## Problem 1: Bernoulli EM

Consider a unsupervised mixture of Bernoullis model. For each data example i, we have D binary values  $x_{ij}$ , we assume there is an unobserved cluster indicator  $y_i \in \{1 ... K\}$ , and that  $P(x_{ij} \sim \text{Ber}(\mu_{y_{ij}}))$ . In this problem, you will show how to fit this model using EM.

- 1. Define the evidence lower bound  $Q(\theta, \{q_n\})$ . Simplify as much as you can.
- 2. Write an expression that can be used to calculate the cluster responsibility  $r_{ik} = P(y_i = k|\theta)$ .
- 3. Show that the M step for ML estimation of a mixture of Bernoullis is given by

$$\mu_{kj} = \frac{\sum_{i} r_{ik} x_{ij}}{\sum_{i} r_{ik}}$$

4. Show that the M step for MAP estimation of a mixture of Bernoullis with a  $Beta(\alpha, \beta)$  prior is given by

$$\mu_{kj} = \frac{\left(\sum_{i} r_{ik} x_{ij}\right) + \alpha - 1}{\left(\sum_{i} r_{ik}\right) + \alpha + \beta - 2}$$

## Problem 2

In many classification problems one has the option either of assigning x to class j or, if you are too uncertain, of choosing a "reject" option. If the cost for rejects is less than the cost of falsely classifying the object, it may be the optimal action. Let  $\alpha_i$  mean you choose action i, for i = 1: C+1, where C is the number of classes and C+1 is the reject action. Let Y = j be the true (but unknown) state of nature. Define the loss function as follows

$$\lambda(\alpha_i|Y=j) = \begin{cases} 0 & \text{if } i=j \text{ and } i,j \in \{1,\dots,C\} \\ \lambda_r & \text{if } i=C+1 \\ \lambda_s, & \text{otherwise} \end{cases}$$

In other words, you incur 0 loss if you correctly classify, you incur  $\lambda_r$  loss (cost) if you choose the reject option, and you incur  $\lambda_s$  loss (cost) if you make a substitution error (misclassification). Both  $\lambda_r$  and  $\lambda_s$  are positive numbers.

- 1. Show that when we decide to choose a class (and not reject), we always pick the most probable one.
- 2. Show that the minimum risk is obtained if we decide to pick the most probable class  $j_{max} = \arg\max_j p(Y=j|x)$  and if  $p(Y=j_{max}|x) \ge 1 \frac{\lambda_r}{\lambda_s}$ ; otherwise we decide to reject.
- 3. Describe qualitatively what happens as  $\lambda_r/\lambda_s$  is increased from 0 to 1 (i.e., the relative cost of rejection increases).

## Assignment 7

## Problem 3

Please write one thing from this course so far that you found confusing, a topic you would like to hear more about, or something you found particularly interesting.