

An Estimate of the Remainder in a Combinatorial Central Limit Theorem

E. Bolthausen

Technische Universität Berlin, Fachbereich Mathematik, Straße des 17. Juni 135,
D-1000 Berlin 12

§1. Introduction

Let $A=(a_{ij})$ be a $n \times n$ matrix of real numbers. Let

$$\mu_A = na_{..}, \quad \sigma_A^2 = \sum_{i,j} (a_{ij} - a_{i.} - a_{.j} + a_{..})^2 / (n-1),$$

where

$$a_{i.} = \sum_j a_{ij} / n, \quad a_{.j} = \sum_i a_{ij} / n, \quad a_{..} = \sum_{i,j} a_{ij} / n^2.$$

Let further $\hat{a}_{ij} = (a_{ij} - a_{i.} - a_{.j} + a_{..}) / \sigma_A$. A theorem which has been proved under various conditions by Hoeffding [3], Motoo [5] and others states that if π is uniformly distributed on the set of permutations of $\{1, 2, \dots, n\}$ then $T_A = (\sum_i a_{i\pi(i)} - \mu) / \sigma = \sum_i \hat{a}_{i\pi(i)}$ is approximately standard normally distributed. We shall investigate the rate of convergence.

Estimates have been obtained by von Bahr [7] and Ho and Chen [2], but they yield the rate $O(n^{-1/2})$ only under some boundedness conditions, like $\sup_{i,j} |\hat{a}_{ij}| = O(n^{-1/2})$.

The special case where $a_{ij} = e_i d_j$ is of particular interest in non-parametric statistics and has been discussed by many authors, e.g. by Husková [4] in the case where the d_j satisfy boundedness and smoothness conditions and most recently and successfully by Does [1], whose results may cover most cases of statistical interest. Usually, d_j is assumed to be given by so-called score generating functions, e.g. $d_j = J(j/(n+1))$, where J is a function on $(0, 1)$, satisfying some smoothness assumption. The main advantages of the approach given here are that no smoothness is needed and the a_{ij} are completely general. Von Bahr and Ho and Chen allow the a_{ij} to be random independent of π . The extensions to cover this case are straightforward and therefore omitted.

Theorem. *There is an absolute constant $K > 0$, such that for all A with $\sigma_A^2 > 0$*

$$\sup_t |P(T_A \leq t) - \Phi(t)| \leq K \sum_{i,j} |\hat{a}_{ij}|^3/n$$

where Φ is the standard normal distribution function.

If one takes a sequence $(a_{ij}^{(n)})$ of $n \times n$ matrices, the theorem gives the convergence rate $n^{-1/2}$ if $\sum_{i,j} |\hat{a}_{ij}^{(n)}|^3/\sqrt{n}$ remains bounded and $\sigma^2 = 1$.

The proof given in §3 is simpler than the Fourier theoretic approaches used e.g. by Does [1]. It is based on an improvement of the Stein method. Stein’s method has also been used by Ho and Chen.

In §2 a proof of the classical Berry-Esseen Theorem is given using a version of the Stein method. This has also been done by Ho and Chen [2], but their proof depends on a concentration inequality and seems not to work for non-identically distributed variables. The approach given here is more flexible and the extension to the non-identically distributed case is straightforward. The proof in the simple situation of §2 gives the motivation for the proof of the theorem stated above.

§2. A Proof of the Classical Berry-Esseen Theorem

If $n \in \mathbb{N}$, $\gamma \geq 1$, let $\mathcal{L}(n, \gamma)$ be the set of sequences $\underline{X} = \{X_1, \dots, X_n\}$ of random variables, such that X_1, \dots, X_n are i.i.d. and $EX_i = 0$, $EX_i^2 = 1$, $E|X_i|^3 = \gamma$. If $\gamma < 1$ $\mathcal{L}(n, \gamma) = \emptyset$. Let $S_k = \sum_{i=1}^k X_i/\sqrt{n}$, $1 \leq k \leq n$. If $z, x \in \mathbb{R}$, $\lambda > 0$, let

$$h_{z, \lambda}(x) = ((1 + (z - x)/\lambda) \wedge 1) \vee 0, \quad h_{z, 0}(x) = 1_{(-\infty, z]}(x).$$

Let

$$\delta(\lambda, \gamma, n) = \sup \{ |E(h_{z, \lambda}(S_n)) - \Phi(h_{z, \lambda})| : z \in \mathbb{R}, \underline{X} \in \mathcal{L}(n, \gamma) \}.$$

Here $\Phi(g)$ is the standard normal expectation of g .

We write $\delta(\gamma, n) = \delta(0, \gamma, n)$. The Berry-Esseen theorem states that

$$\sup \{ \sqrt{n} \delta(\gamma, n)/\gamma : \gamma \geq 1, n \in \mathbb{N} \} < \infty. \tag{2.1}$$

By using $h_{z, 0} \leq h_{z, \lambda} \leq h_{z+\lambda, 0}$ one obtains

$$\delta(\gamma, n) \leq \delta(\lambda, \gamma, n) + \lambda/\sqrt{2\pi}. \tag{2.2}$$

It obviously suffices to bound $\sqrt{n} \delta(\gamma, n)/\gamma$ for $n \geq 2$ which is assumed from now on. We simply write h instead of $h_{z, \lambda}$ if there is no danger of confusion.

Let $f(x) = e^{x^2/2} \int_{-\infty}^x (h(z) - \Phi(h)) e^{-z^2/2} dz$, which satisfies

$$f'(x) - xf(x) = h(x) - \Phi(h). \tag{2.3}$$

If $x \leq 0$, then $|f(x)| \leq \Phi(x)/\varphi(x)$, where φ is the standard normal density and if $x > 0$: $|f(x)| \leq (1 - \Phi(x))/\varphi(x)$. Therefore

$$|f(x)| \leq 1; \quad |xf(x)| \leq 1; \quad |f'(x)| \leq 2 \quad \text{for all } x. \tag{2.4}$$

(The last estimate by using (2.3)). From this one has

$$\begin{aligned} |f'(x+y) - f'(x)| &= |yf(x+y) + x(f(x+y) - f(x)) + h(x+y) - h(x)| \\ &\leq |y| \left(1 + 2|x| + \frac{1}{\lambda} \int_0^1 1_{[z, z+\lambda]}(x+sy) ds \right), \end{aligned} \tag{2.5}$$

$$\begin{aligned} E(f'(S_n) - S_n f(S_n)) &= E(f'(S_n) - \sqrt{n} X_n f(S_n)) \\ &= E \left\{ f'(S_n) - f'(S_{n-1}) - X_n^2 \int_0^1 \left(f' \left(S_{n-1} + t \frac{X_n}{\sqrt{n}} \right) - f'(S_{n-1}) \right) dt \right\} \end{aligned} \tag{2.6}$$

if $\underline{x} \in \mathcal{L}(n, \gamma)$. Using (2.5) one obtains

$$\begin{aligned} E|f'(S_n) - f'(S_{n-1})| &\leq E \left\{ \frac{|X_n|}{\sqrt{n}} \left(1 + 2|S_{n-1}| + \frac{1}{\lambda} \int_0^1 1_{[z, z+\lambda]} \left(S_{n-1} + t \frac{X_n}{\sqrt{n}} \right) dt \right) \right\} \\ &\leq \frac{c}{\sqrt{n}} (1 + \delta(\gamma, n-1)/\lambda) \end{aligned}$$

where we used the independence of S_{n-1} and X_n . Here and in the future c is used as a positive constant which depends only on the formula where it appears.

Similarly

$$E \left| X_n^2 \int_0^1 \left(f' \left(S_{n-1} + t \frac{X_n}{\sqrt{n}} \right) - f'(S_{n-1}) \right) dt \right| \leq \frac{c\gamma}{\sqrt{n}} (1 + \delta(\gamma, n-1)/\lambda).$$

Implementing these estimates into (2.6) and using (2.3) and (2.2) one obtains

$$\delta(\gamma, n) \leq \frac{c\gamma}{\sqrt{n}} (1 + \delta(\gamma, n-1)/\lambda) + \lambda/\sqrt{2\pi}.$$

Choosing now $\lambda = 2c\gamma/\sqrt{n}$ (c here the same as above), then $\delta(\gamma, n) \leq c\gamma/\sqrt{n} + \delta(\gamma, n-1)/2$. Using $\delta(\gamma, 1) \leq 1$, this proves (2.1).

§3. Proof of the Theorem

c again denotes a constant which depends only on the formula where it appears. In contrast, c_1, c_2, \dots are positive constants which depend on nothing.

Let $\beta_A = \sum_{i,j} |\hat{a}_{ij}|^3$. If $n_0 \in \mathbb{N}$ and $\varepsilon_0 > 0$ are arbitrary but fixed then the statement of the theorem is true if $n \leq n_0$ or $\beta_A > \varepsilon_0 n$ (β_A is bounded from below). Therefore, we assume that $n > n_0$ and $\beta_A \leq \varepsilon_0 n$ where ε_0, n_0 will be specified later on but $n_0 \geq 4$.

We first need a truncation: Let

$$a'_{ij} = \begin{cases} \hat{a}_{ij} & \text{if } |\hat{a}_{ij}| \leq 1/2 \\ 0 & \text{if } |\hat{a}_{ij}| > 1/2 \end{cases}$$

and $\Gamma = \{(i, j) : |\hat{a}_{ij}| > 1/2\}$. Clearly, $|\Gamma|$, the number of elements in Γ , is at most $8\beta_A$. Therefore,

$$\begin{aligned} P(\sum_i a'_{i\pi(i)} \neq T_A) &\leq P(\sum_i 1_\Gamma(i, \pi(i)) \geq 1) \\ &\leq E(\sum_i 1_\Gamma(i, \pi(i))) \\ &= |\Gamma|/n \leq 8\beta_A/n. \end{aligned} \tag{3.1}$$

If A' is the matrix $(a'_{i, j})$, then

$$\begin{aligned} |\mu_{A'}| &= \left| \frac{1}{n} \sum_{i,j} a'_{ij} \right| \leq \frac{1}{n} \sum_{(i,j) \in \Gamma} |\hat{a}_{ij}| \\ &\leq \frac{1}{n} |\Gamma|^{2/3} \beta_A^{1/3} \leq c\beta_A/n. \end{aligned} \tag{3.2}$$

We claim that

$$|\sigma_{A'}^2 - 1| \leq c\beta_A/n. \tag{3.3}$$

$$\begin{aligned} |\sigma_{A'}^2 - 1| &= |E(\sum_i a'_{i\pi(i)})^2 - \mu_{A'}^2 - E(\sum_i \hat{a}_{i\pi(i)})^2| \\ &\leq |E(\sum_i a'^2_{i\pi(i)} + \sum_{i \neq j} a'_{i\pi(i)} a'_{j\pi(i)} - \sum_i \hat{a}^2_{i\pi(i)} - \sum_{i \neq j} \hat{a}_{i\pi(i)} \hat{a}_{j\pi(i)})| + \mu_{A'}^2 \\ &= \left| -\frac{1}{n} \sum_{(i,j) \in \Gamma} \hat{a}^2_{ij} + \frac{1}{n(n-1)} \sum_{i \neq j} \sum_{k \neq l} (a'_{ik} a'_{jl} - \hat{a}_{ik} \hat{a}_{jl}) \right| + \mu_{A'}^2 \\ &\leq \frac{1}{n} \sum_{(i,j) \in \Gamma} \hat{a}^2_{ij} + \frac{2}{n(n-1)} \left| \sum_{\substack{i \neq j, k \neq l \\ (i,k) \in \Gamma}} \hat{a}_{ik} \hat{a}_{jl} \right| + \mu_{A'}^2, \end{aligned}$$

$$\begin{aligned} \frac{1}{n} \sum_{(i,j) \in \Gamma} \hat{a}^2_{ij} &\leq c\beta_A/n \quad \text{and} \quad \frac{1}{n(n-1)} \left| \sum_{(i,k) \in \Gamma} \sum_{j \neq i, l \neq k} \hat{a}_{ik} \hat{a}_{jl} \right| \\ &= \frac{1}{n(n-1)} \sum_{(i,k) \in \Gamma} \hat{a}^2_{ik} \leq c\beta_A/n. \end{aligned}$$

Therefore (3.3) is true.

If we choose ϵ'_0 small enough (depending on c in (3.3)), one has

$$|\sigma_{A'}^2 - 1| \leq 1/3 \quad \text{if } \beta_A \leq \epsilon'_0 n.$$

Therefore,

$$\begin{aligned} |\hat{a}'_{ij}| &= \left| \frac{1}{\sigma_{A'}} (a'_{ij} - a'_{i.} - a'_{.j} + a'_{..}) \right| \\ &\leq \frac{3}{4} + \frac{3}{2} (|a'_{i.}| + |a'_{.j}| + |a'_{..}|). \end{aligned}$$

By taking ε'_0 small enough, this is easily seen to be ≤ 1 . Therefore,

$$|\hat{a}'_{ij}| \leq 1. \tag{3.4}$$

A simple calculation shows that

$$\beta_{A'} = \sum_{i,j} |\hat{a}'_{ij}|^3 \leq c_1 \beta_A \tag{3.5}$$

if $\gamma > 0$, let $M_n(\gamma)$ be the set of $n \times n$ matrices B satisfying

$$\sigma_B^2 \neq 0, \quad |\hat{b}_{ij}| \leq 1 \quad \text{and} \quad \sum_{i,j} |\hat{b}_{ij}|^3 \leq \gamma.$$

Let

$$\delta(\lambda, \gamma, n) = \sup \{ |Eh_{z, \lambda}(T_B) - \Phi(h_{z, \lambda})| : z \in \mathbb{R}, B \in M_n(\gamma) \}$$

and $\delta(\gamma, n) = \delta(0, \gamma, n)$.

The considerations above show that if A is a $n \times n$ matrix with $\sigma_A^2 > 0$ and $\beta_A \leq \varepsilon'_0 n$, then

$$\begin{aligned} \sup_t |P(T_A \leq t) - \Phi(t)| &\leq \sup_t |P(\sum_i a'_{i\pi(i)} \leq t) - \Phi(t)| + \frac{8\beta_A}{n} \\ &\leq \delta(c_1 \beta_A, n) + \sup_t |\Phi((t - \mu_{A'})/\sigma_{A'}) - \Phi(t)| + \frac{8\beta_A}{n} \\ &\leq \delta(c_1 \beta_A, n) + c\beta_A/n. \end{aligned} \tag{3.6}$$

We now want to show that $\delta(\gamma, n) \leq c\gamma/n$, which together with (3.6) proves the theorem.

As $\beta_A \geq c\sqrt{n}$, we may assume that $\gamma \geq 1$. We fix $A \in M_n(\gamma)$ and estimate $|Eh_{z, \lambda}(T_A) - \Phi(h_{z, \lambda})|$. Of course, we may assume $a_{ij} = \hat{a}_{ij}$ and therefore $a_{i.} = a_{.i} = 0$; $\frac{1}{n-1} \sum_{i,j} a_{ij}^2 = 1$. We denote the set of these matrices by $M_n^0(\gamma)$.

In order to apply the method of §2 we need some manipulations on the permutations which replace the independence of the summands in §2.

We define a random element (I_1, I_2, J_1, J_2) in N^4 , where $N = \{1, \dots, n\}$, in the following way: (I_1, I_2, J_1) is uniformly distributed on N^3 , and given this, one has $J_2 = J_1$ on $\{I_1 = I_2\}$ and J_2 is uniformly distributed on $N - \{J_1\}$ on $\{I_1 \neq I_2\}$. Let π_1 be a random permutation, which is uniformly distributed on the permutations of N and independent of (I_1, I_2, J_1, J_2) . Define

$$\begin{aligned} I_3 &= \pi_1^{-1}(J_1), & I_4 &= \pi_1^{-1}(J_2), & J_3 &= \pi_1(I_1), & J_4 &= \pi_1(I_2). \\ \underline{I} &= (I_1, I_2, I_3, I_4), & \underline{J} &= (J_1, J_2, J_3, J_4). \end{aligned} \tag{3.7}$$

Of course, $I_1 = I_2$ holds if and only if $I_3 = I_4$. For each fixed $i = (i_1, i_2, i_3, i_4) \in N^4$ which satisfies the condition $i_1 = i_2 \Leftrightarrow i_3 = i_4$, we fix once for all a permutation $t(i)$ of N , which maps i_1 to i_4 and i_2 to i_3 and which leaves the numbers outside $\{i_1, i_2, i_3, i_4\}$ fixed. Let further $s(i_1, i_2)$ be the transposition of i_1 and i_2 . We put $\pi_2 = \pi_1 \circ t(\underline{I})$, $\pi_3 = \pi_2 \circ s(i_1, i_2)$.

Lemma. a) π_1, π_2, π_3 have the same law.

b) π_2 is independent of (I_1, J_1) .

Proof. A simple calculation shows that I and π_1 are independent. Given I , π_2 is a one to one function of π_1 . Therefore, π_2 is also uniformly distributed and independent of I .

Given π_2 , (I_1, I_2) is a one to one function of (I_1, J_1) . As (I_1, I_2) is uniformly distributed on N^2 , it follows that π_2 is independent of (I_1, J_1) . This proves b). Using this, an identical argument as that above shows that π_3 is uniformly distributed.

Let $T_i = \sum_j a_{j\pi_i(j)}$, $i = 1, 2, 3$; $\Delta T_i = T_{i+1} - T_i$, $i = 1, 2$. ΔT_1 depends on (I, J) , ΔT_2 on (I_1, I_2, J_1, J_2) .

Therefore, if $f = f_{z, \lambda}$ from §2:

$$\begin{aligned} E(T_A f(T_A)) &= E(T_3 f(T_3)) = nE(a_{I_1, J_1} f(T_3)) \\ &= nE \left(a_{I_1, J_1} \Delta T_2 \int_0^1 (f'(T_1 + \Delta T_1 + t \Delta T_2) - f'(T_1)) dt \right) \\ &\quad + nE(a_{I_1, J_1} \Delta T_2 f'(T_1)), \end{aligned}$$

where we used the independence of T_2 and a_{I_1, J_1} (Lemma b). Using the independence of π_1 and (I_1, I_2, J_1, J_2) the second summand above equals $nE(a_{I_1, J_1} \Delta T_2) E(f'(T_1))$. Using this and (2.5), one obtains

$$\begin{aligned} &|E(f'(T_A) - T_A f(T_A))| \\ &\leq 2|nE(a_{I_1, J_1} \Delta T_2) - 1| + n \int_0^1 E \left\{ |a_{I_1, J_1} \Delta T_2 (\Delta T_1 + t \Delta T_2)| \left(1 + 2|T_1| \right. \right. \\ &\quad \left. \left. + \frac{1}{\lambda} \int_0^1 \mathbf{1}_{[z, z+\lambda]}(T_1 + s \Delta T_1 + st \Delta T_2) ds \right) \right\} dt \\ &\leq 2|nE(a_{I_1, J_1} \Delta T_2) - 1| + nE(|a_{I_1, J_1} \Delta T_2| (|\Delta T_1| + |\Delta T_2|)) \\ &\quad + 2nE(|a_{I_1, J_1} \Delta T_2| (|\Delta T_1| + |\Delta T_2|) |T_1|) \\ &\quad + \frac{n}{\lambda} E \left(|a_{I_1, J_1} \Delta T_2| (|\Delta T_1| + |\Delta T_2|) \right. \\ &\quad \left. \cdot \int_0^1 \int_0^1 \mathbf{1}_{[z, z+\lambda]}(T_1 + s \Delta T_1 + st \Delta T_2) ds dt \right) \\ &= A_1 + A_2 + A_3 + A_4, \quad \text{say.} \end{aligned} \tag{3.8}$$

A_1 is 0 and A_2 is easy to estimate:

$$E(|a_{I_1, J_1} \Delta T_2| (|\Delta T_1| + |\Delta T_2|))$$

contains summands of the form

$$E(|a_{I_1, J_1} a_{\alpha\beta} a_{\mu\nu}|)$$

where

$$\alpha, \mu \in \{I_1, \dots, I_4\}, \quad \beta, \nu \in \{J_1, \dots, J_4\}.$$

An easy calculation shows that these summands are bounded by $c\gamma/n^2$. So

$$A_2 \leq c\gamma/n. \tag{3.9}$$

In order to estimate A_3 and A_4 we look at the conditional distribution of T_1 given $\underline{I}=\underline{i}, \underline{J}=\underline{j}$. T_1 depends only on π_1 and the conditional distribution of π_1 is easy to describe: π_1 takes any permutation φ which satisfies $\varphi(i_k)=j_k, 1 \leq k \leq 4$, with equal probability. If B is the matrix which is obtained from A by cancelling the rows i_1, i_2, i_3, i_4 and the columns j_1, j_2, j_3, j_4 , the T_1 , conditioned on $\underline{I}=\underline{i}$ and $\underline{J}=\underline{j}$ has the same law as

$$\sum_{i \in \{i_1, \dots, i_4\}} a_{i\pi_1(i)} + \sum_{j=1}^{n-l} b_{j\sigma(j)},$$

where l is the number of (distinct) elements in $\{i_1, \dots, i_4\}$ and σ is uniformly distributed on the permutations of $\{1, \dots, n-l\}$. As $|a_{ij}| \leq 1$ for all i, j , one has

$$\begin{aligned} E(|T_1| | \underline{I}=\underline{i}, \underline{J}=\underline{j}) &\leq 4 + E(|\sum_i b_{i\sigma(i)}|) \\ &\leq 4 + (E(\sum_i b_{i\sigma(i)}^2))^{1/2} \end{aligned}$$

which is bounded, uniformly in \underline{i} and \underline{j} . Therefore, A_3 can be estimated in the same way as A_2 , leading to

$$A_3 \leq c\gamma/n. \tag{3.10}$$

If $1 \leq l \leq 4$, let $M_n^l(\gamma)$ be the set of $(n-l) \times (n-l)$ -matrices, which can be obtained from matrices in $M_n^0(\gamma)$ by cancelling l rows and l columns. Introducing

$$\alpha(\lambda, \gamma, n) = \sup \{ \|P(T_1 \in [z, z + \lambda] | \underline{I}, \underline{J})\|_\infty : z \in \mathbb{R}, A \in M_n^0(\gamma) \}$$

we have:

$$\alpha(\lambda, \gamma, n) \leq \sup \left\{ P \left(\sum_{i=1}^{n-l} b_{i\sigma(i)} \in [z, z + \lambda] \right) : z \in \mathbb{R}, B \in M_n^l(\gamma), 1 \leq l \leq 4 \right\}. \tag{3.11}$$

If $B \in M_n^l(\gamma)$, then $|b_{i.}|, |b_{.j}|, |b_{.l}|$ are $\leq c/n$. Using this, one obtains

$$\left| \sigma_B^2 - \frac{1}{n-l-1} \sum_{i,j} a_{ij}^2 \right| \leq \frac{1}{n-l-1} \sum' a_{ij}^2 + o(1),$$

where the \sum' is the sum over the cancelled matrix elements. Furthermore, $\sum' a_{ij}^2 \leq cn\epsilon_0^{2/3}$ if $\beta_A \leq \epsilon_0 n$ and if $\epsilon_0 \leq \epsilon'_0$ is small enough and n_0 sufficiently large, we have $|\sigma_B^2 - 1| \leq 1/2$ and therefore $\sigma_B^2 \geq 1/2$. Using this, one gets

$$\beta_B = \sum_{i,j} |\hat{b}_{ij}|^3 \leq c_2 \beta_A. \tag{3.12}$$

Therefore

$$\sup_z P(\sum_i b_{i\sigma(i)} \in [z, z + \lambda]) \leq \sup_z P(\sum_i \hat{b}_{i\sigma(i)} \in [z, z + 2\lambda]),$$

which, if $c_2 \varepsilon_0 \leq \varepsilon'_0$ according to (3.6), is

$$\begin{aligned} &\leq 2\delta(c_1 \beta_B, n-l) + c \beta_B/n + \lambda/\sqrt{2\pi} \\ &\leq 2\delta(c_1 c_2 \beta_A, n-l) + c_2 c \beta_A/n + \lambda/\sqrt{2\pi}. \end{aligned}$$

Using (3.11), one has

$$\alpha(\lambda, \gamma, n) \leq 2 \max_{1 \leq l \leq 4} \delta(c_1 c_2 \beta_A, n-l) + c \left(\lambda + \frac{\beta_A}{n} \right).$$

Using this with the estimate (3.9) of A_2 , one obtains

$$A_4 \leq \frac{c\gamma}{n} \left(1 + \frac{1}{\lambda} \frac{\gamma}{n} + \frac{1}{\lambda} \max_{1 \leq l \leq 4} \delta(c_1 c_2 \gamma, n-l) \right).$$

Combining this with (3.9) and (3.10) in (3.8) and using (2.2) and (2.3) one obtains

$$\delta(\gamma, n) \leq c_3 \frac{\gamma}{n} \left(1 + \frac{1}{\lambda} \frac{\gamma}{n} + \frac{1}{\lambda} \max_{1 \leq l \leq 4} \delta(c_1 c_2 \gamma, n-l) \right) + \frac{\lambda}{\sqrt{2\pi}}.$$

Now one may choose λ at ones liking, so we take $\lambda = 2c_1 c_2 c_3 \gamma/n$ leading to

$$\delta(\gamma, n) \leq \frac{c\gamma}{n} + \frac{1}{2c_1 c_2} \delta(c_1 c_2 \gamma, n-l),$$

and if $n \geq 8$, this gives

$$\sup_{\gamma} \frac{n\delta(\gamma, n)}{\gamma} \leq c + \frac{1}{2} \max_{1 \leq l \leq 4} \sup_{\gamma} \frac{(n-l)}{\gamma} \delta(\gamma, n-l).$$

This proves $\delta(\gamma, n) \leq c\gamma/n$ and, using again (3.6), this proves the theorem.

References

1. Does, R.J.M.M.: Berry-Esseen theorems for simple linear rank statistics. *Ann. Probability* **10**, 982-991 (1982)
2. Ho, S.T., Chen, L.H.Y.: An L_p bound for the remainder in a combinatorial central limit theorem. *Ann. Probability* **6**, 231-249 (1978)
3. Hoeffding, W.: A combinatorial central limit theorem. *Ann. Math. Statist.* **22**, 558-566 (1951)
4. Husková, Marie: The Berry-Esseen theorem for rank statistics. *Comment. Math. Univ. Carolina*, **20**, 399-415 (1979)
5. Motoo, M.: On the Hoeffding's combinatorial central limit theorem. *Ann. Inst. Statist. Math.* **8**, 145-154 (1957)
6. Stein, Ch.: A bound for the error in the normal approximation to the distribution of a sum of dependent random variables. *Proc. Sixth Berkeley Sympos. Math. Statist. Probability* **2**, 583-602 (1972)
7. von Bahr, B.: Remainder term estimate in a combinatorial limit theorem. *Z. Wahrscheinlichkeitstheorie verw. Gebiete* **35**, 131-139 (1976)