

本試題是否可以使用計算機： 可使用； 不可使用 (請命題老師勾選)

考試日期：0302，節次：2

- I. (40 points) Suppose that random variable  $X$  takes values  $-1, 0, 1$  with respective probabilities  $p_1 = \theta^2$ ,  $p_2 = 2\theta(1-\theta)$ ,  $p_3 = (1-\theta)^2$  given by the Hardy-Weinberg proportions,  $0 < \theta < 1$ . A sample size of  $n$ ,  $X_1, X_2, \dots, X_n$ , are obtained.
- Find the likelihood function.
  - Find the maximum likelihood estimator (MLE)  $T$  of  $\theta$ .
  - Let  $H_0: \theta = \theta_0$  vs  $H_1: \theta > \theta_0$ . Based on minimal sufficient statistic, obtain the uniformly most powerful test (UMPT) at level  $\alpha$ .
  - Find  $E(T)$  and  $\text{Var}(T)$ .

- II. (24 points) A random variable  $X$  with double exponential distribution has the following *p.d.f.*

$$f_{\mu, \lambda}(x) = \frac{1}{2} \lambda e^{-\lambda|x-\mu|}, x \in R, \mu \in R, \lambda > 0.$$

Suppose a random sample  $X_1, X_2, \dots, X_n$  is available,  $n=2m+1$ ,  $m$  is positive integer.

- For  $\lambda$  fixed, plot the likelihood function for  $\mu$ . For simplicity, you can take  $m=1$ .
  - (continued). Find the maximum likelihood estimator for  $\mu$ .
  - If now  $\mu = \mu_0$  is known, but  $\lambda$  is unknown, construct a  $100(1-\alpha)\%$  confidence interval for  $\lambda$  based on the minimal sufficient statistic,  $0 < \alpha < 1$ .
- III. (36 points) In a two-color microarray experiment, the gene expression level is usually represented as  $R/G$ , where  $R$  is the fluorescence intensity of Cye 5 and  $G$  is the fluorescence intensity of Cye 3. It is a common practice that both  $R$  and  $G$  follow lognormal distributions, and both are correlated, although its association is mild if the experimental process is under control. Suppose  $(\ln R, \ln G)$  follows a bivariate normal with mean  $\mu_1, \mu_0$ , variance  $\sigma_1^2, \sigma_0^2$  and correlation  $\rho$ . Let  $M = \ln R - \ln G$ , the differential gene expression level, and  $A = (\ln R + \ln G)/2$ , the abundance of mRNA.
- Find  $E(R) = \mu_R$  and  $\text{Var}(R) = \sigma_R^2$ , and show that  $E(R)$  has something to do with the  $\text{Var}(R)$  is a function of  $E(R)$ .
  - If  $\sigma_R^2 = f(\mu_R)$ , where  $f(\cdot)$  is a function with continuous derivative, find a transformation of  $R$ ,  $g(R)$ , so that  $\text{Var}(g(R))$  is roughly free of  $\mu_R$ .
  - Find the joint distribution of  $(M, A)$ .
  - Find  $E(M|A)$ , the conditional expectation of  $M$  given  $A$ . Under which condition(s) that  $A$  has nothing to do with  $M$ ?