

THE SET-COMPOUND ONE-STAGE ESTIMATION IN THE NONREGULAR* FAMILY OF DISTRIBUTIONS OVER THE INTERVAL $[\theta, \theta + 1)$

YOSHIKO NOGAMI

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Introduction

This paper is a continuation of author's Ph.D. thesis [6] and Nogami [7].

The set-compound problem simultaneously considers n statistical decision problems each of which is structurally identical to the component problem. The loss is taken to be the average of n component losses.

Let ξ be Lebesgue measure and f a measurable function with $0 \leq f \leq 1$. Define

$$(0.1) \quad q(\theta) = 1 / \int_{\theta}^{\theta+1} f d\xi.$$

Let $\mathcal{P}^*(f)$ be a family of probability measures determined by

$$(0.2) \quad \mathcal{P}^*(f) = \{P_{\theta} \text{ with } p_{\theta} = q(\theta)[\theta, \theta + 1)f, \forall \theta \in \Omega\}$$

where $p_{\theta} = dP_{\theta}/d\xi$ and Ω is a real interval $[c, d]$ with $-\infty < c < d < +\infty$, and we denote the indicator function of a set A by A itself. The component problem is the squared-error loss estimation (SELE) of θ based on $X \sim P_{\theta} \in \mathcal{P}^*(f)$.

Let X_1, \dots, X_n be n independent random variables with each $X_j \sim P_{\theta_j} \in \mathcal{P}^*(f)$. The modified regret of the set-compound decision procedure $\mathbf{t} = (t_1, \dots, t_n)$ is of form

$$(0.3) \quad D(\boldsymbol{\theta}, \mathbf{t}) = E \left(n^{-1} \sum_{j=1}^n (\theta_j - t_j(\mathbf{X}))^2 \right) - R(G_n)$$

where $R(G_n)$ is the Bayes risk against the empiric distribution G_n of $\theta_1, \dots, \theta_n$ in the component problem.

With squared-error loss, let $\boldsymbol{\theta}_{G_n}$ be the procedure whose component procedures are Bayes against G_n : $\boldsymbol{\theta}_{G_n} = (\theta_{1n}, \dots, \theta_{nn})$ with, for each j ,

* The word "nonregular" was quoted from Ferguson ([3], p. 130).

$$(0.4) \quad \theta_{jn} = \int_{x'_j+}^{x_j} \theta q(\theta) dG_n(\theta) / \int_{x'_j+}^{x_j} q dG_n$$

where y' is an abbreviation of $y-1$ and the affix $+$ is intended to describe the integration as over $(X'_j, X_j]$. Henceforth we delete $+$ in lower limits of \int s. Then, we can write

$$R(G_n) = E \left(n^{-1} \sum_{j=1}^n (\theta_j - \theta_{jn})^2 \right).$$

The work here is a generalization and an extension of Fox's work [4], respective to a family $\mathcal{P}^*(f)$ and to the set-compound SELE problem. When P_θ is the uniform distribution on $[\theta, \theta+1)$ for $\theta \in (-\infty, \infty)$ and the θ are i.i.d. with a prior G , Fox [4] showed the convergence to $R(G)$ of the respective expected risks for a one-stage procedure with components direct estimates of the posterior means wrt G .

Nogami ([6], Chapters II and III) introduced a one-stage set-compound estimate θ_τ for $\theta \in [c, d]^n$ (we say this fact so that θ_τ has a rate $1/4$) under $\mathcal{P}^*(f)$ with Lipshitz condition for $1/f$. In this paper we demonstrate (in Section 1) another estimate ϕ^* with a rate $1/3$ without Lipshitz condition for $1/f$ and can expect (from Section 2) that both θ_τ and ϕ^* have the same best exact order $n^{-2/3}$ of convergence of the modified regret. In Section 1 we get an upper bound for $D(\theta, \phi^*)$. In Nogami [8], there is a misprint in the bound of Theorem in Section 2. The bound there should be $(8N+24)m^k \{N^k \cdot (4k-2+k^{1/2})((n-k+1)h^k)^{-1/2} + 2^{-k+1}h^k\}$. Although this bound with $k=1$ gives an upper bound for $D(\theta, \phi^*)$, the result of Section 1 in this paper is stronger than that. Section 2 gives us lower bounds for $D(0, \phi^*)$ at $f \equiv 1$.

Notations. We often let $P(h)$ or $P(h(\omega))$ denote $\int h(\omega) dP(\omega)$. G abbreviates the empiric distribution G_n of $\theta_1, \dots, \theta_n$. For any function h , $h|_a^b$ means $h(b)-h(a)$. When we refer to (a.b) in Section a we simply write (b). \vee and \wedge denote the supremum and the infimum, respectively. \doteq denotes the defining property. P_x means the conditional expectation of $X_1, \dots, X_{j-1}, X_{j+1}, \dots, X_n$ given $X_j \doteq x$. A distribution function also represents the corresponding measure. Define $\bar{z} = n^{-1} \sum_{i=1}^n z_i$.

1. An upper bound for the modified regret $D(\theta, \phi^*)$

In this section we shall get a one-stage procedure ϕ^* for estimating θ and show that it has a rate $1/3$ under the assumption $f(\cdot) \geq m^{-1}$.

Assume

$$(1.1) \quad f(\cdot) \geq m^{-1} (>0), \quad \text{for a constant } m (>1).$$

Before mentioning about the structure of the procedure ϕ^* we introduce the following:

LEMMA 1.1 (Nogami [6]). *Let τ be a signed measure, g be a measurable function and $I=(y', y]$ be an interval with $\int Igd\tau \neq 0$. Let τ_y be the signed measure with density $Ig/\int Igd\tau$ wrt τ . Then,*

$$\int sd\tau_y(s) = y - \int_0^1 \tau_y(y', y' + t) dt.$$

PROOF. By Fubini's theorem applied to the lhs of the second equality below,

$$y - \int sd\tau_y(s) = \int \int_{s=y'}^1 dt d\tau_y(s) = \int_0^1 \tau_y(y', y' + t) dt.$$

For fixed j , $1 \leq j \leq n$, we abbreviate X_j to x . Fix j until (5). Let Q be the measure with the density q wrt G . Define

$$(1.2) \quad u_i(y) = p_i(y)/f(y), \quad i=1, \dots, n.$$

Then, by the definition of p_i

$$(1.3) \quad \bar{u}(y) = Q|_{y'}^y.$$

Thus,

$$(1.4) \quad Q(y) = \sum_{r=0}^{\infty} \bar{u}(y-r).$$

By above Lemma 1.1 applied to (0.4)

$$(1.5) \quad \begin{aligned} \theta_{jn} &= x - \int_0^1 Q|_{x'}^{x'+t} dt / Q|_x^x \\ &= x - \int_0^1 \sum_{r=1}^{\infty} \bar{u}|_{x-r}^{x-r+t} dt / \bar{u}(x). \end{aligned}$$

In view of (2) we estimate $\bar{u}(y)$ by $\hat{\bar{u}}(y) = n^{-1} \sum_{i=1}^n \hat{u}_i$ where for any $h > 0$

$$(1.6) \quad \hat{u}_i(y) = h^{-1} [y \leq X_i < y+h] / f(X_i).$$

We allow h to depend on n and assume $h < 1$ for convenience. Thus, this and (5) (observe $x' < \theta_{jn} \leq x$) suggest that to achieve a small modified regret we might estimate θ_{jn} by

$$(1.7) \quad \phi_{jn}^* = x - 0 \vee \left(\int_0^1 \sum_{r=1}^{\infty} \hat{\bar{u}}|_{x-r}^{x-r+t} dt / \hat{\bar{u}}(x) \right) \wedge 1$$

and thus $\phi^*=(\phi_{1n}^*, \dots, \phi_{nn}^*)$ is an estimate of $\theta_\sigma(X)$ and thus of θ . Note that if the r -th term of the numerator of the quotient of rhs (7) is nonzero, then

$$(1.8) \quad r \leq d+3-c \doteq N+2.$$

Since $X'_j < \phi_{jn}^*$, $\theta_{jn} \leq X_j$, $j=1, 2, \dots, n$,

$$(1.9) \quad n2^{-1}|D(\theta, \phi^*)| \leq \sum_{j=1}^n P_j P_x |\phi_{jn}^* - \theta_{jn}|.$$

We shall invoke the following corollary, a special case of Lemma A.2 of Singh [9], to get a bound of $P_x |\phi_{jn}^* - \theta_{jn}|$.

COROLLARY 1.1 (Singh [9]). *For real random variables Y and Z , and real numbers y and z ,*

$$(1.10) \quad E \left(\left| \frac{Y}{Z} - \frac{y}{z} \right| \wedge 1 \right) \leq 2|z|^{-1} \left\{ E|Y-y| + \left(\left| \frac{y}{z} \right| + 1 \right) E|Z-z| \right\}.$$

Applying Corollary 1.1 and weakening the resulted bound shows that for fixed j ,

$$(1.11) \quad P_x |\phi_{jn}^* - \theta_{jn}| \leq (\bar{u}(x))^{-1} \left[\sum_{r=1}^{N+2} \left\{ \int_0^1 P_x |\bar{u}(x-r+t) - \bar{\bar{u}}(x-r+t)| dt \right. \right. \\ \left. \left. + P_x |\bar{u}(x-r) - \bar{\bar{u}}(x-r)| \right\} + 2P_x |\bar{u}(x) - \bar{\bar{u}}(x)| \right].$$

But with $\bar{u}_j \doteq (n-1)^{-1} \sum_{(j \neq i)}^n u_i$ and $\bar{\bar{u}}_j \doteq (n-1)^{-1} \sum_{(j \neq i)}^n \hat{u}_i$,

$$(1.12) \quad nP_x |\bar{u}(x-r+t) - \bar{\bar{u}}(x-r+t)| - (n-1)P_x |\bar{u}_j(x-r+t) - \bar{\bar{u}}_j(x-r+t)| \\ \leq |u_j(x-r+t) - (hf(x))^{-1}| \leq (2m)/h$$

where the last inequality follows by

$$(1.13) \quad u_j(\cdot) \leq m, \quad \forall j \quad \text{and} \quad 1/f(\cdot) \leq m.$$

Lemma 2.2 below will be used to get a bound of $\sum_{j=1}^n P_j(\text{rhs}(11))$ and is proved in the proof of Lemma 2.1 of Nogami [7] with β there replaced by N .

LEMMA 2.2.

$$(1.14) \quad \sum_{j=1}^n P_j(\bar{u}(X_j))^{-1} \leq nN.$$

By three applications of (12) and an application of (14), and by weakening the resulted bound we obtain

$$\begin{aligned}
 (1.15) \quad & n(2N+6)^{-1} \sum_{j=1}^n P_j(\text{rhs (11)}) \\
 & \leq (n-1) \bigvee_{y \geq 0} \sum_{j=1}^n P_j \{ (\bar{u}(X_j))^{-1} \mathbf{P}_x |\bar{u}_j(X_j - y) - \bar{\hat{u}}_j(X_j - y)| \} \\
 & \quad + 2mNnh^{-1}.
 \end{aligned}$$

Since by the triangular inequality and Hölder's inequality,

$$(1.16) \quad \mathbf{P}_x |\bar{u}_j(y) - \bar{\hat{u}}_j(y)| \leq |\bar{u}_j(y) - \mathbf{P}_x \bar{\hat{u}}_j(y)| + \sigma_n(y)$$

where $\sigma_n^2(y)$ = variance of $\bar{\hat{u}}_j(y)$, to get an upper bound of the first term of rhs (15) we shall obtain bounds for $\bigvee_y \sigma_n(y)$ and $\bigvee_{y \geq 0} \sum_{j=1}^n P_j$ ((first term of rhs (16) at $X_j - y$)/ $\bar{u}(X_j)$).

LEMMA 2.3.

$$\bigvee_y \sigma_n(y) \leq m((n-1)h)^{-1/2}.$$

PROOF. By the definition of σ_n^2

$$(1.17) \quad ((n-1)h)^2 \sigma_n^2(y) \leq \sum_{(j \neq i)}^n \mathbf{P}_x (\hat{u}_i(y))^2 \leq (n-1) \int_y^{y+h} \bar{u}_j(z) f(z) dz$$

which is no more than $m^2(n-1)h$ because of (13).

LEMMA 2.4. For all $y \geq 0$,

$$(1.18) \quad \sum_{j=1}^n P_j (|\bar{u}_j(X_j - y) - \mathbf{P}_x \bar{u}_j(X_j - y)| / \bar{u}(X_j)) \leq nhm.$$

PROOF. Since $\mathbf{P}_x \bar{\hat{u}}_j(z) = h^{-1} \int_z^{z+h} \bar{u}_j(t) dt = \int_0^1 \bar{u}_j(z + hs) ds$, lhs (18) for every $y \geq 0$ equals to

$$\sum_{j=1}^n \int \left| \int_0^1 \bar{u}_j \Big|_{z-y}^{z-y+hs} ds \right| \cdot p_j(z) / \bar{u}(z) dz$$

which is no more than

$$\begin{aligned}
 & \sum_{j=1}^n \int_c^{d+1} \int_0^1 (n-1)^{-1} \sum_{(j \neq i)}^n q(\theta_i) ([\theta_i - hs \leq z - y < \theta_i] + [\theta_i - hs \leq z - y - 1 < \theta_i]) ds \\
 & \quad \cdot p_j(z) / \bar{u}(z) dz.
 \end{aligned}$$

Thus, interchanging integrations and also averages over respective j and i leads to

$$\begin{aligned}
 (1.19) \quad & \text{lhs (18)} \leq \sum_{i=1}^n q(\theta_i) \int_0^1 \int_c^{d+1} ([\theta_i - hs \leq z - y < \theta_i] + [\theta_i - hs \leq z - y - 1 < \theta_i]) \\
 & \quad \cdot (n-1)^{-1} \sum_{(i \neq j)}^n p_j(z) / \bar{u}(z) dz ds.
 \end{aligned}$$

Since $(n-1)^{-1} \sum_{(i^*)_{j=1}^n} p_j(z)/\bar{u}(z) \leq f(z) \leq 1$, by a simple computation and $q(\cdot) \leq m$

$$\text{rhs (19)} \leq 2h \sum_{i=1}^n q(\theta_i) \int_0^1 s ds = h \sum_{i=1}^n q(\theta_i) \leq nhm.$$

We now go back to (15). By Lemmas 2.4 and 2.3 together with an application of Lemma 2.2 we obtain in view of (16) that

$$\text{rhs (15)} \leq mNn^{3/2}h^{-1/2} + (n-1)nhm + 2mNnh^{-1}.$$

Therefore, in view of (9), we finally obtain

THEOREM 1.1. *For all $\theta \in [c, d]^n$,*

$$|D(\theta, \phi^*)| \leq 2(2N+6)\{3mN(nh)^{-1/2} + mh\}.$$

Remark. In Chapter III of Nogami [6], two one-stage procedures; one (denoted by θ_τ) for $\mathcal{P}^*(f)$ under Lipschitz condition for $1/f$ and the other (denoted by ϕ) for $\mathcal{P}^*(1)$, both with a rate $1/4$ are exhibited as a special case ($k=1$) of the k -extended problem. (From the structure of construction ϕ cannot be extended to $\mathcal{P}^*(f)$.) In Chapter II of Nogami [6] Theorem 3 (Theorem 2.1 in Section 2 of this paper) shows that when $\theta=0$ and $f \equiv 1$ (note that in this case θ_τ and ϕ^* are the same estimate for the zero sequence 0), θ_τ with $h^{-1}n^{-1/4} = O(1)$ has exact order $O(h^2)$ of convergence, and Theorems 4 and 5 there give a lower bound and an upper bound for $D(0, \phi)$ at $f \equiv 1$, respectively. In this section we assume no Lipschitz condition for $1/f$ and from above Theorem 1.1 we can see that for ϕ^* with a choice of $h = n^{-1/3}$ (up to constants) $|D(\theta, \phi^*)| = O(n^{-1/3})$, uniformly in $\theta \in [c, d]^n$. Furthermore, from Theorem 2.1 in the next section we shall see that for this choice of h ϕ^* has the best lower bound $n^{-2/3}$ for $D(0, \phi^*)$ at $f \equiv 1$ and this shows that $D(\theta, \phi^*)$ converges to zero at a rate no faster than $n^{-2/3}$.

2. Lower bounds of the modified regret $D(0, \phi^*)$ when $f \equiv 1$

In this section we consider the uniform distribution $P = U[0, 1)$ over the interval $[0, 1)$ as the underlined family of distributions. Lemmas 2.1 through 2.4 will be furnished to prove forthcoming Theorem 2.1 which gives us lower bounds of $D(0, \phi^*)$. Theorem 2.2 is a derivation from Theorem 2.1 and somewhat Section 1 and will be stated without proof.

Let X_1, \dots, X_n be i.i.d. random variables with the common distribution $P = U[0, 1)$. Let $X = (X_1, \dots, X_{n+1})$. Here we consider $\phi^*(X) = (\phi_{1,n+1}^*, \dots, \phi_{n+1,n+1}^*)$. Since $\phi_{1,n+1}^*, \dots, \phi_{n+1,n+1}^*$ are identically distributed and since for all j , $\theta_{j,n+1} = 0$, abbreviating $\phi_{n+1,n+1}^*$ to ϕ^* we see in view

of (0.3) that the modified regret of ϕ^* at $\theta=0$ is given by

$$(2.1) \quad D(0, \phi^*) = P\phi^{*2}$$

where $P = P_1 \times \cdots \times P_{n+1}$.

For fixed $x \in X_{n+1}$, ϕ^* is written as

$$(2.2) \quad \phi^* = (x' \vee \varphi) \wedge x$$

where with $\bar{u}(y) = (n+1)^{-1} \sum_{i=1}^{n+1} \dot{u}_i(y)$,

$$(2.3) \quad \varphi = x - \int_0^1 \sum_{r=0}^{\infty} \bar{u}(\cdot - r)]_{x'+t}^{x'+t} / \bar{u}(x) dt.$$

We shall exhibit an explicit form of φ in a.e. P_x -sense in the following:

LEMMA 2.1. For every $x \in [0, 1]$,

$$(2.4) \quad \varphi = \left\{ \sum_{j=1}^n (X_j - h) [x < X_j \leq x + h] - h \sum_{j=1}^n [0 \leq X_j \leq x] - h \right. \\ \left. + \sum_{j=1}^n [0 \leq X_j \leq x' + h] \right\} / \sum_{j=1}^n [x < X_j \leq x + h] \quad \text{a.e. } P_x.$$

PROOF. Fix j and note that as a function of $t \in [0, 1]$, $\sum_{r=0}^{\infty} [X_j - x' + r - h \leq t < X_j - x' + r]$ is equal to zero, is equal to its first term, or is equal to the sum of its first two terms according to whether $1 < X_j - x' - h$, $X_j - x' - h \leq 1 < X_j - x'$ or $X_j - x' \leq 1$. Integrating over $t \in [0, 1]$ for each case gives

$$\int_0^1 \sum_{r=0}^{\infty} [X_j - x' + r - h \leq t < X_j - x' + r] dt \\ = (x + h - X_j) [x < X_j \leq x + h] + h [X_j \leq x].$$

Hence, it follows

$$(2.5) \quad ((n+1)h) \int_0^1 \sum_{r=0}^{\infty} \bar{u}(\cdot - r)]_{x'+t}^{x'+t} dt \\ = \sum_{j=1}^{n+1} (x + h - X_j) [x < X_j \leq x + h] + h \sum_{j=1}^{n+1} [X_j \leq x] \\ - \sum_{j=1}^{n+1} \sum_{r=0}^{\infty} [x' - r < X_j \leq x' - r + h].$$

But since $[x < x \leq x + h] = 0$, $[x \leq x] = 1$, $\sum_{r=0}^{\infty} [x' - r < x \leq x' - r + h] = 0$ and

a.e. P_x , $\sum_{r=1}^{\infty} [x' - r < X_j \leq x' - r + h] = 0$, we have

$$\text{rhs (5)} = x \sum_{j=1}^n [x < X_j \leq x + h] - \sum_{j=1}^n (X_j - h) [x < X_j \leq x + h]$$

$$+h \sum_{j=1}^n [X_j \leq x] + h - \sum_{j=1}^n [x' < X_j \leq x' + h], \quad \text{a.e. } \mathbf{P}_x.$$

On the other hand, since $[x < x \leq x + h] = 0$,

$$((n+1)h)\bar{u}(x) = \sum_{j=1}^n [x < X_j \leq x + h].$$

Applying these to the definition of φ , we get the asserted expression for φ .

In this section we need only to deal with φ for $x < 1 - h$, where the term $\sum_{j=1}^n [0 \leq X_j \leq x' + h]$ (cf. rhs (4)) vanishes. We also recognize that for $x < 1 - h$, $\mathbf{P}_x[\varphi > x] = 0$. Hence, ϕ^* has the following simpler form:

$$(2.6) \quad \phi^* = \begin{cases} x' \vee \varphi & \text{for } x \in [0, 1 - h) \\ (x' \vee \varphi) \wedge x, & \text{for } x \in [1 - h, 1). \end{cases}$$

Now, we let

$$(2.7) \quad J = [\varphi \geq x', \ x < 1 - h]$$

and recognize by (1) and the definition of ϕ^* that

$$(2.8) \quad D(0, \phi^*) \geq P(\varphi^2 J).$$

Let $\xrightarrow{\mathcal{D}}$ denote convergence in distribution. Also, $N(c, d)$ denotes the normal distribution with mean c and variance d . To get lower bounds for $D(0, \phi^*)$ (Theorem 2.1) we use the relation (8) and the fact that for fixed x , $h^{-1}\varphi J \xrightarrow{\mathcal{D}} -2^{-1}$ and $S_n \doteq (\sqrt{nh}\varphi + 2^{-1}\sqrt{nh^3})J \xrightarrow{\mathcal{D}} N(0, x^2)$. We then apply a convergence theorem (cf. Loève [5] 11.4, A(i)):

$$(2.9) \quad \text{If } U_n \xrightarrow{\mathcal{D}} U, \text{ then } \underline{\lim} E U_n^2 \geq E U^2,$$

where E means expectation, and Theorem A in Appendix. We shall first prepare Lemmas 2.2, 2.3 and 2.4 to prove the above two convergences in distribution for the proof of forthcoming Theorem 2.1.

Let $u = \sum_{j=1}^n [0 \leq X_j \leq x]$, $v = \sum_{j=1}^n [x < X_j \leq x + h]$ and $w = \sum_{j=1}^n X_j [x < X_j \leq x + h]$. We also define

$$X = (w - hv - xv - h)/(hv),$$

$$Y = (u - nx)/\sqrt{nx(1-x)} \quad \text{and}$$

$$Z = (v - nh)/\sqrt{nh}.$$

Then, on the set J , φ of the form (4) is alternatively written as

$$(2.10) \quad \varphi = hX + \frac{x(nh)^{-1/2}Z}{1+(nh)^{-1/2}Z} - \frac{\sqrt{x(1-x)}n^{-1/2}Y}{1+(nh)^{-1/2}Z}.$$

LEMMA 2.2. Given $x \in (0, 1)$, if h is a function of n such that $nh \rightarrow \infty$ and $h \rightarrow 0$, then

$$(Y, Z) \xrightarrow{\mathcal{D}} N(\underline{0}, I)$$

where $\underline{0}$ is the zero vector in R^2 and I is 2×2 identity matrix.

PROOF. For each $x \in (0, 1)$ we restrict to n such that $x < 1-h$. Pick t and s arbitrary, and let

$$V_j = n^{-1/2} \{s(x(1-x))^{-1/2}([0 \leq X_j \leq x] - x) + th^{-1/2}([x < X_j \leq x+h] - h)\},$$

for $j=1, 2, \dots, n$. Then, it is not hard to see that

$$\sum_{j=1}^n V_j = sY + tZ.$$

Since the V_j are i.i.d., the characteristic function K of (Y, Z) at a point $(s, t) \in R^2$ is given by

$$(2.11) \quad K(s, t) = (J(1))^n$$

where J is the characteristic function of $V \doteq V_1$.

Since by XV (6.8) (Feller, [2]), for any complex numbers such that $|\alpha| \leq 1$ and $|\beta| \leq 1$,

$$|\alpha^n - \beta^n| \leq n|\alpha - \beta|,$$

$$(2.12) \quad \left| (J(1))^n - \exp\left(-\frac{1}{2}(s^2 + t^2)\right) \right| \leq n \left| J(1) - \exp\left(-\frac{1}{2n}(s^2 + t^2)\right) \right|.$$

By the triangular inequality and by using $|1 - y - e^{-y}| = O(y^2)$ as $y \rightarrow 0$,

$$(2.13) \quad \text{rhs (12)} \leq n \left| J(1) - 1 + \frac{s^2 + t^2}{2n} \right| + O(n^{-1}).$$

Now, from the Taylor development of characteristic functions by XV (4.14) (Feller, [2]) and from the fact that $J(0)=1$, $J'(0)=i\mathbf{P}_x V=0$ and $J''(0)=-\mathbf{P}_x V^2$, it follows that

$$\left| J(1) - 1 + \frac{1}{2}\mathbf{P}_x V^2 \right| \leq \frac{1}{6}\mathbf{P}_x |V|^3.$$

Now, we verify that

$$\mathbf{P}_x V^2 = n^{-1} \{ (s^2 + t^2) - t^2 h - 2stx(\sqrt{x}(1-x)^{-1/2} + \sqrt{1-x}x^{-1/2})\sqrt{h} \}$$

and

$$\begin{aligned} P_x|V|^3 = & n^{-3/2} \{ |s(x^{-1}-1)^{1/2} - th^{1/2}|^3 x + |t(1-h)h^{-1/2} - sx^{1/2}(1-x)^{-1/2}|^3 h \\ & + |sx^{1/2}(1-x)^{-1/2} + th^{1/2}|^3 (1-x-h) \}. \end{aligned}$$

Hence,

$$0 \leq n^{-1}(s^2 + t^2) - P_x V^2 \leq O(n^{-1}h^{1/2})$$

and

$$P_x|V|^3 = O(n^{-3/2}h^{-1/2}).$$

Hence, applying the triangular inequality leads to

$$\left| J(1) - 1 + \frac{s^2 + t^2}{2n} \right| = O(n^{-1}h^{1/2} + n^{-3/2}h^{-1/2}).$$

Thus, in view of (13), (12) and (11),

$$\left| K(t, s) - \exp\left(-\frac{s^2 + t^2}{2}\right) \right| = O(h^{1/2} + n^{-1/2}h^{-1/2} + n^{-1}).$$

To get the conclusion we invoke the continuity theorem (cf. e.g. Breiman [1], Theorem 11.6).

We shall next prove $X \xrightarrow{P} -2^{-1}$ where $\xrightarrow{P} 0$ means convergence in probability P_x for given x .

LEMMA 2.3. *Under the same assumption as Lemma 2.2,*

$$X \xrightarrow{P} -2^{-1}.$$

PROOF. For given $x \in (0, 1)$, we restrict to n such that $x < 1-h$. Then, X is written as

$$(2.14) \quad X = \left(C / \left(\frac{v}{nh} \right) \right) - v^{-1}$$

where $C = (nh)^{-1} \sum_{j=1}^n U_j$, where $U_j = h^{-1}(X_j - x - h)I_j$ with $I_j = [x < X_j \leq x+h]$.

Since v has the binomial distribution with parameters n and h ,

$$(2.15) \quad \frac{v}{nh} \xrightarrow{P} 1 \text{ as } nh \rightarrow \infty \text{ and } h \rightarrow 0.$$

By simple computations,

$$E U = -\frac{h}{2}$$

and

$$\text{Var}(U) = \frac{h}{12} + \frac{h(1-h)}{4}.$$

Thus, $EC = h^{-1}EU = -2^{-1}$ and $\text{Var}(C) = (nh^2)^{-1} \text{Var}(U) = (12^{-1} + (1-h)/4)/(nh)$. Therefore, by the Chebychev inequality,

$$(2.16) \quad C \xrightarrow{P} -2^{-1} \text{ as } nh \rightarrow \infty \text{ and } h \rightarrow 0.$$

Applying (16), (15), (14) and Slutsky's theorem completes the proof of Lemma 2.3.

Besides the above two lemmas we shall show that $P_x[\varphi \leq x']$ vanishes when $nh \rightarrow \infty$ and $h \rightarrow 0$.

LEMMA 2.4. *Under the same assumption as Lemma 2.2,*

$$P_x[\varphi \leq x'] \rightarrow 0 \quad \text{for fixed } x.$$

PROOF. We restrict to n such that $x < 1-h$. Let $W_j = h[0 \leq X_j \leq x] - (X_j - h - x')[x < X_j \leq x+h]$ for $j=1, 2, \dots, n$. Then, by the representation (4) of φ , $[\varphi \leq x'] = [\bar{W} \geq -n^{-1}h]$ where \bar{W} is the average of i.i.d. W_j 's. Since $P_x W_1 = h(2^{-1}h + x')$,

$$(2.17) \quad P_x[\varphi \leq x'] = P_x[\bar{W} - P_x \bar{W} \geq (1-x-n^{-1}-2^{-1}h)h].$$

But, $\text{Var}(\bar{W}) = n^{-1} \text{Var}(W_1) = hn^{-1} \left\{ 1 - (1-x)(2-x)h + \left(\frac{4}{3} - x\right)h^2 - 4^{-1}h^3 \right\} \leq \left(\frac{7}{3}\right)hn^{-1}$. Hence, by the Chebychev inequality and for large n

$$\text{rhs}(17) \leq (7/3)h^{-1}n^{-1}(1-x-n^{-1}-2^{-1}h)^{-2}$$

which tends to zero when $nh \rightarrow \infty$ and $h \rightarrow 0$.

We are now ready to prove

THEOREM 2.1. (i) *If h is a function of n such that $nh^3 \rightarrow \infty$ and $h \rightarrow 0$, then for any $\frac{1}{4} > \varepsilon > 0$, there exists $N < +\infty$ so that for all $n \geq N$*

$$D(0, \phi^*) > \left(\frac{1}{4} - \varepsilon\right)h^2.$$

(ii) *If h is a function of n such that $nh \rightarrow \infty$, $h \rightarrow 0$ and $nh^3 = O(1)$, then for any $\frac{1}{3} > \varepsilon > 0$, there exists $N < +\infty$ so that for all $n \geq N$*

$$D(0, \phi^*) > \left(\frac{1}{3} - \varepsilon\right)\frac{1}{nh}.$$

PROOF. (i) Since $nh^3 \rightarrow \infty$ and $h \rightarrow 0$ implies $nh \rightarrow \infty$ and $h \rightarrow 0$, we have by Lemmas 2.2, 2.3 and 2.4 that given $x \in (0, 1)$,

$$(2.18) \quad (Y, Z) \xrightarrow{\mathcal{D}} N(0, I), \quad X \xrightarrow{P} -\frac{1}{2}, \quad [\varphi \geq x'] \xrightarrow{P} 1.$$

Hence, in view of (10) it follows from Slutsky's theorem that if $x \in (0, 1)$, then $h^{-1}\varphi J \xrightarrow{\mathcal{D}} -2^{-1}$ (see (7) for the definition of J). By a convergence theorem (9), we have

$$(2.19) \quad \underline{\lim} P_x(h^{-2}\varphi^2 J) \geq \frac{1}{4} [0 < x < 1],$$

and hence by Fatou's theorem applied to the lhs below

$$\underline{\lim} P P_x(h^{-2}\varphi^2 J) \geq P(\text{lhs (19)}) \geq \frac{1}{4}.$$

Thus, by (8) we get that

$$\underline{\lim} h^{-2}D(0, \phi^*) \geq \frac{1}{4}.$$

(i) follows because of the definition of \liminf .

To prove (ii) we first recognize that for this choice of h , (18) still holds. Let $S_n = \{\sqrt{nh}\varphi + 2^{-1}\sqrt{nh^3}\}J$. Then, in view of (10) it follows from Slutsky's theorem that if $x \in (0, 1)$, then

$$S_n \xrightarrow{\mathcal{D}} N(0, x^2).$$

Since $P_x\{(nh)\varphi^2 J\} = P_x(S_n - 2^{-1}\sqrt{nh^3}J)^2 \geq \text{Var}(S_n)$, applying Theorem A in Appendix to the rhs leads to

$$(2.20) \quad \underline{\lim} P_x\{(nh)\varphi^2 J\} \geq x^2 [0 < x < 1].$$

Thus, by Fatou's lemma applied to the lhs below

$$\begin{aligned} \underline{\lim} P P_x(nh\varphi^2 J) &\geq P(\text{lhs (20)}) \\ &\geq \int_0^1 y^2 dy = \frac{1}{3}. \end{aligned}$$

Therefore by (8) we get that $\underline{\lim} (nh)D(0, \phi) \geq \frac{1}{3}$ and the definition of \liminf leads to (ii).

Theorem 2.1 (i) implies that at any parameter sequence $(\theta_1, \theta_2, \dots)$ where $\theta_1 = \theta_2 = \dots$, ϕ^* with the choice $h = n^{-1/3}$ has modified regret converging to zero at a rate no faster than $n^{-2/3}$.

Remark. By usage of the method obtaining Theorem 1.1 and the result of Theorem 2.1 we can verify the following:

THEOREM 2.2. (i) *If h is a function of n such that $nh^3 \rightarrow \infty$ and $h \rightarrow 0$, then there exists a positive constant b_1 such that for sufficiently large n ,*

$$b_1^{-1}h^2 \leq D(0, \phi^*) \leq b_1h^2.$$

(ii) *If h is a function of n such that $nh \rightarrow \infty$, $h \rightarrow 0$ and $nh^3 = O(1)$, then there exists a positive constant b_2 such that for sufficiently large n ,*

$$b_2^{-1}(nh)^{-1} \leq D(0, \phi^*) \leq b_2(nh)^{-1}.$$

From this theorem we can see that if ϕ^* is defined by (1.7) with h such that $nh^3 = b_0$, then there exists a positive constant b_3 so that for sufficiently large n , $b_3^{-1}n^{-2/3} \leq D(0, \phi^*) \leq b_3n^{-2/3}$. From this fact we may expect existence of ϕ^* where $D(\theta, \phi^*)$ is of the best exact order $n^{-2/3}$, uniformly in $\theta \in \Omega^n$.

Appendix

The following theorem (A Fatou theorem for variances) is used in Section 2.

THEOREM A. *If $\{U_n\}$ is a sequence of random variables converging in distribution to a random variable U , then*

$$\liminf \text{Var}(U_n) \geq \text{Var}(U).$$

PROOF. It suffices to show that for $\{U_n\}$ such that $\text{Var}(U_n) \rightarrow \text{finite}$.

With $\mu_n = E U_n$ and $\sigma_n^2 = \text{Var} U_n$, the Chebychev inequality gives $P[|U_n - \mu_n| < \sqrt{2}\sigma_n] \geq 1/2$ while tightness provides a finite b independent of n for which $P[|U_n| \leq b] > 1/2$. The nonemptiness of the intersection of these events shows $|\mu_n| < b + \sqrt{2}\sigma_n$ so that $\{\mu_n\}$ is bounded.

Letting $\{\mu_m\}$ be a convergent subsequence with limit μ_∞ , $U_m - \mu_m \xrightarrow{D} U - \mu_\infty$ and hence (cf. Loève [5] 11.4, A(i))

$$\lim \text{Var}(U_n) = \lim E(U_m - \mu_m)^2 \geq E(U - \mu_\infty)^2 \geq \text{Var} U.$$

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DAITO BUNKA UNIVERSITY*

REFERENCES

- [1] Breiman, L. (1968). *Probability*, Addison-Wesley.
- [2] Feller, W. (1971). *An Introduction to Probability Theory and its Applications Volume II* (2nd ed.), Wiley, New York.
- [3] Ferguson, T. S. (1967). *Mathematical Statistics and a Decision Theoretic Approach*, Academic Press, New York.
- [4] Fox, R. (1968). Contribution to compound decision theory and empirical Bayes squared-error loss estimation, RM-214, Department of Statistics and Probability, Michigan State University.
- [5] Loève, Michel (1963). *Probability Theory* (3rd ed.), Van Nostrand, Princeton.
- [6] Nogami, Y. (1975). A non-regular squared-error loss set-compound estimation problem, RM-345, Department of Statistics and Probability, Michigan State University.
- [7] Nogami, Y. (1978). The set-compound one-stage estimation in the nonregular family of distributions over the interval $(0, \theta)$, *Ann. Inst. Statist. Math.*, **30**, A, 35-43.
- [8] Nogami, Y. (1979). The k -extended set-compound estimation problem in a nonregular family of distributions over $[\theta, \theta+1)$, *Ann. Inst. Statist. Math.*, **31**, A, 169-176.
- [9] Singh, Radhey S. (1974). Estimation of derivatives of average of μ -densities and sequence-compound estimation in exponential families, RM-318, Department of Statistics and Probability, Michigan State University.
- [10] Wahba, Grace (1975). Optimal convergence properties of variable knot, kernel, and orthogonal series methods for density estimation, *Ann. Statist.*, **3**, 15-29.

* Now at The University of Tsukuba.