

MAT3

MATHEMATICAL TRIPOS **Part III**Friday 13 June 2025 1:30 pm to 3:30 pm

PAPER 210**TOPICS IN STATISTICAL THEORY****Before you begin please read these instructions carefully**

Candidates have TWO HOURS to complete the written examination.

Attempt no more than **THREE** questions.There are **FOUR** questions in total.

The questions carry equal weight.

STATIONERY REQUIREMENTSCover sheet
Treasury tag
Script paper
Rough paper**SPECIAL REQUIREMENTS**

None

You may not start to read the questions printed on the subsequent pages until instructed to do so by the Invigilator.
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1 (a) Let X, X_1, \dots, X_n be independent and identically distributed on a measurable space $(\mathcal{X}, \mathcal{A})$ and let \mathcal{G} denote a collection of integrable, real-valued functions on \mathcal{X} . What does it mean for \mathcal{G} to satisfy a *Uniform Law of Large Numbers (ULLN)*? [You may assume that a relevant supremum is measurable.]

Given $\epsilon > 0$, let $N(\epsilon, \mathcal{G}) \in \mathbb{N} \cup \{\infty\}$ denote the minimal $N \in \mathbb{N} \cup \{\infty\}$ for which there exist integrable functions $g_1^L, g_1^U, \dots, g_N^L, g_N^U : \mathcal{X} \rightarrow \mathbb{R}$ with the properties that $\mathbb{E}|g_j^U(X) - g_j^L(X)| \leq \epsilon$ for every $j \in [N]$, and for every $g \in \mathcal{G}$, there exists $j^* \equiv j^*(g) \in [N]$ such that $g_{j^*}^L(x) \leq g(x) \leq g_{j^*}^U(x)$ for all $x \in \mathcal{X}$. Prove that if $N(\epsilon, \mathcal{G}) < \infty$ for every $\epsilon > 0$, then \mathcal{G} satisfies a ULLN.

(b) State the Glivenko–Cantelli theorem, and deduce it from (a).

(c) Now suppose that the distribution function of X is continuous on \mathbb{R} . Let $\mathcal{B}(\mathbb{R})$ denote the Borel σ -algebra on \mathbb{R} and for $A \in \mathcal{B}(\mathbb{R})$, define $g_A : \mathbb{R} \rightarrow \mathbb{R}$ by $g_A(x) := \mathbb{1}_{\{x \in A\}}$. Does $\mathcal{G} := \{g_A : A \in \mathcal{B}(\mathbb{R})\}$ satisfy a ULLN? Justify your answer.

2 In the context of kernel density estimation, define what is meant by a *kernel* K and the *scaled kernel* K_h , where $h > 0$. Given a bounded, Borel measurable function $g_1 : \mathbb{R} \rightarrow \mathbb{R}$ and an integrable function $g_2 : \mathbb{R} \rightarrow \mathbb{R}$, define their *convolution* $g_1 * g_2$.

Let f be a density on \mathbb{R} with $f(x) = 0$ for $x \in (-\infty, 0]$, and suppose that $\mu := \int_0^\infty xf(x) dx < \infty$ and $\bar{\mu} := \int_0^\infty f(x)/x dx < \infty$. Let $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} g$, where g is the density given by $g(x) := xf(x)/\mu$ for $x \in \mathbb{R}$, and suppose that our goal is to estimate f . Suppose further that μ is known, and consider the estimator \hat{f}_n given by

$$\hat{f}_n(x) := \frac{\mu}{nh} \sum_{i=1}^n \frac{1}{X_i} K\left(\frac{x - X_i}{h}\right),$$

where K is a bounded kernel and $h > 0$. Find an exact expression, involving a convolution, for the bias of $\hat{f}_n(x)$. Compare this expression with the bias of a kernel density estimator based on observations having density f .

Show that

$$\int_{-\infty}^{\infty} \text{Var } \hat{f}_n(x) dx = \frac{\mu \bar{\mu} R(K)}{nh} - \frac{1}{n} \int_{-\infty}^{\infty} (K_h * f)^2(x) dx \quad (1)$$

for some $R(K)$ that you should define. Prove further that $\mu \bar{\mu} > 1$ and hence compare (1) with the corresponding integrated variance of a kernel density estimator based on independent observations having density f .

3 For $\beta, L > 0$, define the *Hölder class* $\mathcal{H}(\beta, L)$ on \mathbb{R} .

Consider a vector $Y = (Y_1, \dots, Y_n)^\top$ of responses generated via

$$Y_i = m(x_i) + \epsilon_i,$$

where $x_i = i/n$ for $i \in [n]$, where $m \in \mathcal{H}(\beta, L)$ and where $\epsilon_1, \dots, \epsilon_n$ are independent with $\mathbb{E}(\epsilon_i) = 0$ and $\text{Var}(\epsilon_i) \leq \sigma^2$ for $i \in [n]$. Fix $x \in [0, 1]$, define $K : \mathbb{R} \rightarrow [0, 1/2]$ by $K(u) := \mathbb{1}_{\{|u| \leq 1\}}/2$, let $p \in \mathbb{N}_0$ and let $h > 0$. Give the definition of the *local polynomial estimator* $\hat{m}_n(x; p, h)$ of $m(x)$ of degree p , bandwidth h and kernel K . Show that, for suitable matrices $X \in \mathbb{R}^{n \times (p+1)}$ and $W \in \mathbb{R}^{n \times n}$, and subject to a positive definiteness condition that you should state, $\hat{m}_n(x; p, h)$ can be expressed as an appropriate component of $\hat{\beta} = (\hat{\beta}_0, \hat{\beta}_1, \dots, \hat{\beta}_p)^\top$, where $\hat{\beta} := (X^\top W X)^{-1} X^\top W Y$.

Define the *effective kernel* $\{w_i(x) : i \in [n]\}$ of $\hat{m}_n(x; p, h)$. Under the same positive definiteness condition as above, prove that if R is a polynomial of degree at most p , then

$$\frac{1}{n} \sum_{i=1}^n w_i(x) R(x_i) = R(x).$$

Writing $Q(u) := (1, u, u^2/2, \dots, u^p/p!)^\top \in \mathbb{R}^{p+1}$ for $u \in \mathbb{R}$, explain why the minimum eigenvalue Λ_p of the matrix $\int_0^1 Q(u) Q(u)^\top du \in \mathbb{R}^{(p+1) \times (p+1)}$ is positive. Using the fact that the minimum eigenvalue $\lambda_0 \equiv \lambda_{0,n,h,x}(p)$ of $n^{-1} X^\top W X$ satisfies

$$\inf_{x \in [0,1]} \lambda_{0,n,h,x}(p) \geq \frac{1}{2} \left(\Lambda_p - \frac{6e}{nh} \right),$$

when $n \geq 2$ and $h \leq 1/4$, prove that for $p \geq \lceil \beta \rceil - 1$ there exist $n_0 \equiv n_0(p) \in \mathbb{N}$, $a \equiv a(p) > 0$ and $C \equiv C(p) > 0$ such that

$$\sup_{x \in [0,1]} \sup_{m \in \mathcal{H}(\beta, L)} \mathbb{E}[\{\hat{m}_n(x; p, h) - m(x)\}^2] \leq C \left(\frac{\sigma^2}{nh} + L^2 h^{2\beta} \right)$$

for $n \geq n_0$ and $h \in [a/n, 1/4]$.

4 Let P and Q be probability measures on a σ -finite measure space $(\mathcal{X}, \mathcal{A}, \mu)$, having Radon–Nikodym derivatives p and q respectively with respect to μ . Write down expressions for the *total variation distance* $\text{TV}(P, Q)$ and *Hellinger distance* $H(P, Q)$ as integrals with respect to μ . Prove that

$$\text{TV}^2(P, Q) \leq H^2(P, Q) \leq 2\text{TV}(P, Q).$$

Writing P^n and Q^n for the n -fold product distributions of P and Q respectively, and using the fact that these probability measures have Radon–Nikodym derivatives p^n and q^n respectively with respect to the σ -finite product measure μ^n , prove that

$$H^2(P^n, Q^n) = 2 - 2\left(1 - \frac{1}{2}H^2(P, Q)\right)^n.$$

State Le Cam’s two-point lemma.

Let $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} U[\theta, \theta + 1]$ for some $\theta \in \mathbb{R}$. Prove that there exists a universal constant $c > 0$ such that

$$\inf_{\hat{\theta} \in \hat{\Theta}} \sup_{\theta \in \mathbb{R}} \mathbb{E}_{\theta} \{ (\hat{\theta}(X_1, \dots, X_n) - \theta)^2 \} \geq \frac{c}{n^2}$$

for all $n \in \mathbb{N}$, where $\hat{\Theta}$ denotes the set of Borel measurable functions from \mathbb{R}^n to \mathbb{R} . [You may use the fact that $(1 - x)^a \geq 1 - ax$ for $x \in [0, 1]$ and $a \geq 1$.]

END OF PAPER