

BIOS760: 2011 FALL SEMESTER FINAL EXAM

1. Assume that X_1, \dots, X_n are i.i.d from the uniform distribution in $[-\theta, \theta]$, where θ is a positive and unknown parameter. Let $|X|_{(n)}$ denote $\max(|X_1|, \dots, |X_n|)$ and $g(\theta)$ be a continuously differentiable function of θ satisfying $g'(\theta) > 0$.

- (a) (5 points) Show that $|X|_{(n)}$ is a complete and sufficient statistic for θ .
- (b) (5 points) Show that

$$n \left(1 - \frac{|X|_{(n)}}{\theta} \right)$$

converges in distribution to a non-degenerate distribution. Give the explicit expression of the limiting distribution.

- (c) (5 points) What is the maximum likelihood estimator for $g(\theta)$? Derive the asymptotic distribution of the maximum likelihood estimator. Furthermore, construct an asymptotic confidence interval with coverage $(1 - \alpha)$ for $g(\theta)$ based on the maximum likelihood estimator.
- (d) (5 points) Find the UMVUE for $g(\theta)$.
- (e) (5 points) Derive the asymptotic distribution of the UMVUE. Furthermore, construct an asymptotic confidence interval with coverage $(1 - \alpha)$ for $g(\theta)$ based on the UMVUE.

2. Let X and ξ be two independent random variables and moreover, ξ follows a gamma distribution with density $\Gamma(\lambda)^{-1} \lambda^\lambda \xi^{\lambda-1} \exp\{-\lambda\xi\}$ where λ is a known constant larger than 1. We assume that conditional on (X, ξ) , a response variable Y follows the distribution

$$P(Y > t|X, \xi) = \exp\{-te^{\beta X}\xi\},$$

where β is an unknown parameter. In a real study, ξ is not observed so we only obtain n i.i.d data $(X_i, Y_i), i = 1, \dots, n$.

- (a) (5 points) Write down the conditional density of Y given X then give the observed likelihood function.
- (b) (2 points) Under what condition(s) is β identifiable?
- (c) (5 points) Show that $E[Y|X] = ce^{-\beta X}$ for some constant c and identify the constant c . Thus, we can construct an estimating equation for β as

$$\sum_{i=1}^n X_i \{Y_i - ce^{-\beta X_i}\} = 0.$$

Give the asymptotic distribution of the estimator which solves this equation. You do not need justify this distribution and may leave mathematical expressions in the result.

- (d) (5 points) Write the score equation for the maximum likelihood estimator and derive the Newton-Raphson iteration for solving this equation.
- (e) (8 points) Another way of calculating the maximum likelihood estimator is via the EM algorithm. In this case, we treat ξ_1, \dots, ξ_n as missing data. Give the details in the E-step and M-step.