)}\|^2}}{\sum_{j=1}^M \alpha_i^{(l)}(j)}$
- ☐ $\frac{\sum_{i=1}^K \alpha_i^{(l)}(j) \bar{x}(j)}{\sum_{i=1}^K \alpha_i^{(l)}(j)}$
- ☐ $\frac{\sum_{j=1}^M \alpha_i^{(l)}(j) p_i \times \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{N}{2}} e^{-\frac{1}{2\sigma^2} \|\bar{x}(j) - \bar{\mu}_i^{(l-1)}\|^2}}{\sum_{j=1}^M p_i \alpha_i^{(l)}(j)}$
- ☒ $\frac{\sum_{j=1}^M \alpha_i^{(l)}(j) \bar{x}(j)}{\sum_{j=1}^M \alpha_i^{(l)}(j)}$



Your answer is correct.

The correct answer is:

$$\frac{\sum_{j=1}^M \alpha_i^{(l)}(j) \bar{x}(j)}{\sum_{j=1}^M \alpha_i^{(l)}(j)}$$

Question 6

Correct

Mark 1.00 out of 1.00

The **entropy** $H(X)$ of this source is

- ☒ $\sum_{i=1}^n p(x_i) \log_2 \frac{1}{p(x_i)}$
- ☐ $\sum_{i=1}^n p(x_i) \log_2 p(x_i)$
- ☐ $\sum_{i=1}^n \frac{1}{p(x_i)} \log_2 \frac{1}{p(x_i)}$
- ☐ $\sum_{i=1}^n \log_2 \frac{1}{p(x_i)}$



Your answer is correct.

The correct answer is:

$$\sum_{i=1}^n p(x_i) \log_2 \frac{1}{p(x_i)}$$

Question 7

Incorrect

Mark 0.00 out of 1.00

Consider a source with 8 equiprobable symbols. What is its entropy?

- ☒ 1
- ☐ 1.5
- ☐ 3
- ☐ 2



Your answer is incorrect.

The correct answer is:

3

Question 8

Correct

Mark 1.00 out of 1.00

The **conditional entropy** $H(X|Y)$ is defined as

- ☐ $\sum_{j=1}^m p(y_j) H(Y = y_j | X)$
- ☐ $\sum_{j=1}^m H(X | Y = y_j)$
- ☐ $\sum_{i=1}^n p(x_i) H(Y | X = x_i)$
- ☒ $\sum_{j=1}^m p(y_j) H(X | Y = y_j)$



Your answer is correct.

The correct answer is:

$$\sum_{j=1}^m p(y_j) H(X | Y = y_j)$$

Question 9

Correct

Mark 1.00 out of 1.00

To construct the decision tree classifier (DTC), one has to choose the **feature** that

- ☒ maximizes the **information gain**
- ☐ minimizes the **information gain**
- ☐ has zero **information gain**
- ☐ that has **information gain** equal to unity



Your answer is correct.

The correct answer is:

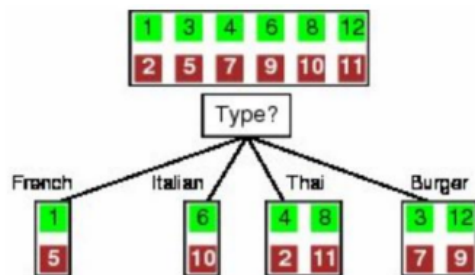
maximizes the **information gain**

Question 10

Correct

Mark 1.00 out of 1.00

What is the information gain for the type feature depicted in the figure below?



- ☐ 0.82
- ☐ 0.36
- ☒ 0
- ☐ 0.54



Your answer is correct.

The correct answer is:

0