

Statistics 581, Final Exam

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This exam is to be taken without the use of any books or notes.

1. (40) points) **Define** each of the following terms. In each case, briefly provide an appropriate context for your definition.
 - (a) The information matrix for θ in a regular parametric model $\mathcal{P} = \{P_\theta : \theta \in \Theta \subset R^d\}$.
 - (b) The efficient score function for a parameter θ_1 when $\theta = (\theta_1, \theta_2)$.
 - (c) The efficient influence function \tilde{l}_1 for a parameter θ_1 when $\theta = (\theta_1, \theta_2)$.
 - (d) The efficient influence function for \tilde{l}_ν for a differentiable parameter $q(\theta) = \nu(P_\theta)$ in a regular parametric model \mathcal{P} .
 - (e) An asymptotically linear estimator of a parameter $\nu(P)$ with influence function ψ .

2. (32 points) **State** four of the following five results, providing the appropriate (brief) context for your statement:
 - (a) The (elementary) Skorokhod theorem.
 - (b) The Cramér - Wold device.
 - (c) A result about the finite-dimensional limiting distributions of the sample quantile process $\{\sqrt{n}(\mathbb{F}_n^{-1}(t) - F^{-1}(t)) : 0 < t < 1\}$ specifying the assumption(s) carefully.
 - (d) Le Cam's third lemma.
 - (e) The Glivenko-Cantelli theorem.

Do either problem 3 or problem 4:

3. (50 points). (A two-sample model.) Suppose that $Z \sim \text{Bernoulli}(\eta)$, and conditional on Z , a random variable X has conditional distributions given Z described by $(X|Z) \sim \text{Exponential}(\mu Z + \nu(1 - Z))$ where $\mu, \nu > 0$. Thus

$$\begin{aligned} P(X > x|Z = 1) &= \exp(-\mu x), \quad x \geq 0, \\ P(X > x|Z = 0) &= \exp(-\nu x), \quad x \geq 0. \end{aligned}$$

- (a) Let $\theta = (\mu, \nu, \eta) \equiv (\theta_1, \theta_2, \theta_3) \in (0, \infty) \times (0, \infty) \times (0, 1)$, and write $(X, Z) \sim P_\theta$. Show that the joint density p_θ of (X, Z) (with respect to the product of Lebesgue measure λ on $(0, \infty)$ and counting measure $\#$ on $\{0, 1\}$) is given by

$$p_\theta(x, z) = (\eta\mu e^{-\mu x})^z ((1 - \eta)\nu e^{-\nu x})^{1-z} 1_{(0, \infty)}(x) 1_{\{0, 1\}}(z).$$

- (b) Find the score(s) for $\theta = (\mu, \nu, \eta)$ and the information matrix for θ (when $n = 1$).
- (c) If $(X_1, Z_1), \dots, (X_n, Z_n)$ are i.i.d. as $(X, Z) \sim P_{\theta_0}$, what is the Cramér - Rao bound for estimation of $\nu(P_\theta) \equiv \theta_1 - \theta_2 = \mu - \nu$?
- (d) For a sample of (X, Z) pairs as in (c), find the maximum likelihood estimator

$\hat{\theta}_n$ of θ .

(e) Find the limiting distribution of $\sqrt{n}(q(\hat{\theta}_n) - q(\theta_0))$ when $\theta = \theta_0$ is true.

(f) Propose a test statistic for testing $H : \mu = \nu$ versus $K : \mu \neq \nu$ based on the result of (e). What is the limiting distribution of your test statistic under H ? What is the limiting distribution of your test statistic under local alternatives of the form $\mu = \nu + tn^{-1/2}$?

4. (48 points).

Suppose that X_1, \dots, X_n are independent and identically distributed real-valued random variables with distribution function F and density f .

(a) Consider the sample median $\mathbb{F}_n^{-1}(1/2)$ and the sample mean \bar{X}_n . Give conditions under which the sample median $\mathbb{F}_n^{-1}(1/2)$ is an asymptotically linear estimator of the population median $F^{-1}(1/2)$. Identify the influence function $\psi(x)$ and the limiting distribution of $\sqrt{n}(\mathbb{F}_n^{-1}(1/2) - F^{-1}(1/2))$.

(b) Give conditions under which the sample median $\mathbb{F}_n^{-1}(1/2)$ and the sample mean \bar{X}_n have a joint limiting distribution; i.e. conditions which imply that the random vector

$$\begin{pmatrix} \sqrt{n}(\mathbb{F}_n^{-1}(1/2) - F^{-1}(1/2)) \\ \sqrt{n}(\bar{X}_n - \mu) \end{pmatrix}$$

converges in distribution where $\mu = \mu_F = E_F(X_1)$. Find the limiting distribution explicitly.

(c) A simple test for asymmetry of a distribution function is based on the difference of the mean and median: $\gamma(F) \equiv \mu_F - F^{-1}(1/2)$. Note that $\gamma_F = 0$ if F is symmetric about some point, while $\gamma(F)$ is positive for F skewed to the right, and negative for F skewed to the left. Use the results of (b) to find the limiting distribution of

$$\sqrt{n}(\gamma(\mathbb{F}_n) - \gamma(F)) = \sqrt{n}(\bar{X}_n - \mathbb{F}_n^{-1}(1/2) - (\mu_F - F^{-1}(1/2))).$$

Compute the limiting variance in terms of expectations of functions of F . Is $\gamma(\mathbb{F}_n)$ asymptotically linear?

Do either problem 5 or problem 6.

5. (48 points).

Suppose that $\underline{N}_n \equiv \sum_{i=1}^n \underline{\Delta}_i \sim \text{Mult}_k(n, \underline{p})$ where $\underline{\Delta}_i \sim \text{Mult}_k(1, \underline{p})$ are independent. The usual MLE $\hat{\underline{p}}_n$ of \underline{p} is $\hat{\underline{p}}_n = n^{-1}\underline{N}_n$. For $a \in [0, 1)$ and $\underline{p}_0 = (1/k, \dots, 1/k)$, consider the alternative estimator $\tilde{\underline{p}}_n$ of \underline{p} defined by

$$\tilde{\underline{p}}_n = \begin{cases} \hat{\underline{p}}_n & \text{if } \|\hat{\underline{p}}_n - \underline{p}_0\|_2 > n^{-1/4}, \\ a\hat{\underline{p}}_n + (1-a)\underline{p}_0 & \text{if } \|\hat{\underline{p}}_n - \underline{p}_0\|_2 \leq n^{-1/4}. \end{cases}$$

Let $\underline{Z} \sim N_k(\underline{0}, \text{diag}(\underline{p}) - \underline{p}\underline{p}^T)$, $\underline{Z}_0 \sim N_k(\underline{0}, \text{diag}(\underline{p}_0) - \underline{p}_0\underline{p}_0^T)$.

(a) Find the limiting distribution of $\sqrt{n}(\tilde{\underline{p}}_n - \underline{p}_0)$ when $\underline{p} \neq \underline{p}_0$ and when $\underline{p} = \underline{p}_0$. (Express these in terms of \underline{Z} and \underline{Z}_0 .)

(b) Let $\underline{p}_n = \underline{p}_0 + n^{-1/2}\underline{c}$ where $\underline{1}^T \underline{c} = 0$. Find the limiting distribution of $\sqrt{n}(\tilde{\underline{p}}_n - \underline{p}_n)$ under \underline{p}_n .

(c) Is $\tilde{\underline{p}}_n$ a locally regular estimator of \underline{p} at $\underline{p} = \underline{p}_0$? Explain why or why not.

(d) Find the limit of $E_{\underline{p}_n} n \|\tilde{\underline{p}}_n - \underline{p}_n\|^2$ where $\|v\|^2 \equiv \sum_{j=1}^k v_j^2$ for $v \in \mathbb{R}^k$. How does this compare to the limit of $E_{\underline{p}_n} n \|\hat{\underline{p}}_n - \underline{p}_n\|^2$?

6. (48 points).

Consider a parametric model $\mathcal{P} = \{P_\theta : \theta \in \Theta \subset \mathbb{R}^d\}$ satisfying the hypotheses A0-A4 of section 4.1 of the Chapter 4 notes. Suppose that we are using the Rao (or score statistic) $R_n = \underline{Z}_n(\theta_0)^T I(\theta_0)^{-1} \underline{Z}_n(\theta_0)$ for testing the null hypothesis $H : \theta = \theta_0$ versus $K : \theta \neq \theta_0$ where $\underline{Z}_n(\theta_0) \equiv n^{-1/2} \sum_{i=1}^n \dot{l}_\theta(\theta_0 | X_i)$.

(a) Suppose that $\theta_n = \theta_0 + tn^{-1/2}$ and consider $l_n(\theta_n) - l_n(\theta_0) = \sum_{i=1}^n \log\{p_{\theta_n}(X_i)/p_{\theta_0}(X_i)\}$. Use the expansion developed in HW (and stated in part (vi) of Theorem 4.1.2) which implies LAN at θ_0 , together with the linearity of $\underline{Z}_n(\theta_0)$, to find the joint limit distribution of

$$\begin{pmatrix} \underline{c}^T \underline{Z}_n(\theta_0) \\ l_n(\theta_n) - l_n(\theta_0) \end{pmatrix} = \begin{pmatrix} \underline{c}^T \underline{Z}_n(\theta_0) \\ \log \left(\frac{dP_{\theta_n}^n}{dP_{\theta_0}^n} \right) \end{pmatrix}$$

under P_{θ_0} .

(b) Use the result of (a) together with Le Cam's 3rd lemma to find the joint limiting distribution of

$$\begin{pmatrix} \underline{c}^T \underline{Z}_n(\theta_0) \\ l_n(\theta_n) - l_n(\theta_0) \end{pmatrix} = \begin{pmatrix} \underline{c}^T \underline{Z}_n(\theta_0) \\ \log \left(\frac{dP_{\theta_n}^n}{dP_{\theta_0}^n} \right) \end{pmatrix}$$

under P_{θ_n} .

(c) Use the result of (b) to find the limiting distribution of $\underline{Z}_n(\theta_0)$ under P_{θ_n} .

(d) Use the result of (c) to find the limiting distribution of the Rao statistic R_n under P_{θ_n} . What does this imply about the power of the Rao statistic?

7. (50 points).

Suppose that $P = P_0 = N(0, 1)$, $Q = P_\theta = N(\theta, 1)$ on $(\mathbb{X}, \mathcal{A}) = (\mathbb{R}, \mathcal{B})$.

(a) Compute $K(P, Q) = K(P_0, P_\theta)$.

(b) Compute $H^2(P, Q) = 1 - \rho(P, Q)$ and $\rho(P, Q) = \int \sqrt{p(x)q(x)} dx$. [It might be easiest to compute $\rho(P, Q)$ first recalling that if $Z \sim N(0, 1)$ then $E \exp(tZ) = \exp(t^2/2)$.]

(c) Compute $d_{TV}(P, Q) = 1 - \eta(P, Q)$ and $\eta(P, Q) = \int p(x) \wedge q(x) dx$. [It might be easiest to compute $\eta(P, Q)$ first.]

(d) Show in general that $K(P, Q) \geq 2H^2(P, Q)$, thereby strengthening the fact $K(P, Q) \geq 0$ that we proved in class. [Hint: write both $K(P, Q)$ and $H^2(P, Q)$ in terms of $Y = (p(X)/q(X))^{1/2}$ and use the inequality $\log(1+x) \geq x/(1+x)$ for $x \geq 0$. You will need to relate $E_Q Y$ and $E_Q Y^2$ to $H^2(P, Q)$.]

(e) Use the results of (a) and (d) to find a lower bound for $K(P^n, Q^n)$ in terms of $H^2(P, Q)$ or $\rho(P, Q)$; here P^n and Q^n are the probability distributions of X_1, \dots, X_n i.i.d. as P and Q^n respectively.