

作业十

1. 第一题

1.

$$H(X) = -\frac{2}{3} \log \frac{2}{3} - \frac{1}{3} \log \frac{1}{3} = -\frac{2}{3} + \log 3$$
$$H(Y) = -\frac{2}{3} \log \frac{2}{3} - \frac{1}{3} \log \frac{1}{3} = -\frac{2}{3} + \log 3$$

2.

$$H(X | Y) = \frac{1}{3}H(X | Y = 0) + \frac{2}{3}H(X | Y = 1) = \frac{1}{3}(0) + \frac{2}{3} \left(-2 \cdot \frac{1}{2} \log \frac{1}{2} \right) = \frac{2}{3}$$
$$H(Y | X) = \frac{2}{3}H(Y | X = 0) + \frac{1}{3}H(Y | X = 1) = \frac{2}{3} \left(-2 \cdot \frac{1}{2} \log \frac{1}{2} \right) + \frac{1}{3}(0) = \frac{2}{3}$$

3.

$$H(X, Y) = H(X | Y) + H(Y) = \log 3$$

4.

$$I(X; Y) = p(0, 0) \log \frac{p(0, 0)}{p(x=0)p(y=0)} + p(1, 0) \log \frac{p(1, 0)}{p(x=1)p(y=0)} + p(0, 1) \log \frac{p(0, 1)}{p(x=0)p(y=1)} + p(1, 1) \log \frac{p(1, 1)}{p(x=1)p(y=1)} = -\frac{4}{3} + \log 3$$

2. 第二题

1.

$$\alpha \sim \mathcal{N} \left(\begin{bmatrix} 10 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right), \quad \beta \sim \mathcal{N} \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 8.4 & 2.0 \\ 2.0 & 1.7 \end{bmatrix} \right)$$

考虑这两个高斯分布的混合 $\alpha + \beta$:

$$p(x_1) = \int_{-\infty}^{\infty} 0.4 \cdot \mathcal{N} \left(\begin{bmatrix} 10 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) dx_2 + \int_{-\infty}^{\infty} 0.6 \cdot \mathcal{N} \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 8.4 & 2.0 \\ 2.0 & 1.7 \end{bmatrix} \right) dx_2$$
$$= 0.4 \mathcal{N}(10, 1) + 0.6 \mathcal{N}(0, 8.4)$$

$$p(x_2) = \int_{-\infty}^{\infty} 0.4 \cdot \mathcal{N} \left(\begin{bmatrix} 10 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) dx_1 + \int_{-\infty}^{\infty} 0.6 \cdot \mathcal{N} \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 8.4 & 2.0 \\ 2.0 & 1.7 \end{bmatrix} \right) dx_1$$
$$= 0.4 \mathcal{N}(2, 1) + 0.6 \mathcal{N}(0, 1.7)$$

2.

$$\mathbb{E} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{bmatrix} 10 \times 0.4 \\ 2 \times 0.4 \end{bmatrix} = \begin{bmatrix} 4 \\ 0.8 \end{bmatrix}$$

3. 第三题

1.

$$p(y | x) = \frac{p(x, y)}{p(x)} = \frac{p(x, w)}{p(x)},$$

又因为 x, w 是独立的, 所以

$$y | x \sim \mathcal{N}(Ax + b, Q).$$

$$\begin{aligned} p(y | x) &= p(w) = p(y - Ax - b) \\ &= \frac{1}{(2\pi)^{\frac{D}{2}} |Q|^{\frac{1}{2}}} \exp \left\{ -\frac{1}{2} (y - Ax - b)^T Q^{-1} (y - Ax - b) \right\}. \end{aligned}$$

2. 由 $x \sim \mathcal{N}(x | \mu_x, \Sigma_x)$, 可得

$$p(x) = \frac{1}{(2\pi)^{\frac{D}{2}} |\Sigma_x|^{\frac{1}{2}}} \exp \left\{ -\frac{1}{2} (x - \mu_x)^T \Sigma_x^{-1} (x - \mu_x) \right\}.$$

首先求均值 μ_y :

$$\begin{aligned} \mu_y &= \mathbb{E}[y] = \mathbb{E}[Ax + b + w] \\ &= A\mu_x + b. \end{aligned}$$

然后, 求协方差 Σ_y :

$$\begin{aligned} \Sigma_y &= \text{Cov}(y) = \mathbb{E}[(y - \mu_y)(y - \mu_y)^T] \\ &= \mathbb{E}[(AX - A\mu_x + w)(AX - A\mu_x + w)^T] \\ &= A \mathbb{E}[(x - \mu_x)(x - \mu_x)^T] A^T + \mathbb{E}[ww^T] \\ &= A\Sigma_x A^T + Q. \end{aligned}$$

因此

$$y \sim \mathcal{N}(A\mu_x + b, A\Sigma_x A^T + Q).$$

4. 第四题

原问题可以描述为:

$$\begin{aligned} \max \quad & p(x) = - \int_{-\infty}^{+\infty} p(x) \ln p(x) dx \\ \text{subject to} \quad & \int_{-\infty}^{+\infty} p(x) dx = 1, \\ & \int_{-\infty}^{+\infty} xp(x) dx = \mu, \\ & \int_{-\infty}^{+\infty} (x - \mu)^2 p(x) dx = \sigma^2. \end{aligned}$$

构造拉格朗日函数：

$$L(p, \lambda) = - \int_{-\infty}^{+\infty} p(x) \ln p(x) dx + \lambda_1 \left(\int_{-\infty}^{+\infty} p(x) dx - 1 \right) \\ + \lambda_2 \left(\int_{-\infty}^{+\infty} xp(x) dx - \mu \right) + \lambda_3 \left(\int_{-\infty}^{+\infty} (x - \mu)^2 p(x) dx - \sigma^2 \right).$$

其中 $\lambda = (\lambda_1, \lambda_2, \lambda_3)$ ，继而整理得：

$$L(p, \lambda_1, \lambda_2, \lambda_3) = \int_{-\infty}^{+\infty} [-\ln p(x) + \lambda_1 + \lambda_2 x + \lambda_3 (x - \mu)^2] p(x) dx \\ - \lambda_1 - \lambda_2 \mu - \lambda_3 \sigma^2.$$

令

$$\frac{\partial}{\partial p(x)} [-\ln p(x) + \lambda_1 + \lambda_2 x + \lambda_3 (x - \mu)^2] = -\ln p(x) - 1 + \lambda_1 + \lambda_2 x + \lambda_3 (x - \mu)^2 = 0,$$

则有

$$p(x) = e^{\lambda_1 - 1 + \lambda_2 x + \lambda_3 (x - \mu)^2} \\ = C e^{\lambda_3 \left(x - \mu + \frac{\lambda_2}{2\lambda_3}\right)^2}.$$

因为 $p(x) > 0$ ，所以 $C > 0$ 。且因为 $p(x)$ 关于 $x = \mu - \frac{\lambda_2}{2\lambda_3}$ 对称，所以 $\mathbb{E}[p(x)] = \mu - \frac{\lambda_2}{2\lambda_3}$ 。因此 $\lambda_2 = 0$ 。由此，

$$p(x) = C e^{\lambda_3 (x - \mu)^2},$$

代入约束条件，可得

$$C = \frac{1}{\sqrt{2\pi\sigma^2}}, \quad \lambda_3 = -\frac{1}{2\sigma^2}.$$

因此，

$$p(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{(x - \mu)^2}{2\sigma^2} \right\}.$$

因此可知在均值和方差约束下，最大熵分布为高斯分布。

5. 第五题

事件 $A = \{\text{拿到芒果}\}$ ， $B = \{\text{从第二个袋子拿到芒果}\}$ 。

$$P(B | A) = \frac{0.4 \times 0.5}{0.6 \times \frac{2}{3} + 0.4 \times 0.5} = \frac{1}{3}.$$