UNIVERSITY OF ILLINOIS AT URBANA-CHAMPAIGN
Department of Electrical and Computer Engineering

ECE 417 PRINCIPLES OF SIGNAL ANALYSIS
Spring 2014

EXAM 2 SOLUTIONS

Tuesday, April 1, 2014

- This is a CLOSED BOOK exam.

- There are a total of 100 points in the exam. Each problem specifies its point total. Plan your work accordingly.

- You must SHOW YOUR WORK to get full credit.

Problem 1   (15 points)

There have been seven recorded alien invasions of Earth:

(a) May 1902, six Vulcan ships landed in Fort Lauderdale.

(b) December 1928, twelve Vulcan ships landed in Pensacola.

(c) March 1930, four Vulcan ships landed in Miami.

(d) July 1950, eleven Klingon ships landed in Orlando.

(e) August 1992, two Klingon ships landed in St. Augustine.

(f) January 1993, eight Klingon ships landed in Daytona.

(g) May 2003, seven Klingon ships landed in Palm Beach.

The United Nations has commissioned you to create a Classifier of Invasions by Aliens (CIA). Your CIA should be a function defined by

$$f_{CIA}(x) \equiv \Pr\{\text{KLINGONS}|\text{Number of ships} = x\}$$

Draw $f_{CIA}(x)$ as a function of x, for $0 < x < 15$, using a **3-nearest-neighbor** rule to estimate the probability. You may assume that Klingons and Vulcans are the only alien races that exist, thus $\Pr\{\text{KLINGONS}|x\} = 1 - \Pr\{\text{VULCANS}|x\}$
   IMPORTANT: Specify the value of x at each discontinuity.

Solution

$$p(y|x) = \frac{\text{Number of neighbors with label } y}{K}$$

Therefore the pmf changes whenever the set of K nearest neighbors changes. The training data, sorted, are $\{2, 4, 6, 7, 8, 11, 12\}$ with labels $\{K, V, V, K, K, V, K\}$. Potential boundaries therefore occur at $\frac{2+7}{2} = \frac{9}{2}$, $\frac{4+8}{2} = 6$, $\frac{6+11}{2} = \frac{17}{2}$, and $\frac{7+12}{2} = \frac{19}{2}$. The pmf is thefore

$$p(y = K|x) = \begin{cases} \frac{1}{3} & x < 6 \\ \frac{2}{3} & 6 < x \end{cases}$$

Problem 2   (30 points)

A pelican fishes by sweeping its beak through the water. Each sweep catches many fish. The total weight of fish caught in a single sweep is an instance of a random variable, X, that is well described by a Gaussian mixture model:

$$p_X(x) = \sum_{k=1}^{2} c_k \mathcal{N}(x; \mu_k, \sigma_k^2)$$

Unfortunately, you don't know what are the correct values of the parameters c_k, μ_k, and σ_k.

(a) You have received the following suggestions for the parameters. For each candidate set of parameters, say whether or not $p_X(x)$ would be a valid probability density if computed using this set of parameters; if not, say why not.

    (i) Alice suggests $c_1 = 1, c_2 = 1, \mu_1 = 10, \mu_2 = 20, \sigma_1 = 10, \sigma_2 = 10$. Would $p_X(x)$ computed using this parameter set be a valid probability density? If not, why not?
      SOLUTION: No, because $c_1 + c_2 \neq 1$.

    (ii) Barb suggests $c_1 = 0.1, c_2 = 0.9, \mu_1 = 0, \mu_2 = 20, \sigma_1 = 10, \sigma_2 = 10$. Would $p_X(x)$ computed using this parameter set be a valid probability density? If not, why not?
      SOLUTION: yes.

    (iii) Carol suggests $c_1 = 0.5, c_2 = 0.5, \mu_1 = 10, \mu_2 = 20, \sigma_1 = -10, \sigma_2 = 10$. Would $p_X(x)$ computed using this parameter set be a valid probability density? If not, why not?
      SOLUTION: Either no or yes is an acceptable answer, depending on your justification. If you said "no, because standard deviation can not be negative," that would be an acceptable answer. The correct answer, though, is actually "yes, because σ_1^2 is still positive, therefore a normal distribution computed using σ_1^2 as the variance would still be a valid pdf."

(b) You follow a pelican named Pete, and measure the weight of fish he retrieves on four consecutive dips, resulting in the following training dataset:

$$\{x_1, \ldots, x_4\} = \{5, 25, 15, 10\}$$

Using the parameter set $c_1 = 0.5, c_2 = 0.5, \mu_1 = 10, \mu_2 = 20, \sigma_1 = 10, \sigma_2 = 10$, compute $\gamma_k(x_t) = \Pr\left\{k^{\text{th}} \text{ Gaussian} | x_t\right\}$ for $1 \leq t \leq 4$, $1 \leq k \leq 2$. You might find the table of Gaussian PDFs on page 2 of this exam to be useful.

$$\gamma_1(x_1) = \frac{0.035}{0.035 + 0.013}$$

$$\gamma_1(x_2) = \frac{0.013}{0.035 + 0.013}$$

$$\gamma_1(x_3) = \frac{1}{2}$$

$$\gamma_1(x_4) = \frac{0.04}{0.064}$$

$$\gamma_2(x_1) = \frac{0.013}{0.035 + 0.013}$$

$$\gamma_2(x_2) = \frac{0.035}{0.035 + 0.013}$$

$$\gamma_2(x_3) = \frac{1}{2}$$

$$\gamma_2(x_4) = \frac{0.024}{0.064}$$

(c) Recall that the training data are

$$\{x_1, \ldots, x_4\} = \{5, 25, 15, 10\}$$

Suppose that, after a few iterations of EM, you wind up with the following gamma probabilities:

$$\{\gamma_2(x_1), \gamma_2(x_2), \gamma_2(x_3), \gamma_2(x_4)\} = \{0.1, 0.8, 0.6, 0.6\}$$

Find the re-estimated values of c_2, μ_2, and σ_2^2 resulting from this iteration of EM.

Solution:

$$c_2 = \frac{2.1}{4}, \quad \mu_2 = \frac{(0.1)(5) + (0.8)(25) + (0.6)(15) + (0.6)(10)}{2.1},$$

$$\sigma_2^2 = \frac{(0.1)(5 - \mu_2)^2 + (0.8)(25 - \mu_2)^2 + 0.6(15 - \mu_2)^2 + 0.6(10 - \mu_2)^2}{2.1}$$

Problem 3   (15 points)

You're training an audiovisual bird classifier: based on measurements of the birdsong frequency (f) and the bird color (c), the bird is classified as a sparrow ($s = 1$) if and only if

$$\eta \ln p(c|s = 1) + (1 - \eta) \ln p(f|s = 1) > \eta \ln p(c|s = 0) + (1 - \eta) \ln p(f|s = 0)$$

In truth, all sparrows have pitch $f < 0.5$, and color $c < 0.5$, while all other birds have pitch $f > 0.5$ and color $c > 0.5$. Unfortunately, your training algorithm is broken, so it learned these distributions:

$$p(f|s = 0) = \begin{cases} 1 & 0 \leq f \leq 1 \\ 0 & \text{else} \end{cases} \,, \quad p(f|s = 1) = \begin{cases} 1 & 0 \leq f \leq 1 \\ 0 & \text{else} \end{cases} \,, \quad p(c|s = 0) = \begin{cases} 1 & 0 \leq c \leq 1 \\ 0 & \text{else} \end{cases}$$

In fact, only one of the pdfs was learned to be non-uniform:

$$p(c|s = 1) = \begin{cases} 2 - 2c & 0 \le c \le 1 \\ 0 & \text{else} \end{cases}$$

Despite these horrible training results, it is still possible to choose a value of η so that your audiovisual fusion system has zero error. What value of η gives your classifier zero error?

Solution:

Any value of η is OK for which $\{\eta \ln(2 - 2c) > 0\} \Leftrightarrow c < 0.5$, and this is true for any positive value of η.

Problem 4   (15 points)

Good days and bad days follow each other with the following probabilities:

q_{t-1}	$p(q_t = G|q_{t-1} = \cdot)$	$p(q_t = B|q_{t-1} = \cdot)$
G	0.7	0.3
B	0.4	0.6

In winter in Champaign, the temperature on a good day is Gassian with mean $\mu_G = 50$, $\sigma_G = 20$. The temperature on a bad day is Gaussian with mean $\mu_B = 10$, $\sigma_G = 20$. A particular sequence of days has temperatures

$$\{x_1 = 10, x_2 = 20, x_3 = 30\}$$

What is the probability $p(X|q_1 = B)$, the probability of seeing this sequence of temperatures given that the first day was a bad day?

Solution:

$$p(X|q_1 = B) = \left(\frac{1}{50}\right)\left(\frac{1}{20}\right)\left(\frac{1}{20}\right)\left(\frac{6}{25}\right)((0.6)(0.35)(0.6) + (0.4)(0.13)(0.3) + (0.6)(0.35)(0.4) + (0.4)(0.13)(0.7))$$

Problem 5   (25 points)

Suppose that

$$\begin{aligned} a_{ij} &= p(q_t = j|q_{t-1} = i) \\ b_j(x_t) &= p(x_t|q_t = j) \\ g_t &= p(x_t|x_1, \ldots, x_{t-1}) \end{aligned}$$

And define the scaled forward algorithm to compute

$$\tilde{\alpha}_t(i) = p(q_t = i|x_1, \ldots, x_t) = \frac{p(x_t, q_t = i|x_1, \ldots, x_{t-1})}{g_t} = \frac{p(x_1, \ldots, x_t, q_t = i)}{g_1 g_2 \cdots g_t}$$

(a) Devise an algorithm to iteratively compute g_t and $\tilde{\alpha}_t(i)$. Fill in the right-hand side of each equation, using only the terms a_{jk}, $b_j(x_\tau)$, g_τ, and $\tilde{\alpha}_\tau(j)$ for $1 \le j \le N$, $1 \le k \le N$, $1 \le \tau \le t$.

    (i) **INITIALIZE:** $g_1 =$

    (ii) **INITIALIZE:** $\tilde{\alpha}_1(i) =$

    (iii) **ITERATE:** $g_t =$

    (iv) **ITERATE:** $\tilde{\alpha}_t(i) =$

    (v) **TERMINATE:** $p(X) =$

Solution:

    (i) **INITIALIZE:** $g_1 = \sum_{j=1}^{N} \pi_j b_j(x_1)$

    (ii) **INITIALIZE:** $\tilde{\alpha}_1(i) = \frac{\pi_i b_i(x_1)}{g_1}$

    (iii) **ITERATE:** $g_t = \sum_{i=1}^{N} \sum_{j=1}^{N} \tilde{\alpha}_{t-1}(i) a_{ij} b_j(x_t)$

    (iv) **ITERATE:** $\tilde{\alpha}_t(i) = \frac{\sum_{j=1}^{N} \tilde{\alpha}_{t-1}(i) a_{ij} b_j(x_t)}{g_t}$

    (v) **TERMINATE:** $p(X) = \prod_{t=1}^{T} g_t$

(b) Suppose $\beta_t(i) = p(x_{t+1}, \ldots, x_T | q_t = i)$. Then

$$\tilde{\alpha}_t(i)\beta_t(i) = p(f|g)$$

for some list of variables f, and some other list of variables g. Specify what variables should be included in each of these two lists.

Solution:

$$f = \{q_t = i, x_{t+1}, \ldots, x_T\}$$
$$g = \{x_1, \ldots, x_t\}$$