

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

ECE 417 PRINCIPLES OF SIGNAL ANALYSIS  
Spring 2014

**EXAM 2**

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 | Score |
|---------|-------|
| 1       |       |
| 2       |       |
| 3       |       |
| 4       |       |
| 5       |       |
| Total   |       |

Name: \_\_\_\_\_

## Gaussian Probability Densities (to Two Significant Figures)

| $x$ | $\frac{1}{\sqrt{2\pi}}e^{-x^2/2}$ |
|-----|-----------------------------------|
| 0   | 0.40                              |
| 0.5 | 0.35                              |
| 1   | 0.24                              |
| 1.5 | 0.13                              |
| 2   | 0.05                              |
| 2.5 | 0.02                              |
| 3   | 0.00                              |

## Other Possibly Useful Formulas

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n}$$

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega})e^{j\omega n} d\omega$$

$$h[n] = \frac{\sin \omega_c n}{\pi n} \leftrightarrow H(e^{j\omega}) = \begin{cases} 1 & |\omega| < \omega_c \\ 0 & \text{otherwise} \end{cases}$$

$$u[n] - u[n - N] \leftrightarrow e^{-j\frac{\omega(N-1)}{2}} \frac{\sin(\omega N/2)}{\sin(\omega/2)}$$

$$\delta[n] \leftrightarrow 1$$

$$e^{j\alpha n} \leftrightarrow 2\pi\delta(\omega - \alpha)$$

$$X[k] = \sum_{n=0}^{N-1} x[n]e^{-j2\pi kn/N}$$

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k]e^{j2\pi kn/N}$$

$$S = \sum_{k=1}^n (\vec{x}_k - \vec{m})(\vec{x}_k - \vec{m})^T$$

**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.**

**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  $c_k$ ,  $\mu_k$ , and  $\sigma_k$ .

- (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?
  
  
  
  
  
  
  
  
  
  
  - (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?
  
  
  
  
  
  
  
  
  
  
  - (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?

- (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:

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

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  $\gamma_k(x_t) = \Pr\{k^{\text{th}} \text{ Gaussian} | x_t\}$  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.

(c) Recall that the training data are

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

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  $c_2$ ,  $\mu_2$ , and  $\sigma_2^2$  resulting from this iteration of EM.

**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:

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

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

**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 Gaussian 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?

**Problem 5 (25 points)**

Suppose that

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

And define the scaled forward algorithm to compute

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

- (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  $a_{jk}$ ,  $b_j(x_\tau)$ ,  $g_\tau$ , and  $\tilde{\alpha}_\tau(j)$  for  $1 \leq j \leq N$ ,  $1 \leq k \leq N$ ,  $1 \leq \tau \leq t$ .

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

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

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

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

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

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

$$f = \{ \quad \quad \quad \}$$

$$g = \{ \quad \quad \quad \}$$