

## Quiz Section #9: Supplementary Exercises

CSE 312: Foundations of Computing II

- Suppose  $x_1, x_2, \dots, x_n$  are independent samples from a normal distribution whose mean is known to be zero, but whose variance is unknown. What is the maximum likelihood estimator for its variance?
  - Suppose the mean is known to be  $\mu$  but the variance is unknown. How does the maximum likelihood estimator for the variance differ from the maximum likelihood estimator when both mean and variance are unknown?
- Let  $f(x | \theta) = \theta x^{\theta-1}$  for  $0 \leq x \leq 1$ , where  $\theta$  is any positive real number. Let  $x_1, x_2, \dots, x_n$  be independent samples from this distribution. Derive the maximum likelihood estimator  $\hat{\theta}$ .
- You are given 100 independent samples  $x_1, x_2, \dots, x_{100}$  from  $\text{Ber}(p)$ , where  $p$  is unknown. These 100 samples sum to 30. You would like to estimate the distribution's parameter  $p$ . Give all answers to 3 significant digits.
  - What is the maximum likelihood estimator  $\hat{p}$  of  $p$ ?
  - Is  $\hat{p}$  an unbiased estimator of  $p$ ?
  - Give your best approximation for the 95% confidence interval of  $p$ .
  - Give your best approximation for the 90% confidence interval of  $p$ .
  - Give three different reasons why your answers to (c) and (d) are only approximations.
  - Explain why it makes sense that the interval in (d) is bigger (or smaller, depending on your answers) than the interval in (c).
- Suppose that  $\hat{\theta}$  is a biased estimator for  $\theta$  with  $E[\hat{\theta}] = \alpha\theta$ , for some constant  $\alpha > 0$ . Find an unbiased estimator for  $\theta$  and prove that it is unbiased.
  - In lecture, we saw that the maximum likelihood estimator for the population variance  $\theta_2$  of  $N(\theta_1, \theta_2)$  is the sample variance

$$\hat{\theta}_2 = \frac{1}{n} \sum_{i=1}^n (x_i - \hat{\theta}_1)^2$$

where  $\hat{\theta}_1$  is the sample mean. It can be shown that  $E[\hat{\theta}_2] = \frac{n-1}{n} \cdot \theta_2$ , so that  $\hat{\theta}_2$  is biased and always underestimates the variance  $\theta_2$ . Use your result from part (a) to find an unbiased estimator of the variance  $\theta_2$ .