

Now, using ② -

$$q(z) \propto P(x, z | \theta)$$

$$q(z) = \frac{P(x, z | \theta)}{\sum_z P(x, z | \theta)} \left\{ \text{Normalizing the sum to 1.} \right\}$$

$$= \frac{P(x, z | \theta)}{P(x | \theta)}$$

$$= P(z | x, \theta) \left\{ \text{By definition of conditional probability} \right\}.$$

(i) Now for the E Step:-

$$\text{we will set } q(z) := P(z | x, \theta) \text{ ———— ④}$$

~~This~~ This E-step will create a lower bound on the log likelihood function which is tight at the current value of θ .

(ii) for the M-step:-

We will maximize the lower bound with respect to our parameters θ .

$$\theta := \underset{\theta}{\operatorname{argmax}} \sum_z q(z) \log \frac{P(x, z | \theta)}{q(z)} \text{ ———— ⑤}$$

Hence ④ & ⑤ gives the E & M step of the equation.