## Advanced probabilistic methods - Sketch

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$$\mathcal{L}(q) = \mathbb{E}_q[\log p(\mathbf{X}, \mathbf{Z})] - \mathbb{E}_q[\log q(\mathbf{Z})] \tag{1}$$

$$\mathbb{E}_{q} \log p(\mathbf{X}, \mathbf{Z}) = \mathbb{E}_{q} \left[ \log p(\tau) + \log p\theta + \log p(\mathbf{z} \mid \tau) + \log p(\mathbf{x} \mid \mathbf{z}\theta) \right]$$
(2)

$$= \mathbb{E}_{q(\tau)} \left[ \log p(\tau) \right] + \mathbb{E}_{q(\theta)} \left[ \log p\theta \right] + \mathbb{E}_{q(\mathbf{z})q(\tau)} \left[ \log p(\mathbf{z} \mid \tau) \right] + \mathbb{E}_{q(\mathbf{z})q(\theta)} \left[ \log p(\mathbf{x} \mid \mathbf{z}, \theta) \right]$$
(3)

$$\mathbb{E}_q[\log q(\mathbf{Z})] = \mathbb{E}_q\left[-\log q(\mathbf{z}) - \log q(\tau) - \log q(\theta)\right] \tag{4}$$

$$= -\mathbb{E}_{q(\mathbf{z})}\left[\log q(\mathbf{z})\right] - \mathbb{E}_{q(\tau)}\left[\log q(\tau)\right] - \mathbb{E}_{q(\theta)}\left[\log q(\theta)\right] \tag{5}$$

$$\mathcal{L}(q) = \mathbb{E}_{q(\tau)} \left[ \log p(\tau) \right] + \mathbb{E}_{q(\theta)} \left[ \log p(\theta) \right] + \mathbb{E}_{q(\mathbf{z})q(\tau)} \left[ \log p(\mathbf{z} \mid \tau) \right] + \mathbb{E}_{q(\mathbf{z})q(\theta)} \left[ \log p(\mathbf{x} \mid \mathbf{z}, \theta) \right]$$
(6)

$$-\mathbb{E}_{q(\mathbf{z})}\left[\log q(\mathbf{z})\right] - \mathbb{E}_{q(\tau)}\left[\log q(\tau)\right] - \mathbb{E}_{q(\theta)}\left[\log q(\theta)\right] \tag{7}$$

$$p(\theta) = \mathcal{N}(\theta \mid 0, \beta_0^{-1}) \tag{8}$$

$$\mathbb{E}_{q(\theta)} \left[ \log p(\theta) \right] = \mathbb{E}_{q(\theta)} \left[ \log \mathcal{N}(\theta \mid 0, \beta_0^{-1}) \right] \tag{9}$$

$$\propto \mathbb{E}_{q(\theta)} \left[ -\frac{1}{2} \theta^2 \beta_0 \right] \tag{10}$$

$$= -\frac{1}{2} \mathbb{E}_{q(\theta)} \left[ \theta^2 \right] \beta_0 \tag{11}$$

$$\mathbb{E}_{q(\theta)} \left[ \theta^2 \right] = \operatorname{var}(\theta) + \mathbb{E}_{q(\theta)} \left[ \theta \right]^2$$
(12)

$$=\beta_2^{-1} + m_2^2 \tag{13}$$

$$\mathbb{E}_{q(\theta)}\left[\log p(\theta)\right] \propto -\frac{\beta_0}{2}(\beta_2^{-1} + m_2^2) \tag{14}$$

$$q(\theta) = \mathcal{N}(\theta \mid m_2, \beta_2^{-1}) \tag{15}$$

$$\mathbb{E}_{q(\theta)} \left[ \log q(\theta) \right] = \mathbb{E}_{q(\theta)} \left[ \log \mathcal{N}(\theta \mid m_2, \beta_2^{-1}) \right]$$
(16)

$$\propto \mathbb{E}_{q(\theta)} \left[ -\frac{1}{2} (\theta - m_2)^2 \beta_2 \right] \tag{17}$$

$$= -\frac{1}{2} \mathbb{E}_{q(\theta)} \left[ (\theta - m_2)^2 \right] \beta_2 \tag{18}$$

$$= -\frac{1}{2} \mathbb{E}_{q(\theta)} \left[ (\theta^2 - 2\theta m_2 + m_2^2) \beta_2 \right]$$
 (19)

$$\propto -\frac{1}{2} \mathbb{E}_{q(\theta)} \left[ \theta^2 - 2\theta m_2 \right] \beta_2 \tag{20}$$

$$= -\frac{\beta_2}{2} (\mathbb{E}_{q(\theta)} \left[ \theta^2 \right] - \mathbb{E}_{q(\theta)} \left[ 2\theta m_2 \right]) \tag{21}$$

$$= -\frac{\beta_2}{2}(\beta_2^{-1} + m_2^2 - 2m_2^2) \tag{22}$$

$$= -\frac{\beta_2}{2}(\beta_2^{-1} - m_2^2) \tag{23}$$

$$=\frac{m_2^2\beta_2}{2} - \frac{1}{2} \tag{24}$$

(25)

$$p(x \mid z, \theta) = \prod_{n=1}^{N} \mathcal{N}(x_n \mid 0, 1)^{z_{n1}} \mathcal{N}(x_n \mid \theta, 1)^{z_{n2}}$$
(26)

$$E_{q(z)q(\theta)} \left[ \log p(x \mid z, \theta) \right] = E_{q(z)q(\theta)} \left[ \sum_{n=1}^{N} z_{n1} \log \mathcal{N}(x_n \mid 0, 1) + \sum_{n=1}^{N} z_{n2} \log \mathcal{N}(x_n \mid \theta, 1) \right]$$
(27)

$$\propto E_{q(z)q(\theta)} \left[ \sum_{n=1}^{N} z_{n1} \left( -\frac{1}{2} x_n^2 \right) + \sum_{n=1}^{N} z_{n2} \left( -\frac{1}{2} \right) (x_n - \theta)^2 \right]$$
 (28)

$$=E_{q(z)q(\theta)}\left[\sum_{n=1}^{N} z_{n1} \left(-\frac{1}{2}x_{n}^{2}\right) + \sum_{n=1}^{N} z_{n2} \left(-\frac{1}{2}\right) \left(x_{n}^{2} - 2x_{n}\theta + \theta^{2}\right)\right]$$
(29)

$$= \sum_{n=1}^{N} E_{q(z)} [z_{n1}] \left( -\frac{1}{2} x_{n}^{2} \right) + \sum_{n=1}^{N} E_{q(z)} [z_{n2}] \left( -\frac{1}{2} \right) (x_{n}^{2} - 2x_{n} E_{q(\theta)} [\theta] + E_{q(\theta)} [\theta^{2}])$$
(30)

$$E_{q(z)}\left[z_{nk}\right] = r_{nk} \tag{31}$$

$$E_{q(\theta)}[\theta] = m_2 \tag{32}$$

$$E_{q(\theta)} \left[ \theta^2 \right] = \beta_2^{-1} + m_2^2$$
 (33)

$$E_{q(z)q(\theta)}\left[\log p(x\mid z,\theta)\right] \propto \sum_{n=1}^{N} r_{n1} \left(-\frac{1}{2}x_n^2\right) + \sum_{n=1}^{N} r_{n2} \left(-\frac{1}{2}\right) \left(x_n^2 - 2x_n m_2 + \beta_2^{-1} + m_2^2\right)$$
(34)

$$= \sum_{n=1}^{N} r_{n1} \left( -\frac{1}{2} x_n^2 \right) + \sum_{n=1}^{N} r_{n2} \left( -\frac{1}{2} \right) \left( (x_n - m_2)^2 + \beta_2^{-1} \right)$$
 (35)

We denote  $x_i$  as the number of black pebbles in the *i*th bag and n the total amount of pebbels.

$$x_i \sim \text{Bin}(x_i \mid n, a) = \binom{n}{x_i} a^{x_i} (1 - a)^{x_i}$$
 (36)

$$a \sim \text{Beta}(a \mid 1, 1) = 1 \tag{37}$$

$$p(x \mid M_1) = \int_0^1 \text{Bin}(x_1 \mid n, a) \text{Bin}(x_2 \mid n, a) \text{Beta}(a \mid 1, 1)$$
(38)

$$= \int_0^1 \binom{n}{x_1} a^{x_1} (1-a)^{n-x_1} \binom{n}{x_2} a^{x_2} (1-a)^{n-x_2} da$$
 (39)

$$= \binom{n}{x_1} \binom{n}{x_2} \int_0^1 a^{x_1 + x_2} (1 - a)^{2n - x_1 - x_2} da$$
 (40)

$$= \binom{n}{x_1} \binom{n}{x_2} \int_0^1 a^{x_1 + x_2 - 1 + 1} (1 - a)^{x_1 + x_2 - 1 + 1} da \tag{41}$$

$$= \binom{n}{x_1} \binom{n}{x_2} B(x_1 + x_2 + 1, 2n - x_1 - x_2 + 1)$$
(42)

For the second model we have

$$x_i \sim \operatorname{Bin}(n, a_i) = \binom{n}{x_i} a_i^{x_i} (1 - a_i)^{n - x_i}$$

$$\tag{43}$$

$$b \sim \text{Beta}(1,1) = 1 \tag{44}$$

(45)

We again note i = b for black marbles and i = w for the white ones.

$$p(x \mid M_2) = \int_0^1 \int_0^1 \text{Bin}(x_1 \mid n, a_1) \text{Bin}(x_2 \mid n, a_2) \text{Beta}(a_1 \mid 1, 1) \text{Beta}(a_2 \mid 1, 1) da_1 da_2$$
 (46)

$$= \int_0^1 \int_0^1 \binom{n}{x_1} a_1^{x_1} (1 - a_1)^{n - x_1} \binom{n}{x_2} a_2^{x_2} (1 - a_2)^{n - x_2} da_1 da_2 \tag{47}$$

$$= \binom{n}{x_1} \binom{n}{x_2} \int_0^1 a_1^{x_1} (1 - a_1)^{n - x_1} da_1 \int_0^1 a_2^{x_2} (1 - a_2)^{n - x_2} da_2$$
(48)

$$= \binom{n}{x_1} \binom{n}{x_2} \int_0^1 a_1^{x_1 - 1 + 1} (1 - a_1)^{n - x_1 - 1 + 1} da_1 \int_0^1 a_2^{x_2 - 1 + 1} (1 - a_2)^{n - x_2 - 1 + 1} da_2$$
 (49)

$$= \binom{n}{x_1} \binom{n}{x_2} B(x_1+1, n-x_1+1) B(x_2+1, n-x_2+1)$$
(50)

(51)

We can now formulate the posterior odds as

$$\frac{p(x \mid M_1)}{p(x \mid M_2)} = \frac{\binom{n}{x_1} \binom{n}{x_2} B(x_1 + x_2 + 1, 2n - x_1 - x_2 + 1)}{\binom{n}{x_1} \binom{n}{x_2} B(x_1 + 1, n - x_1 + 1) B(x_2 + 1, n - x_2 + 1)}$$
(52)

$$= \frac{B(x_1 + x_2 + 1, 2n - x_1 - x_2 + 1)}{B(x_1 + 1, n - x_1 + 1)B(x_2 + 1, n - x_2 + 1)}$$
(53)