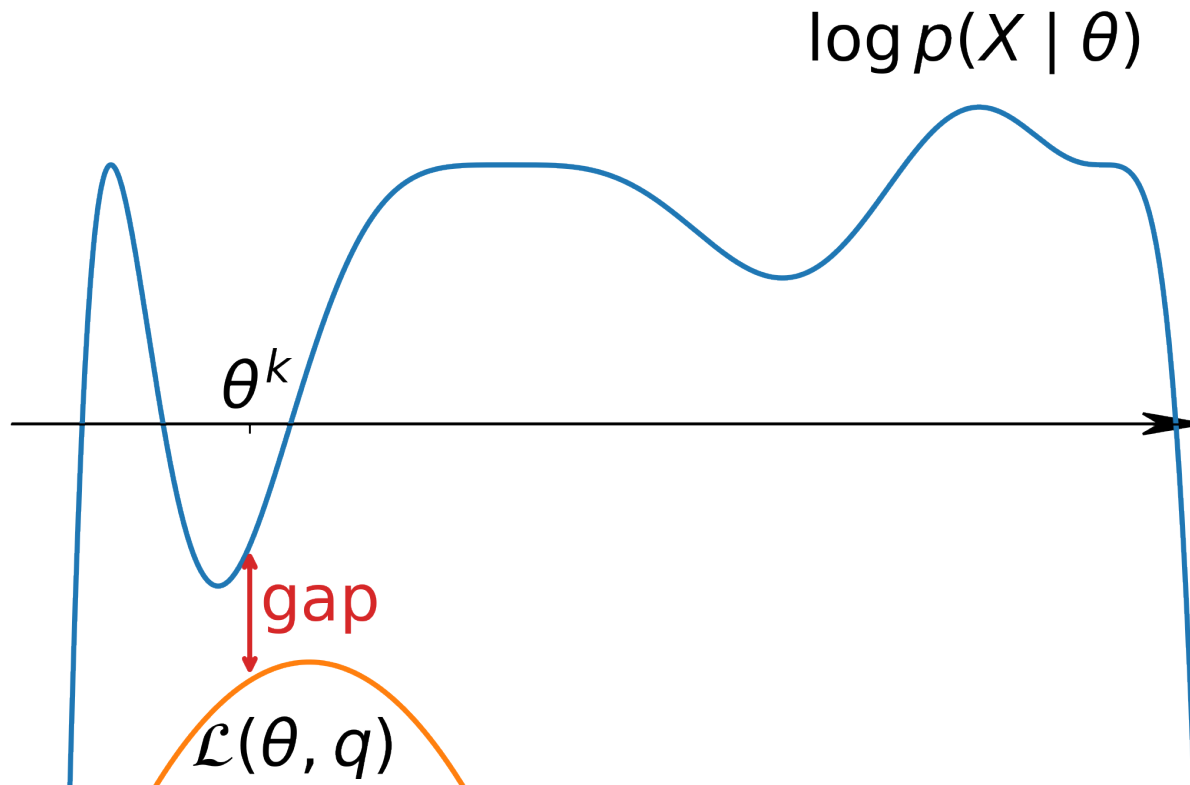


# E-step details

$$\log p(X \mid \theta) \geq \mathcal{L}(\theta, q)$$

**E-step:**  $\max_q \mathcal{L}(\theta^k, q)$



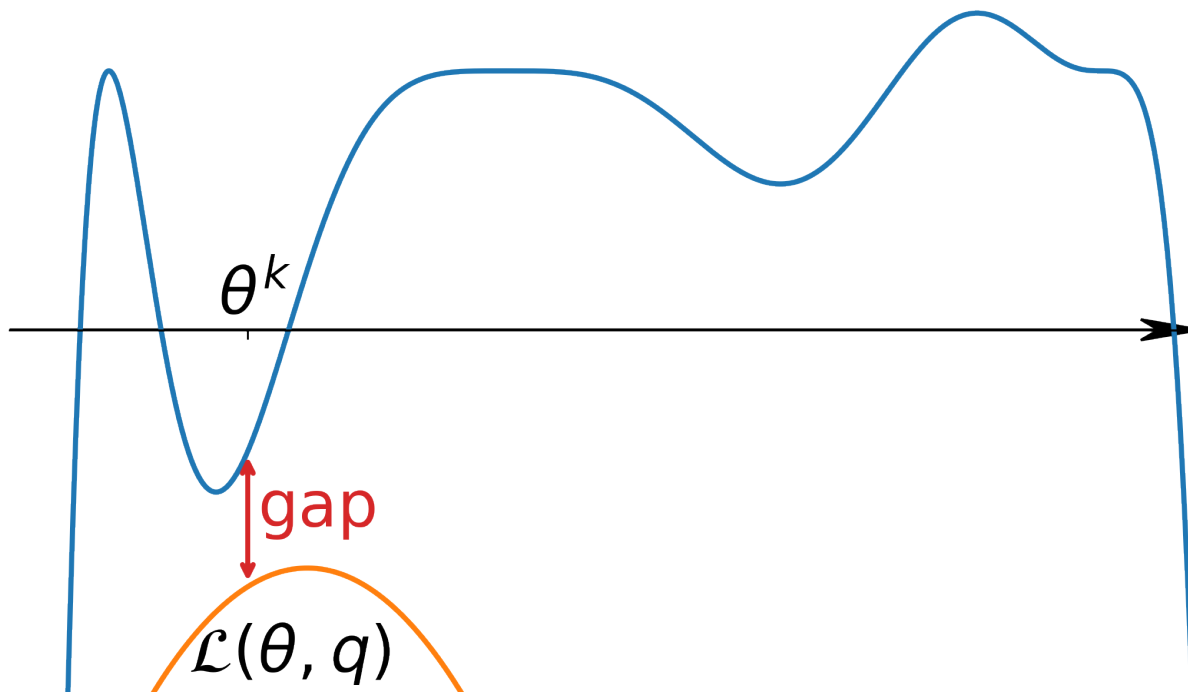
# E-step summary

We want to minimise the GAP between log likelihood and the lower bound. Once its 0, then we got a good fit!

$$\log p(X \mid \theta) - \mathcal{L}(\theta, q) = \sum_i \mathcal{KL}(q(t_i) \parallel p(t_i \mid x_i, \theta))$$

^ This turns out to be equal to the KL divergence! Proof in typora

$$\text{E-step: } \arg \max_{q(t_i)} \mathcal{L}(\theta^k, q) = \frac{p(t_i \mid x_i, \theta)}{\log p(X \mid \theta)}$$



# E-step summary

$$\log p(X \mid \theta) - \mathcal{L}(\theta, q) = \sum_i \mathcal{KL}(q(t_i) \parallel p(t_i \mid x_i, \theta))$$

**E-step:**  $\arg \max_{q(t_i)} \mathcal{L}(\theta^k, q) = p(t_i \mid x_i, \theta)$

$\log p(X \mid \theta)$

