

EE 641 Final Exam  
December 12, Fall 2022

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

**Q1: Instructions (4pt)**

**Rules:** I understand that this is an open book exam that shall be done within the allotted time of 180 minutes. I can use my notes, previous posted exams and exam solutions, and web resources. However, I will not communicate with any other person other than the official exam proctors during the exam; I will not seek or accept help from any other persons other than the official proctors; and I will not use GPT-3 or any other variant of an AI response engine.

**Signature:** \_\_\_\_\_

**Q2: Generating Random Variables** (10pt)

Let  $X$  be a random variable with the CDF given by

$$F(\lambda) = P\{X \leq \lambda\} ,$$

where  $F$  is continuous and strictly monotone increasing.

**Q2.1:**

Give a method for generating a new random variable,  $X'$ , with the same distribution as  $X$ .

**Q2.2:**

Prove that if  $F'(\lambda)$  is the CDF of  $X'$ , then  $F'(\lambda) = F(\lambda)$ .

**Q3: Properties of Discrete Distribution (25pt)**

Consider  $X = (X_0, \dots, X_{N-1})$  where  $X_n$  are i.i.d. random variables such that

$$P\{X_n = i\} = \theta_i .$$

Also let  $p_\theta(x)$  denote the associated family of distributions such that  $\theta \in S$  where  $S$  denotes the  $M$  dimensional simplex given by

$$S = \left\{ \theta \in S : \forall i \in \{0, \dots, M-1\} , \theta_i \geq 0 \text{ and } \sum_{i=0}^{M-1} \theta_i = 1 \right\} .$$

**Q3.1:**

Show that

$$K_i = \sum_{n=0}^{N-1} \delta(X_n - i) ,$$

is a sufficient statistic for the family of distributions  $p_\theta(x)$ .

**Q3.2:**

Show that  $p_\theta(x)$  is an exponential distribution with natural sufficient statistics of  $\{K_i\}_{i=0}^{M-1}$ .

**Q3.3:**

Derive the maximum likelihood estimate of  $\theta$  given the observations  $X$ .

**Q4: ADMM Optimization (20pt)**

Let  $X = (X_0, \dots, X_{N-1})$  be i.i.d. random variables with distribution

$$P\{X_n = m\} = \pi_m ,$$

where  $\pi_i \geq 0$  and  $\sum_{m=0}^{M-1} \pi_m = 1$ .

Also, let  $Y = (Y_0, \dots, Y_{N-1})$  be conditionally independent discrete random variables given  $X$  with each  $Y_n$  having the conditional distribution given by

$$P\{Y_n = j | X_n = i\} = P_{i,j}$$

where  $P_{i,j} \geq 0$  and  $\sum_{j=0}^{M-1} P_{i,j} = 1$ .

Furthermore, let  $\theta = (\pi_0, P_{0,0}, \dots, P_{0,M-1}, \dots, \pi_{M-1}, P_{M-1,0}, \dots, P_{M-1,M-1})$  parameterize the model.

**Q4.1:**

Using the statistic,

$$N_i = \sum_{n=0}^{N-1} \delta(X_n - i) ,$$

write out an expression for the density of  $X$ .

**Q4.2:**

Using the statistic,

$$K_{i,j} = \sum_{n=0}^{N-1} \delta(X_n - i) \delta(Y_n - j) ,$$

write out an expression for the conditional density of  $Y$  given  $X$ .

**Q4.3:**

Write out the negative log likelihood,  $-\log p_\theta(x, y)$ , in terms of the sufficient statistics  $N_i$  and  $K_{i,j}$ .

**Q4.4:**

Write out the maximum likelihood estimate of  $\theta$  given the complete data,  $(X, Y)$ .

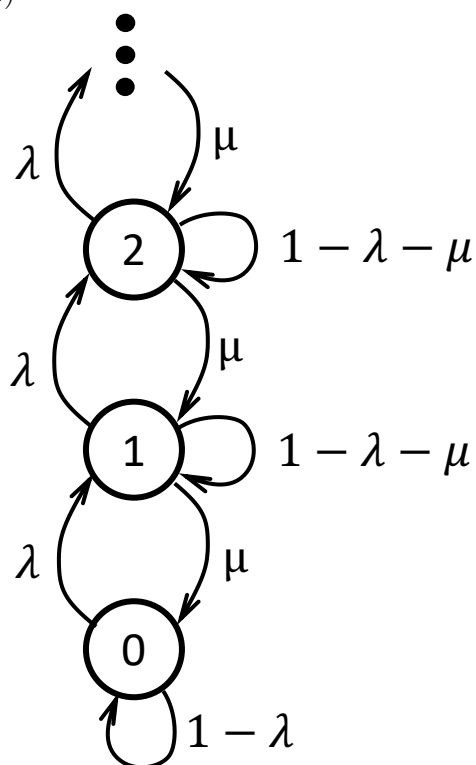
**Q4.5:**

Write out the explicit expression for the E-step of the EM algorithm for estimating  $\theta$  given  $Y$ .

**Q4.6:**

Write out the explicit expression for the M-step of the EM algorithm for estimating  $\theta$  given  $Y$ .

**Q5: Markov Chain** (25pt)



Let  $\{X_n\}_{n=0}^{\infty}$  be a homogeneous Markov chain with states  $\{0, 1, 2, \dots\}$  and state-transition diagram as shown above. Furthermore, assume that  $\rho = \lambda/\mu < 1$ .

**Q5.1:**

Write out an explicit form for the state transition probabilities,  $P_{i,j}$ .

**Q5.2:**

Is there a solution to the detailed balance equations for this Markov chain? If so, give the solution.

**Q5.3:**

Is there a solution to the full balance equations for this Markov chain? If so, give the solution.

**Q5.4:**

Is the Markov chain reversible? Justify your answer.

**Q5.5:**

Determine the stationary distribution for the Markov chain.

**Q6: Plug-and-Play Methods (25pt)**

Define

$$f(x) = \frac{1}{2\sigma_y^2} \|y - Ax\|^2$$

and its associated proximal map as

$$\hat{x} = F(z) = \arg \min_x \left\{ f(x) + \frac{1}{2\sigma^2} \|x - z\|^2 \right\} .$$

Let  $H(z)$  be a firmly non-expansive function so that

$$X \approx H(X + W) ,$$

where  $X$  is a typical image and  $W \sim N(0, \sigma^2 I)$ . Then define

$$T = (2H - I)(2F - I) .$$

Furthermore, assume that the fixed point problem  $Tw^* = w^*$  has a unique solution denoted by  $w^*$ .

**Q6.1:**

Use the results of Appendix B, Properties B.5, B.3, and B.1 to prove that  $T$  is non-expansive.

**Q6.2:**

Give an algorithm for computing the solution to the fixed point problem  $Tw^* = w^*$ . Why do you know that this algorithm converges to  $w^*$ .

**Q6.3:**

Prove that there is a solution to the equilibrium equation

$$F(x^* - u^*) = x^*$$

$$H(x^* + u^*) = x^* .$$

(Hint: Reverse the argument of Section 10.3.3 page 161.)

**Q6.4:**

In the equilibrium equations,

$$F(x^* - u^*) = x^*$$

$$H(x^* + u^*) = x^* ,$$

give an interpretation for the quantities  $x^*$  and  $u^*$ .

**Q6.5:**

Explain how one might obtain an agent,  $H(z)$ ?

**Q6.6:**

What is the advantage of this approach over more conventional MAP estimation?