

1 EM

1.1 E Step

For E step and just for now, all the probabilities $P(\text{hidden} \mid \text{observed})$ are computed using marginalization.

1.2 M Step

For E step we want to find

$$\operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \log(p(h, y_k | x_k))$$

which can be formulated as

$$\begin{aligned} & \operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \log(b(h, y_k, x_k)) - \sum_{k=1}^N \sum_h q(h) \log(b(x_k)) \\ &= \operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \log(b(h, y_k, x_k)) - \sum_{k=1}^N \log(b(x_k)) \sum_h q(h) \\ &= \operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \log(b(h, y_k, x_k)) - \sum_{k=1}^N \log(b(x_k)) \\ &= \operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \log(b(h, y_k, x_k)) - \sum_{k=1}^N \log(\sum_{h', y'} b(h', y', x_k)) \end{aligned}$$

Simplifying it "Assuming nothing will change"

$$\begin{aligned} &= \operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \log(b(h, y_k, x_k)) - \sum_{k=1}^N \sum_{h', y'} \log(b(h', y', x_k)) \\ &= \operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \log(e^{\theta_{(h, y_k, x_k)}}) - \sum_{k=1}^N \sum_{h', y'} \log(e^{\theta_{(h', y', x_k)}}) \end{aligned}$$

Where $\theta_{(h, y, x)} = \sum_{\theta_{i, j}} \theta_{i, j}$ *such that* $i, j \in h \cup y \cup x$

$$= \operatorname{argmax}_{\theta} \sum_{k=1}^N \sum_h q(h) \theta_{(h, y_k, x_k)} - \sum_{k=1}^N \sum_{h', y'} \theta_{(h', y', x_k)}$$