

Q 5 (a)

$$l(\theta) = \log \prod_{i=1}^m P(x^{(i)} | y^{(i)}; \theta) P(y^{(i)})$$

$$= \log \prod_{i=1}^m \left[\prod_{j=1}^n P(x_j^{(i)} | y^{(i)}; \theta) \right] P(y^{(i)})$$

$$= \left[\sum_{i=1}^m \sum_{j=1}^n \log P(x_j^{(i)} | y^{(i)}; \theta) \right] + \sum_{i=1}^m \log P(y^{(i)})$$

$$= \sum_{i=1}^m \sum_{j=1}^n \log \left(\binom{P(i)}{x_j(i)} \alpha_j | y=y(i) \right)^{P(i)-x_j(i)} + \sum_{i=1}^m \log (P_y(i))$$

\downarrow
 Prior probability

$$L = \sum_{i=1}^m \sum_{j=1}^n \log \left(\binom{P(i)}{x_j(i)} + \log \alpha_j | y=y(i) + \log (1 - \alpha_j | y) \right)^{P(i)-x_j(i)}$$

To take the derivative of the log likelihood we can ignore $\sum_{j=1}^n$ since we are taking derivative with respect to $\alpha_j | y=y(i)$

$$\frac{\partial L}{\partial \alpha_j | y=j} = \sum_{i=1}^m 0 + \frac{x_j(i)}{\alpha_j | y=y(i)} + \frac{P(i) - x_j(i)}{1 - \alpha_j | y=y(i)}$$

Equating this derivative to 0 we get

$$\sum_{i=1}^m x_j(i) (1 - \alpha_j | y=y(i)) + P(i) - x_j(i) \alpha_j | y=y(i) = 0$$

$$\alpha = \frac{\sum_{i=1}^m x_i(i)}{\sum_{i=1}^m x_i(i) + P(i) - x_i(i)}$$

considering $a = x_i(i)$
 $b = P(i) - x_i(i)$

Hence $\sum a(1-x) + bx$
 $x = \frac{\sum a}{\sum a + b}$

$$= \frac{\sum_{i=1}^m x_i(i)}{\sum_{i=1}^m P(i)}$$

Hence $\alpha = \frac{\sum_{i=1}^m x_i(i)}{\sum_{i=1}^m P(i)}$