

EE 641 Midterm Exam  
Fall 2004

Name: \_\_\_\_\_ Starting time: \_\_\_\_\_  
Ending time: \_\_\_\_\_

**Instructions**

The following is a take home exam.

- This exam contains 4 problems and should be completed in 75 minutes.
- Answer questions precisely and completely. Credit will be subtracted off for vague answers.
- You may not use any notes, textbooks, or calculators; but you may use a single sided single page of hand copied notes.

Good luck.

**Problem 1.** (20pt)

Let  $\{X_n\}_{n=0}^N$  be a Markov Chain with

$$P\{X_0 = k\} = \pi_k$$

for  $1 \leq k \leq M$ , and

$$P\{X_n = m | X_{n-1} = l\} = P_{l,m}$$

for  $1 \leq l, m \leq M$ . Let  $\pi$  denote the  $1 \times M$  vector of initial probabilities, and let  $P$  denote the  $M \times M$  transition matrix.

- a) Derive an expression,  $p(x|\theta)$ , the likelihood for the entire sequence.
- b) Derive the ML estimate of  $\theta = (\pi, P_{l,m})$  given  $\{X_n\}_{n=0}^N$ .

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**Problem 2.** (25pt)

Let  $\{X_n\}_{n=0}^{N-1}$  be i.i.d. random variables with

$$P\{X_n = k\} = \pi_k$$

for  $0 \leq k \leq M - 1$ .

Let  $\{Y_n\}_{n=0}^{N-1}$  be random variables with conditional distribution

$$p(y|x, \theta) = \prod_{n=0}^{N-1} \frac{1}{\mu_{x_n}} \exp\{-y_n/\mu_{x_n}\} \mathbf{1}(y_n \geq 0)$$

where  $\mathbf{1}(y_n \geq 0)$  is function which is 1 when  $y_n \geq 0$  and zero otherwise, and  $\theta = (\pi, \mu)$

- a) Derive  $p(y, x|\theta)$ .
- b) Derive  $f(x_n|y_n, \theta) = p(x_n|y_n, \theta)$ .
- c) Derive the ML estimate of  $\theta$  given both  $Y$  and  $X$ .
- d) Derive the EM update equations for estimation of  $\theta$  from  $Y$ .

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**Problem 3.** (30pt)

Let  $X$  be an  $N$  pixel Gaussian MRF with density

$$p(x) = \frac{1}{(2\pi\sigma^2)^{N/2}} |B|^{N/2} \exp \left\{ -\frac{1}{2\sigma^2} x^t B x \right\}$$

Furthermore, let  $Y$  be an  $N$  pixel random field with pixels that are conditionally i.i.d. and Poisson given  $X$ , and

$$E[Y|X] = X$$

- a) Calculate the maximum likelihood estimate of  $\sigma^2$  given  $X$  and  $Y$ .
- b) Calculate  $p(y, x|\sigma)$ , the joint probability density of  $Y$  and  $X$  given  $\sigma$ .
- c) Calculate an expression for the MAP estimate of  $X$  given  $Y$ .
- d) Derive an expression for the gradient descent update of the MAP cost functional.
- e) Derive an expression for the ICD update of a signal pixel  $s$ .

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**Problem 4.** (25pt)

Consider the cost functional

$$\|y - Hx\|^2 + x^t Bx$$

which takes on its unique minimum value at

$$\hat{x} = \arg \min_{x \in \mathbb{R}^N} \left\{ \|y - Hx\|^2 + x^t Bx \right\}$$

a) Derive the gradient descent algorithm for this minimization problem, and show it has the form of the difference equation

$$x^{(k+1)} = (I - \omega P) x^{(k)} + \omega H^t y$$

where  $P = H^t H + B$ .

Further assume that  $P = T^t \Lambda T$  where  $T$  is an orthonormal matrix, and  $\Lambda$  is a diagonal matrix with positive entries  $\lambda_1, \dots, \lambda_N$ .

b) Let  $e^{(k)} = x^{(k)} - \hat{x}$ , and derive a difference equation for  $e^{(k)}$ .

c) Determine the range of values of  $\omega$  that will allow stable convergence of the gradient descent algorithm.

d) Explain what the trade-offs are in the selection of  $\omega$ .



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