## EECS545 Lecture 15 Quiz Solutions

1. (T/F) The objective function of K-means decreases monotonically.

Solution: True.

- 2. Which of the following about EM algorithm is true? Select all that apply.
  - (a) The EM algorithm can be used for MLE (maximum likelihood) estimate problems involving latent variables.
  - (b) The EM algorithm monotonically increases the lower bound of the log-likelihood  $L(q,\theta)$ .
  - (c) If the posterior  $P(\mathbf{Z} \mid \mathbf{X}; \theta)$  is tractable, the EM algorithm always monotonically increases the observed data log-likelihood of the data.
  - (d) For all complex models where  $P(\mathbf{Z} \mid \mathbf{X}; \theta)$  is not tractable, EM monotonically decreases the log-likelihood of data.
  - (e) The EM algorithm can find the global maximum data likelihood if ran sufficiently long.

**Solution:** (a), (b), and (c). See the lecture notes.

- 3. Which of the following is true about the E-step of the EM algorithm? Select all that apply.
  - (a) E-step computes complete data log-likelihood
  - (b) E-step computes the posterior probability of the latent variables
  - (c) E-step updates the parameters of the model
  - (d) In a single E-step, the log-likelihood of the observed data is increased.
  - (e) In a single E-step, the lower bound  $L(q,\theta)$  of the log-likelihood of the observed data is increased.

**Solution:** (b) and (e). In E-step, we compute the posterior  $P(\mathbf{Z} \mid \mathbf{X}; \theta)$  and set it as  $q(\mathbf{Z})$  given fixed parameters  $\theta$  of the model. This increases the lower bound on the log-likelihood of the observed data, but the log-likelihood of the observed data is kept constant because the parameter does not change.

4. (T/F) q(Z) is fixed during M-step.

Solution: True.

- 5. In the latent variable models we discussed in the class, what does each of the following terms mean, respectively (in the same order)?
  - complete likelihood
  - posterior
  - observed data likelihood
  - (a)  $p(\mathbf{X}, \mathbf{Z} \mid \theta)$   $p(\mathbf{Z} \mid \mathbf{X}, \theta)$   $p(\mathbf{X} \mid \theta)$
  - (b)  $p(\mathbf{X} \mid \theta) = p(\mathbf{X}, \mathbf{Z} \mid \theta) = p(\mathbf{X} \mid \mathbf{Z}, \theta)$
  - (c)  $p(\mathbf{X}, \mathbf{Z} \mid \theta)$   $p(\mathbf{X} \mid \mathbf{Z}, \theta)$   $p(\mathbf{X} \mid \theta)$
  - (d)  $p(\mathbf{X}, \mathbf{Z} \mid \theta) \quad p(\mathbf{X} \mid \theta) \quad p(\mathbf{X} \mid \mathbf{Z}, \theta)$
  - (e)  $p(\mathbf{X}, \mathbf{Z} \mid \theta)$   $p(\mathbf{X} \mid \mathbf{Z}, \theta)$   $p(\mathbf{Z} \mid \mathbf{X}, \theta)$

## Solution: (a)

- complete likelihood:  $p(\mathbf{X}, \mathbf{Z} \mid \theta)$ .
- posterior:  $p(\mathbf{Z} \mid \mathbf{X}, \theta)$ .
- observed data likelihood:  $p(\mathbf{X} \mid \theta)$ .

## Note that

•  $p(\mathbf{X} \mid \mathbf{Z}, \theta)$  is called the conditional data likelihood.