

# A REMARK ON THE APPROXIMATION OF THE SAMPLE DF IN THE MULTIDIMENSIONAL CASE

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Let  $X_1, X_2, \dots, X_n$  be i.i.d.r.v.s on the  $k$ -dimensional unit cube with

$$(1) \quad P(X_1 < \mathbf{t}) = \lambda(\mathbf{t}) \quad \text{if} \quad \mathbf{0} \leq \mathbf{t} \leq \mathbf{1},$$

where  $\mathbf{t} = (t_1, t_2, \dots, t_k)$ , and  $\mathbf{1} = (1, 1, \dots, 1)$  are  $k$ -dimensional vectors,

$\lambda(\mathbf{t}) = \prod_{i=1}^k t_i$  is the  $k$ -dimensional volume of the rectangle determined by the origin and the point  $\mathbf{t}$ , and for two vectors  $\mathbf{a}$  and  $\mathbf{b}$  the inequality  $\mathbf{a} < \mathbf{b}$  means that each coordinate of  $\mathbf{a}$  is less than the corresponding coordinate of  $\mathbf{b}$ . The empirical distribution function  $F_n(\mathbf{t})$  based on the sample  $X_1, X_2, \dots, X_n$  is the function

$$(2) \quad F_n(\mathbf{t}) = \frac{1}{n} \sum_{i: X_i < \mathbf{t}} 1 \quad \text{if} \quad \mathbf{0} \leq \mathbf{t} \leq \mathbf{1},$$

and the  $k$ -dimensional Brownian bridge  $B(\mathbf{t})$  is defined by

$$(3) \quad B(\mathbf{t}) = W(\mathbf{t}) - \lambda(\mathbf{t})W(\mathbf{1}) \quad \text{if} \quad \mathbf{0} \leq \mathbf{t} \leq \mathbf{1},$$

where  $W(\mathbf{t})$  is a  $k$ -dimensional Wiener process, i.e.,  $W(\mathbf{t})$  is a Gaussian process with independent increments, variance equal to the  $k$ -dimensional volume. In this remark we investigate the approximation of  $F_n(\mathbf{t})$  by  $B(\mathbf{t})$  in the case  $k = 2$ .

The one-dimensional case was investigated in [2], where we proved that there is a version of  $F_n$  and  $B$  such that

$$(4) \quad P\left(\sup_{\mathbf{0} \leq \mathbf{t} \leq \mathbf{1}} |n(F_n(\mathbf{t}) - \lambda(\mathbf{t})) - n^{\frac{1}{2}}B(\mathbf{t})| > C \log n + x\right) < Ke^{-\lambda x}$$

olds for all  $x$ , where  $C, K, \lambda$  are positive absolute constants (Theorem 3). Investigating the approximation of the whole sequence  $\{F_n, n = 1, 2, \dots\}$  we

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got a theorem of two-dimensional character (Theorem 4). As we shall see, the result presented here is strongly connected with this latter theorem.

**THEOREM.** *In the case  $k = 2$  for any  $n$  there is a version of  $F_n$  and  $B$  such that*

$$(5) \quad P\left(\sup_{0 \leq t \leq 1} |n(F_n(t) - \lambda(t)) - n^{\frac{1}{2}} B(t)| > (C \log n + x) \log n\right) < K e^{-\lambda x}$$

*holds for all  $x$ , where  $C, K, \lambda$  are positive absolute constants.*

**PROOF.** The proof is based on the following version of Theorem 5 of [2]:

**LEMMA.** *In the case  $k = 1$  for any  $n$  there are measurable functions*

$$(6) \quad e_i(w, f) \quad i = 1, 2, \dots, n$$

*defined on the product of the spaces  $C(0, 1)$  and  $D(0, 1)$  such that for any independent pair of the one-dimensional Wiener process  $W$  and empirical DF  $F_n$  the random variables*

$$(7) \quad \varepsilon_i = e_i(W, F_n) \quad i = 1, 2, \dots, n$$

*are i.i.d.r.v-s with distribution*

$$(8) \quad P(\varepsilon_i = 0) = P(\varepsilon_i = 1) = \frac{1}{2},$$

*the set of random variables  $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n$  is independent of  $F_n$ , and*

$$(9) \quad P\left(\sup_{0 \leq t \leq 1} \left| \frac{n}{2} (\tilde{F}(t) - \lambda(t)) - \left(\frac{n}{2}\right)^{\frac{1}{2}} W(t) \right| > (C \log n + x)\right) < K e^{-\lambda x}$$

*holds for all  $x$ , where  $C, K, \lambda$  are positive absolute constants, and  $\tilde{F}_n$  is defined by*

$$(10) \quad \tilde{F}_n(t) = \frac{2}{n} \sum_{i=1}^{n F_n(t)} \varepsilon_i.$$

The proof of this lemma is similar to the proof of Theorem 5 of [2]. We shall use the conditional quantile transformation, and the dyadic scheme in the following way. Suppose  $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n$  are arbitrary i.i.d.r.v-s of distribution (8), and the set  $\{\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n\}$  is independent of  $F_n$ . Let  $m$  and

$$0 = t_0 < t_1 < \dots < t_m = 1$$

be arbitrary and let  $\tilde{F}_n$  defined by (10). The conditional distribution of  $\frac{n}{2} \tilde{F}_n(t_i)$  under the condition that  $\{F_n(t), 0 \leq t \leq 1\}$  and

$$\{\tilde{F}_n(t_0), \tilde{F}_n(t_1), \dots, \tilde{F}_n(t_{i-1}), \tilde{F}_n(t_{i+1}), \dots, \tilde{F}_n(t_m)\}$$

are given is a hypergeometric distribution with parameters depending on the condition, and the conditional distribution of  $\tilde{F}_n(\mathbf{1})$  given  $\{F_n(\mathbf{t}), \mathbf{0} \leq \mathbf{t} \leq \mathbf{1}\}$  is a binomial distribution with parameters  $\left(n, \frac{1}{2}\right)$ . Hence we can transform the appropriate parts of  $W(\mathbf{t})$  step by step to  $\tilde{F}_n(\mathbf{1}), \tilde{F}_n\left(\frac{1}{2}\right), \tilde{F}_n\left(\frac{1}{4}\right), \tilde{F}_n\left(\frac{3}{4}\right)$  using in each step the conditional distribution of the new variable on the condition that  $\{F_n(\mathbf{t}), \mathbf{0} \leq \mathbf{t} \leq \mathbf{1}\}$  and the just defined  $\tilde{F}_n(\mathbf{t}_i)$ -s are given. This is the same construction as the construction of the proof of Theorem 5, the only difference is that here  $\tilde{F}_n(\mathbf{1})$  is also random variable. Hence the further details of the proof are omitted.

The theorem follows from the lemma in the same way as Theorem 4 follows from Theorem 5 in [2]. Hence its proof is also omitted.

REMARK 1. In the case  $k = 1$  we proved in [2] that our result is the best possible in the sense that there are positive absolute constants  $A, B$  such that for any  $n$  and any version of  $F_n, B$

$$P\left(\sup_{\mathbf{0} \leq \mathbf{t} \leq \mathbf{1}} |n(F_n(\mathbf{t}) - \lambda(\mathbf{t})) - n^{\frac{1}{2}}B(\mathbf{t})| > A \log n\right) \geq B.$$

In the case  $k = 2$  the situation is different: we do not know whether for a given  $\varepsilon > 0$  are there positive constants  $A, B$  such that

$$(11) \quad P\left(\sup_{\mathbf{0} \leq \mathbf{t} \leq \mathbf{1}} |n(F_n(\mathbf{t}) - \lambda(\mathbf{t})) - n^{\frac{1}{2}}B(\mathbf{t})| > A(\log n)^{1+\varepsilon}\right) \geq B.$$

REMARK 2. In case  $k > 2$  the best known available results are given by CSÖRGŐ and RÉVÉSZ [1]. They give an approximation of order  $n^{\frac{k-1}{2k}}$ . We do not know whether (11) is true for any  $k > 1$  or not.

## REFERENCES

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