On self–normalised sums. Supplement.

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Abstract

This supplement aims to correct a misprint and clarify the point about the accuracy of Berry–Esseen–type inequalities for self–normalised sums and Student's statistic established in [1].

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Let X, X_1, X_2, \ldots be a sequence of independent and identically distributed ran-Let $X, X_1, X_2, ...$ be a sequence of independent and identically distributed random variables. Denote $t_n = \sum_{i=1}^n X_i / \sqrt{\sum_{i=1}^n X_i^2}$ (here and in the sequel we use the notation from [1]).

Theorems 9 and 10 in [1] establish uniform and non–uniform Berry–Esseen–type inequalities for self–normalised sums t_n and Student's statistic.

The lower and upper bounds for $\mathbb{P}(t_n < x) - \Phi(x)$ in [1] are given with explicit constants, and the only moment assumption is the finiteness of the 3rd moment. The upper bound is of order $n^{-1/2}$. The lower bound is of order $n^{-2/7}$; it is of order $n^{-1/2}$ if the 4th moment is finite. In particular,

$$
\sup_{x} |\mathbb{P}(t_n < x) - \Phi(x)| \le C_* \frac{\mathbb{E}|X|^3}{\sqrt{n}} + \frac{8|\mathbb{E}X|^3}{e\sqrt{2\pi n}} + \frac{\sqrt{\mathbb{D}X^2}}{\sqrt{n}} \left(\frac{2}{\pi e}\right)^{1/4} + o(n^{-1/2}), \quad (1)
$$

where C_* be the constant from the uniform Berry–Esseen inequality (the term $8\mathbb{E}|X|^3/(e\sqrt{2\pi n})$ in [1], p. 424, may be replaced by $8|\mathbb{E}X|^3/(e\sqrt{2\pi n})$.

The author has received a feedback that the bounds of Theorems 9 and 10 are not easy to understand. One distinguished scientist wrote that (1) was the only inequality with explicit constants he was able to find in [1].

The aim of this letter is to clarify that point and correct a misprint. We show that Theorems 9 and 10 provide bounds with explicit constants. Moreover, the bounds are sharp in the following sense: for a class of probability distributions, the estimates are asymptotically as sharp as those of the uniform and non–uniform Berry–Esseen inequalities.

There is a misprint in the definition of $R(t, a, b)$ on page 423: "c" in the denominator must be erased. Examples below illustrate the sharpness of the Berry– Esseen–type inequalities for self–normalised sums.

Example 1. Let $\mathbb{P}(X = -1) = \mathbb{P}(X = 1) = 1/2$. Then Theorem 9 with $\varepsilon = 0$ and $N = 1$ yields

$$
\Delta_n \equiv \sup_x |\mathbb{P}(t_n < x) - \Phi(x)| \le C_* / \sqrt{n} \tag{2}
$$

The right–hand side of (2) coincides with that of the Berry–Esseen theorem.

Put $\varepsilon_{n,x} = 0$, and notice that

$$
\rho_n = p_n = \delta_{n,x} = \gamma_n^+ = 0, \ R_n^+(x,3) = C_+ n^{-1/2},
$$

where C_+ the constant from the non–uniform Berry–Esseen inequality. Theorem 10 yields

$$
|\mathbb{P}(t_n < x) - \Phi(x)| \le C_+ n^{-1/2} (1 + x^3)^{-1} \tag{3}
$$

.

The right–hand side of estimate (3) coincides with that of the non–uniform Berry– Esseen inequality.

Example 2. Let $\mathbb{P}(X = -N) = \mathbb{P}(X = N) = 1/(2N^2)$, $\mathbb{P}(X = 0) = 1 - N^{-2}$. Denote $R_{n,N} = \frac{N}{\sqrt{n}}$ \overline{n} h $C_*^{\overline{1/3}}+C_{+}^{\overline{1/3}}N/\sqrt{n}$ بر
3 and $r_{n,N} = \frac{24\sqrt{6} N^2}{e\sqrt{\pi n}}$ $\frac{d^2\sqrt{6}N^2}{e\sqrt{\pi e}n}$. It is easy to check that

$$
\mathbb{E}X = 0, \mathbb{E}|X|^m = N^{m-2} \ (m \in \mathbb{N}), \ \rho_n = m_0 = 0,
$$

$$
\beta_n = e^{-n/(36N^2)}, \ \max\{R_n(3, 1, 1); R_n(3, 2, 2)\} \le R_{n,N},
$$

$$
r_n < 2r_{n,N}, \ \mathbb{P}\left(\sum_{i=1}^n (X_i^2 - 1)/n > \sqrt{2\varepsilon}\right) \le e^{-\varepsilon n/(2N^2)}
$$

Theorem 9 gives

$$
\Phi(x(1-\varepsilon)) - R_{n,N} - 2r_{n,N} - e^{-\varepsilon n/(2N^2)} \le \mathbb{P}(t_n < x) \le \Phi(x) + R_{n,N} + r_{n,N} \,. \tag{4}
$$

Assume that $N = N(n) \to \infty$, $N^2/n \to 0$. With $\varepsilon = 2n^{-1}N^2 \ln(n/N^2)$, (4) implies

$$
\Delta_n \le C_* N n^{-1/2} + O(n^{-1}N^2 \ln (n/N^2)).
$$

In other words, estimate (4) is asymptotically as sharp as that of the Berry–Esseen theorem.

In order to apply Theorem 10, note that $\gamma_n^+ \leq \gamma_{n,x}^+$ and $R_n^+(x,3) \leq R_{n,x}^+$, where

$$
\gamma_{n,x}^+ = 9x^2 N^2 / 25n \,, \ R_{n,x}^+ = C_+ N n^{-1/2} \left(1 + 3xN / (5\sqrt{n}) \right)^3.
$$

Besides, $\mu_3^> = 0$ if $x \geq 6N/\sqrt{n}$. Theorem 10 yields

$$
\Phi\left(x(1-\varepsilon)(1-\gamma_{n,x}^{+})\right) - R_{n,x}^{+}\left(1+x^{3}(1-\gamma_{n,x}^{+})^{3}/(1-\varepsilon)^{3}\right)^{-1} - e^{-\varepsilon n/(2N^{2})}
$$
\n
$$
\leq \mathbb{P}(t_{n} < x) \leq \Phi(x) + R_{n,x}^{+}\left(1+x^{3}(1-\gamma_{n,x}^{+})^{3}\right)^{-1} + e^{-n/(36N^{2})} \tag{5}
$$

as $6N/\sqrt{n} \leq x \leq$ √ $\overline{n}/(3N)$. With $\varepsilon = 2n^{-1}N^2 \ln(nN^{-2}) (2 + x^2)$, inequality (5) implies

$$
-C_{+}\left(1+3xN/(5\sqrt{n})\right)^{3} - v_{n}n^{-1}N^{2}\ln(n/N^{2}) \leq (1+x^{3})\left[\mathbb{P}(t_{n} < x) - \Phi(x)\right]\sqrt{n}/N
$$

$$
\leq C_{+}\left(1+3xN/(5\sqrt{n})\right)^{3}\left(1+x^{3}\right)\left(1+x^{3}(1-\gamma_{n,x}^{+})\right)^{-1} + v_{n}n^{-1}N^{2},
$$

where $v_n \to 0$ uniformly in $x \in \left[\frac{6N}{\sqrt{n}}\right]$; $\sqrt{n}/(3N)$. Let $\{u_n\}$ be a sequence of positive numbers such that $u_n \to 0$ and $u_n \sqrt{n}/N \to \infty$. Then

$$
(1+x^3)|P(t_n < x) - \Phi(x)|\sqrt{n}/N \le C_+ + o(1)
$$
\n(6)

uniformly in $x \in [0; u_n]$ √ \overline{n}/N .

In the expanding interval $[0; u_n]$ √ \overline{n}/N , estimate (6) is asymptotically as sharp as that of the non–uniform Berry–Esseen bound. We notice in [2] that a non–uniform Berry–Esseen–type inequality for self–normalised sums may not, in general, hold on the whole line.

References

- [1] S.Y. Novak (2000) On self–normalised sums. Math. Methods Statist., v. 9, No 4, 415–436.
- [2] S.Y. Novak (2001) On self–normalised sums and Student's statistic. Brunel University: Technical Report No TR/18/01.