

## Homework Assignment 0

### Statistical Methods for Analysis with Missing Data, Winter 2019

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Submit your solutions via Canvas. Feel free to hand-write your solutions and submit a scanned copy. Due by 12:00pm (noon) on Jan 16, 2019.

The grade for this assignment will be either 1 (satisfactory) or 0. This assignment contains four problems, each of which contains subproblems, totaling 17 of them. Your solutions will be graded as 1 if you have at least 10 correct subproblems, including one correct subproblem from each problem.

1. Let the random variables  $X, Y, Z, W$  have a joint distribution with density  $p$ , such that

$$p(x, y, z, w) = p_X(x)p_{Y|X}(y | x)p_{Z|X}(z | x)p_{W|X,Z}(w | x, z),$$

where  $p_X$ ,  $p_{Y|X}$ ,  $p_{Z|X}$ , and  $p_{W|X,Z}$  are the densities of the distributions of  $X$ ,  $(Y | X)$ ,  $(Z | X)$ , and  $(W | X, Z)$ , respectively.

- (a) Determine which of the following independence statements are true, and provide proofs for the true statements:
  - i.  $Y \perp\!\!\!\perp W$ ;   ii.  $Y \perp\!\!\!\perp W | X$ ;   iii.  $W \perp\!\!\!\perp (X, Z)$ ;
  - iv.  $Z \perp\!\!\!\perp Y | X$ ;   v.  $Z \perp\!\!\!\perp (Y, W) | X$ .
- (b) Write down the conditional density of:
  - i.  $X | Y, Z, W$ ;   ii.  $X | Y, Z$ ;   iii.  $X | Y, W$
- (c) Write down the expression for the expected value of  $XW$ ,  $E(XW)$ , in terms of the density  $p$ .
- (d) Say we use the simpler notation

$$p(x, y, z, w) = p(x)p(y | x)p(z | x)p(w | x, z)$$

to denote the  $p$  density above. Why is this notation inaccurate?

2. Let  $X_1, \dots, X_K \stackrel{i.i.d.}{\sim} \text{Bernoulli}(\pi)$ .
  - (a) Write down the probability mass function of the random vector  $X = (X_1, \dots, X_K)$ , denoted  $p(x | \pi)$ , where  $x$  is a generic realization of  $X$
  - (b) Write down the likelihood function of  $\pi$ . What is the difference between the likelihood function and the probability mass function derived in (a)?
  - (c) Derive the maximum likelihood estimator of  $\pi$ , and call it  $\hat{\pi}$
  - (d) Compute the expected value of  $\hat{\pi}$
  - (e) Compute the expectation  $E_X[\log p(X | \pi')]$ ,  $\pi' \neq \pi$
  - (f) Find the value that maximizes  $E_X[\log p(X | \pi')]$  as a function of  $\pi'$
  - (g) Suppose that  $K$  is an even number. Let  $Y_1 = (X_1, \dots, X_{K/2})$  and  $Y_2 = (X_{K/2+1}, \dots, X_K)$ . Derive the distribution of  $W = Y_1 Y_2^T$ .  $T$  denotes the transpose operator.
  - (h) Derive the conditional distribution of  $X$  given  $XX^T$
3. Let  $X = (X_1, \dots, X_K) \sim \text{MultivariateNormal}(\mu, \Sigma)$ . Let  $A$  be a  $K \times B$  matrix,  $B \leq K$ , such that its  $(i, j)$ th entry  $a_{ij} \in \{0, 1\}$ ,  $\sum_{j=1}^B a_{ij} \leq 1$  for all  $i = 1, \dots, K$ , and  $\sum_{i=1}^K a_{ij} \geq 1$  for all  $j = 1, \dots, B$ . For this problem you can use commonly known facts about the multivariate normal distribution.
  - (a) What is the distribution of  $Y = XA$ ? Explain.
  - (b) What is the distribution of  $(Y_1, Y_2) | (Y_3, \dots, Y_B)$ . Explain.
4. Let  $X = (X_1, \dots, X_K) \sim \text{Multinomial}(N, (\pi_1, \dots, \pi_K))$ .
  - (a) Derive the distribution of  $(X_1 + X_2, X_3, \dots, X_K)$ .
  - (b) What is the distribution of  $Y = XA$ , where  $A$  is a matrix as in problem 3., except that here  $\sum_{j=1}^B a_{ij} = 1$  for all  $i = 1, \dots, K$ . No need to write down the proof.
  - (c) Suppose that  $K$  is an even number. Let  $Y_1 = (X_1, \dots, X_{K/2})$  and  $Y_2 = (X_{K/2+1}, \dots, X_K)$ . Derive the distribution of  $X = (Y_1, Y_2)$  given  $Y_1 Y_1^T$  and  $Y_2 Y_2^T$ .  $T$  denotes the transpose operator.

Note: 4(c) is complicated in general. Acceptable solutions include (in order of difficulty): i. solve for  $N = 1$ ; ii. simply write generic expression for the conditional distribution for generic  $N$ . iii. solve for  $K = 4$  and generic  $N$ .