

THE STRONG UNIFORM CONSISTENCY OF NEAREST NEIGHBOR DENSITY ESTIMATES

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Let X_1, \dots, X_n be independent, identically distributed random vectors with values in \mathbb{R}^d and with a common probability density f . If $V_k(x)$ is the volume of the smallest sphere centered at x and containing at least k of the X_1, \dots, X_n then $f_n(x) = k/(nV_k(x))$ is a nearest neighbor density estimate of f . We show that if $k = k(n)$ satisfies $k(n)/n \rightarrow 0$ and $k(n)/\log n \rightarrow \infty$ then $\sup_x |f_n(x) - f(x)| \rightarrow 0$ w.p. 1 when f is uniformly continuous on \mathbb{R}^d .

Introduction. Suppose that X_1, \dots, X_n are independent, identically distributed random vectors with values in \mathbb{R}^d and with a common probability density f . If $V_k(x)$ is the volume of the smallest sphere centered at x and containing at least k of the random vectors X_1, \dots, X_n , then Loftsgaarden and Quesenberry (1965), to estimate $f(x)$ from X_1, \dots, X_n , let

$$(1) \quad f_n(x) = k/(nV_k(x))$$

where $k = k(n)$ is a sequence of positive integers satisfying

$$(2) \quad \begin{array}{ll} \text{(a)} & k(n) \uparrow \infty \\ \text{(b)} & k(n)/n \rightarrow 0. \end{array}$$

(The factor $k - 1$ was used instead of k by Loftsgaarden and Quesenberry; this has no effect on any of the asymptotic results stated here.) They showed that $f_n(x)$ is a consistent estimate of $f(x)$ at each point where f is continuous and positive. This result can also easily be inferred from the work of Fix and Hodges (1951). For $d = 1$, Moore and Henrichon (1969) showed that

$$\sup_x |f_n(x) - f(x)| \rightarrow 0 \quad \text{in probability}$$

if f is uniformly continuous and positive on \mathbb{R} and if, additionally,

$$(3) \quad k(n)/\log n \rightarrow \infty.$$

Wagner (1973) showed that $f_n(x)$ is a strongly consistent estimate of $f(x)$ at each continuity point of f if, in addition to (2b),

$$(4) \quad \sum_1^\infty e^{-\alpha k(n)} < \infty \quad \text{for all } \alpha > 0.$$

(Notice that (4) is always implied by (3) but (2a) and (4) are needed to imply (3).) The result of this paper is the following theorem.

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THEOREM. *If f is uniformly continuous on \mathbb{R}^d and if $k(n)$ satisfies (2b) and (3) then*

$$\sup_x |f_n(x) - f(x)| \rightarrow_n 0 \quad \text{w.p. 1.}$$

If

$$\hat{f}_n(x) = \sum_{i=1}^n K((x - X_i)/r(n))/nr(n)^d,$$

where K is the uniform probability density for the unit sphere in \mathbb{R}^d and $\{r(n)\}$ is a sequence of positive numbers, the recent results of Moore and Yackel (1977) (see Theorem 3.1) and the above theorem immediately yield that

$$\sup_x |\hat{f}_n(x) - f(x)| \rightarrow 0 \quad \text{w.p. 1}$$

whenever f is uniformly continuous on \mathbb{R}^d and $r(n) \rightarrow 0$, $nr(n)^d/\log n \rightarrow \infty$. This fact, an improvement over the previously published convergence results for the kernel estimate with a uniform kernel (e.g., see Theorem 2.1 of Moore and Yackel (1977)), also is a special case of Theorem 4.9 of Devroye (1976) who proves the same statement for all kernels K which are bounded probability densities with compact support and whose discontinuity points have a closure with Lebesgue measure 0.

PROOF. To simplify notation we assume below that multiplications are always carried out before division. Let $\varepsilon > 0$ and choose $\delta > 0$ such that

$$|f(y) - f(x)| < \varepsilon/2$$

whenever x and y are within a sphere of volume δ . Deferring measurability arguments for the moment,

$$\begin{aligned} P\{\sup_x |f_n(x) - f(x)| > \varepsilon\} \\ = P\{\bigcup_x [V_k(x) < k/n(f(x) + \varepsilon)]\} + P\{\bigcup_{x: f(x) > \varepsilon} [V_k(x) > k/n(f(x) - \varepsilon)]\}. \end{aligned}$$

The event $\bigcup_x [V_k(x) < k/n(f(x) + \varepsilon)]$ implies that, for some x , there must be a sphere centered at x with volume less than $k/n(f(x) + \varepsilon)$ and containing k of the random vectors X_1, \dots, X_n . If $k/n\varepsilon < \delta$ then the probability measure of such a sphere must be less than $k(f(x) + \varepsilon/2)/n(f(x) + \varepsilon)$ so that, for one of these spheres S ,

$$\begin{aligned} \mu_n(S) - \mu(S) &> \frac{k}{n} - \frac{k(f(x) + \varepsilon/2)}{n(f(x) + \varepsilon)} \\ &= \frac{k\varepsilon}{2n(f(x) + \varepsilon)} \geq \frac{k\varepsilon}{2n(F + \varepsilon)} \end{aligned}$$

where F is the maximum of f on \mathbb{R}^d , μ is the measure on the Borel subsets of \mathbb{R}^d corresponding to f and μ_n is the empirical measure on the Borel subsets of \mathbb{R}^d for X_1, \dots, X_n . Thus, for $k/n\varepsilon < \delta$,

$$\begin{aligned} (5) \quad P\{\bigcup_x [V_k(x) < k/n(f(x) + \varepsilon)]\} \\ \leq P\{\sup_{S \in \mathcal{S}_n} |\mu_n(S) - \mu(S)| > k\varepsilon/2n(F + \varepsilon)\} \end{aligned}$$

where \mathcal{S}_n is the class of all spheres in \mathbb{R}^d whose volume is less than $4k/n\varepsilon$.

Next, with $4k/n\varepsilon < \delta$,

$$\bigcup_{x: f(x) > \varepsilon} [V_k(x) > k/n(f(x) - \varepsilon)] \subseteq \bigcup_{x: f(x) > \varepsilon} [V_k(x) > k/n(f(x) - (3\varepsilon/4))]$$

which implies that, for some x with $f(x) > \varepsilon$, there is a sphere S centered at x , with volume $\leq 4k/n\varepsilon$, and

$$\begin{aligned} \mu(S) &\geq k(f(x) - \varepsilon/2)/n(f(x) - (\frac{3}{4})\varepsilon), \\ \mu_n(S) &\leq k/n, \quad \text{and} \\ \mu(S) - \mu_n(S) &\geq k\varepsilon/4n(f(x) - (\frac{3}{4})\varepsilon). \end{aligned}$$

Thus

$$(6) \quad P\{\bigcup_{x: f(x) > \varepsilon} [V_k(x) > k/n(f(x) - \varepsilon)]\} \leq P\{\sup_{S \in \mathcal{S}_n} |\mu(S) - \mu_n(S)| \geq k\varepsilon/4nF\},$$

so that

$$P\{\sup_x |f_n(x) - f(x)| \geq \varepsilon\} \leq 2P\{\sup_{S \in \mathcal{S}_n} |\mu_n(S) - \mu(S)| \geq k\varepsilon/4n(F + \varepsilon)\}.$$

The proof will be completed if we show that for each $\varepsilon > 0$

$$(7) \quad \sum_n P\{\sup_{S \in \mathcal{S}_n} |\mu_n(S) - \mu(S)| \geq k\varepsilon/4n(F + \varepsilon)\} < \infty.$$

To prove (7) we employ a variation of the argument used by Vapnik and Chervonenkis (1971). In this variation use will be made of the following result. If Y_1, \dots, Y_n represent independent drawings without replacement from a population of k 0's and 1's then, for $\varepsilon > 0$ and $k \geq n$,

$$(8) \quad P\{ |(\sum_1^n Y_i)/n - \mu| \geq \varepsilon \} \leq 2e^{-n\varepsilon^2/(2\mu + \varepsilon)}$$

where μ , the {number of 1's}/ k , is assumed to be $\leq \frac{1}{2}$. Additionally (8) holds when Y_1, \dots, Y_n are Bernoulli random variables with parameter $\mu \leq \frac{1}{2}$. (Use the two-sided version of Theorem 3 of Hoeffding (1963) along with $\mu \leq \frac{1}{2}$ and $\log(1 + (\varepsilon/\mu)) \geq 2\varepsilon/(2\mu + \varepsilon)$. See also Section 6 of this paper.)

Now, if $\sup_{\mathcal{S}} \mu(A) \leq M$ and $n \geq 8M/\delta^2$, an easy modification of Lemma 1 of Vapnik and Chervonenkis (1971) yields

$$(9) \quad P[\sup_{\mathcal{S}} |\mu_n(A) - \mu(A)| \geq \delta] \leq 2P[\sup_{\mathcal{S}} |\mu_n(A) - \mu'_n(A)| \geq \delta/2]$$

where $\mu'_n(A)$ is the empirical measure for A with X_{n+1}, \dots, X_{2n} and \mathcal{S} is any class of Borel sets in \mathbb{R}^d for which

$$\sup_{\mathcal{S}} |\mu_n(A) - \mu(A)| \quad \text{and} \quad \sup_{\mathcal{S}} |\mu_n(A) - \mu'_n(A)|$$

are random variables. Putting $\mathcal{S} = \mathcal{S}_n$ we see that M can be taken to be $4kF/n\varepsilon$. Since, for $\alpha > 0$,

$$(10) \quad \begin{aligned} &P[\sup_{\mathcal{S}_n} |\mu_n(A) - \mu'_n(A)| \geq \delta/2] \\ &\leq P[\sup_{\mathcal{S}_n} |\mu_n(A) - \mu'_n(A)| \geq \delta/2; \sup_{\mathcal{S}_n} \mu_{2n}(A) \leq \alpha M] \\ &\quad + P[\sup_{\mathcal{S}_n} \mu_{2n}(A) > \alpha M] \end{aligned}$$

we see, using (3) and putting $\delta = k\varepsilon/4n(F + \varepsilon)$, that (7) follows whenever both

terms of the right-hand side of (10) are summable for some $\alpha > 0$. Looking at the first term, we note that it equals

$$\int_{\mathbb{R}^{2nd}} \frac{1}{(2n)!} \sum I_{[\sup_{\mathcal{S}_n} |\mu_n(A) - \mu'_n(A)| \geq \delta/2]} I_{[\sup_{\mathcal{S}_n} \mu_{2n}(A) \leq \alpha M]} dQ$$

where I_E is the indicator of the set $E \subseteq \mathbb{R}^d$ and Q is the probability measure on \mathbb{R}^{2nd} for X_1, \dots, X_{2n} and where the inner summation is taken over all $(2n)!$ permutations of x_1, \dots, x_{2n} . But this last integral equals

$$\begin{aligned} & \int_{\mathbb{R}^{2nd}} \frac{1}{(2n)!} \sum I_{[\sup_{\mathcal{S}_n} \mu_{2n}(A) \leq \alpha M]} \sup_{\mathcal{S}'_n} I_{[|\mu_n(A) - \mu'_n(A)| \geq \delta/2]} dQ \\ &= \int_{\mathbb{R}^{2nd}} \frac{1}{(2n)!} \sum I_{[\sup_{\mathcal{S}_n} \mu_{2n}(A) \leq \alpha M]} \sup_{\mathcal{S}'_n} I_{[|\mu_n(A) - \mu'_n(A)| \geq \delta/2]} dQ \\ &\leq \int_{\mathbb{R}^{2nd}} \sum_{A \in \mathcal{S}'_n} I_{[\sup_{\mathcal{S}_n} \mu_{2n}(A) \leq \alpha M]} \left\{ \frac{1}{(2n)!} \sum I_{[|\mu_n(A) - \mu_{2n}(A)| \geq \delta/4]} \right\} dQ \end{aligned}$$

where $\mathcal{S}'_n = \mathcal{S}'_n(x_1, \dots, x_{2n})$ is any finite subclass of \mathcal{S}_n which yields the same class of intersections with $\{x_1, \dots, x_{2n}\}$ and where the inner summation is again taken over the $(2n)!$ permutations of x_1, \dots, x_{2n} . The quantity within $\{\cdot\}$ is bounded above, using (8), by

$$2e^{-n\delta^2/(32\mu_{2n}(A)+4\delta)}$$

whenever $\mu_{2n}(A) \leq \frac{1}{2}$. Since $M = 4kF/n\epsilon$ we see, from (3), that for all n sufficiently large the last integral is upper-bounded by

$$2 \int_{\mathbb{R}^{2nd}} e^{-n\delta^2/(32\alpha M+4\delta)} (\sum_{A \in \mathcal{S}'_n} 1) dQ.$$

Choosing \mathcal{S}'_n to be a smallest possible subclass, we have (Vapnik and Chervonenkis (1971), Cover (1965)) that $(\sum_{A \in \mathcal{S}'_n} 1) \leq 1 + (2n)^{d+3}$ and, using (3) again, that the first term of (10) is summable for all $\alpha > 0$.

Looking at the second term of (10), let r be the radius of a sphere in \mathbb{R}^d whose volume is $4k/n\epsilon$. If some sphere of radius r contains l of the points X_1, \dots, X_{2n} then there must be at least one sphere or radius $2r$, centered at one of the points X_1, \dots, X_{2n} which contains at least l points. Thus

$$P[\sup_{\mathcal{S}_n} \mu_{2n}(A) > \alpha M] \leq 2nP[\mu_{2n}(S_{X_1}(2r)) > \alpha M]$$

where $S_x(t)$ denotes the sphere of radius t centered at x . But

$$\begin{aligned} & P[\mu_{2n}(S_{X_1}(2r)) > \alpha M] \\ & \leq \max_{x \in \mathbb{R}^d} P[\mu_{2n-1}(S_x(2r)) > (\alpha 2nM - 1)/(2n - 1)] \\ & \leq \max_{x \in \mathbb{R}^d} P[\mu_{2n-1}(S_x(2r)) - \mu(S_x(2r)) > [(\alpha 2nM - 1)/(2n - 1)] - 2^d 4kF/n\epsilon]. \end{aligned}$$

At this point it is not difficult, using (3) and (8), to show that the second term of (9) is summable as long as $\alpha > 2^d$.

Finally, to complete the proof, it is easy to see that all of the uncountable unions over x are indeed events and that the various supremums over \mathcal{S}_n are indeed random variables.

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