



IMPROVEMENT OF THE NON-UNIFORM VERSION OF BERRY-ESSEEN INEQUALITY VIA PADITZ-SIGANOV THEOREMS

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ABSTRACT. We improve the constant in a non-uniform bound of the Berry-Esseen inequality without assuming the existence of the absolute third moment by using the method obtained from the Paditz-Siganov theorems. Our bound is better than the results of Thongtha and Neammanee in 2007 ([14]).

Key words and phrases: Berry-Esseen inequality, Paditz-Siganov theorems, central limit theorem, uniform and non-uniform bounds.

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1. INTRODUCTION AND MAIN RESULTS

The Berry-Esseen inequality is one of the most important inequalities in the theory of probability. This inequality was independently discovered by two mathematicians, Andrew C. Berry ([2]) and Carl-Gustav Esseen ([5]) in 1941 and 1945 respectively. Let X_1, X_2, \dots, X_n be independent random variables with zero mean and $\sum_{i=1}^n E[X_i]^2 = 1$. Define $W_n = X_1 + X_2 + \dots + X_n$. Then $\text{Var } W_n = 1$. Let F_n be the distribution function of W_n and Φ the standard normal distribution function, i.e.,

$$F_n(x) = P(W_n \leq x) \quad \text{and} \quad \Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{t^2}{2}} dt.$$

The central limit theorem shows that F_n converges pointwise to Φ as $n \rightarrow \infty$ and the bounds of this convergence are,

$$(1.1) \quad \sup_{x \in \mathbb{R}} |P(W_n \leq x) - \Phi(x)| \leq C_0 \sum_{i=1}^n E|X_i|^3$$

and

$$(1.2) \quad |P(W_n \leq x) - \Phi(x)| \leq \frac{C_1}{1 + |x|^3} \sum_{i=1}^n E|X_i|^3$$

for uniform and non-uniform versions respectively, where both C_0 and C_1 are positive constants and stated under the assumption that $E|X_i|^3 < \infty$ for $i = 1, 2, \dots, n$.

In the case of identical X_i 's, Siganov ([11]) and Chen ([5]) improved the constant down to 0.7655 and 0.7164, respectively. For non-uniform bounds, Nageav ([7]) was the first to obtain (1.2) and Michel ([6]) calculated the constant to be 30.84.

Without assuming identically distributed X'_i 's, Beek ([15]) sharpened the constant down to 0.7975 in 1972 for the uniform version. The best bound was found by Siganov ([11]) in 1986.

Theorem 1.1 (Siganov, 1986). *Let X_1, X_2, \dots, X_n be independent random variables such that $EX_i = 0$ and $E|X_i|^3 < \infty$ for $i = 1, 2, \dots, n$. Assume that $\sum_{i=1}^n EX_i^2 = 1$. Then*

$$\sup_{x \in \mathbb{R}} |P(W_n \leq x) - \Phi(x)| \leq 0.7915 \sum_{i=1}^n E|X_i|^3,$$

where $W_n = X_1 + X_2 + \dots + X_n$.

For the non-uniform version, Bikeliis ([1]) generalized (1.2) to this case and Paditz ([9]) calculated C_1 to be 114.7 in 1977. He also improved his result down to 31.935 in 1989.

Theorem 1.2 (Paditz ([10]), 1989). *Under the assumptions of Theorem 1.1, we have*

$$|P(W_n \leq x) - \Phi(x)| \leq \frac{31.935}{1 + |x|^3} \sum_{i=1}^n E|X_i|^3.$$

In 2001, Chen and Shao ([3]) gave new versions of (1.1) and (1.2) without assuming the existence of third moments. Their results are

$$(1.3) \quad \sup_{x \in \mathbb{R}} |P(W_n \leq x) - \Phi(x)| \leq 4.1 \sum_{i=1}^n \{E|X_i|^2 I(|X_i| \geq 1) + E|X_i|^3 I(|X_i| < 1)\}$$

and

$$(1.4) \quad |P(W_n \leq x) - \Phi(x)| \leq C_2 \sum_{i=1}^n \left\{ \frac{EX_i^2 I(|X_i| \geq 1 + |x|)}{(1 + |x|)^2} + \frac{E|X_i|^3 I(|X_i| < 1 + |x|)}{(1 + |x|)^3} \right\},$$

where C_2 is a positive constant and $I(A)$ is an indicator random variable such that

$$I(A) = \begin{cases} 1 & \text{if } A \text{ is true,} \\ 0 & \text{otherwise.} \end{cases}$$

In 2005, Neammanee ([8]) combined the concentration inequality in ([3]) with a coupling approach to calculate the constant in (1.4), giving,

$$(1.5) \quad \begin{aligned} |P(W_n \leq x) - \Phi(x)| \\ \leq C_3 \sum_{i=1}^n \left\{ \frac{EX_i^2 I(|X_i| \geq 1 + |\frac{x}{4}|)}{(1 + |\frac{x}{4}|)^2} + \frac{E|X_i|^3 I(|X_i| < 1 + |\frac{x}{4}|)}{(1 + |\frac{x}{4}|)^3} \right\}, \end{aligned}$$

where C_3 is 21.44 for large values of x such that $|x| \geq 14$.

Thongtha and Neammanee ([14]) improved the concentration inequality used in ([8]) and gave a better constant, i.e., 9.7 for $|x| \geq 14$. The method which was used in ([8]) is Stein's method which was first introduced by Stein ([12]) in 1972. In this work, we provide a better constant by using Paditz-Siganov theorems. The results are as follows.

Theorem 1.3. *We have*

$$|P(W_n \leq x) - \Phi(x)| \leq C \sum_{i=1}^n \left\{ \frac{EX_i^2 I(|X_i| \geq 1 + |x|)}{(1 + |x|)^2} + \frac{E|X_i|^3 I(|X_i| < 1 + |x|)}{(1 + |x|)^3} \right\},$$

where

$$C = \begin{cases} 49.89 & \text{if } 0 \leq |x| < 1.3, \\ 59.45 & \text{if } 1.3 \leq |x| < 2, \\ 73.52 & \text{if } 2 \leq |x| < 3, \\ 76.17 & \text{if } 3 \leq |x| < 7.98, \\ 45.80 & \text{if } 7.98 \leq |x| < 14, \\ 39.39 & \text{if } |x| \geq 14. \end{cases}$$

To compare Theorem 1.3 with the result of Thongtha and Neammanee ([14]) in (1.5), we give Corollary 1.4.

Corollary 1.4. *We have*

$$|P(W_n \leq x) - \Phi(x)| \leq C \sum_{i=1}^n \left\{ \frac{EX_i^2 I(|X_i| \geq 1 + |\frac{x}{4}|)}{(1 + |\frac{x}{4}|)^2} + \frac{E|X_i|^3 I(|X_i| < 1 + |\frac{x}{4}|)}{(1 + |\frac{x}{4}|)^3} \right\},$$

where

$$C = \begin{cases} 9.54 & \text{if } 0 \leq |x| < 1.3, \\ 19.74 & \text{if } 1.3 \leq |x| < 2, \\ 18.38 & \text{if } 2 \leq |x| < 3, \\ 14.63 & \text{if } 3 \leq |x| < 7.98, \\ 5.13 & \text{if } 7.98 \leq |x| < 14, \\ 3.55 & \text{if } |x| \geq 14. \end{cases}$$

We note from Corollary 1.4 that our result is better than a bound from Thongtha and Neammanee in ([14]).

2. PROOF OF THE MAIN RESULTS

In this section, we will prove Theorem 1.3 by using the Paditz-Siganov theorems. Corollary 1.4 can be obtained easily from Theorem 1.3. To prove these results, let

$$\begin{aligned} Y_{i,x} &= X_i I(|X_i| < 1 + x), & S_x &= \sum_{i=1}^n Y_{i,x}, \\ \alpha_x &= \sum_{i=1}^n EX_j^2 I(|X_j| \geq 1 + x), & \beta_x &= \sum_{i=1}^n E|X_j|^3 I(|X_j| < 1 + x), \\ \gamma_x &= \frac{\beta_x}{2} \quad \text{and} \quad \delta_x = \frac{\alpha_x}{(1+x)^2} + \frac{\beta_x}{(1+x)^3} \text{ for } x > 0. \end{aligned}$$

Proposition 2.1. *For each $n \in \mathbb{N}$, we have*

- (1) $\sum_{i=1}^n E|Y_{i,x} - EY_{i,x}|^3 \leq \beta_x + \frac{7\alpha_x}{1+x}$,
- (2) $1 - 2\alpha_x \leq \text{Var } S_x \leq 1$, and
- (3) If $\alpha_x \leq 0.11$, then $0 < \frac{1}{\sqrt{\text{Var } S_x}} \leq 1 + 1.452\alpha_x$.

Proof. 1. By the fact that

$$(2.1) \quad |EX_i I(|X_i| < 1+x)| = |EX_i I(|X_i| \geq 1+x)|,$$

$$E|X_i|^2 \leq \sum_{i=1}^n EX_i^2 = 1 \quad \text{and} \quad E^2 X_i \leq EX_i^2,$$

we have

$$\begin{aligned} & \sum_{i=1}^n E|Y_{i,x} - EY_{i,x}|^3 \\ &= \sum_{i=1}^n E|X_i I(|X_i| < 1+x) - EX_i I(|X_i| < 1+x)|^3 \\ &\leq \sum_{i=1}^n [E|X_i|^3 I(|X_i| < 1+x) + 3EX_i^2 I(|X_i| < 1+x)|EX_i I(|X_i| < 1+x)| \\ &\quad + 3E|X_i I(|X_i| < 1+x)|E^2 X_i I(|X_i| < 1+x)| + |EX_i I(|X_i| < 1+x)|^3] \\ &\leq \sum_{i=1}^n E|X_i|^3 I(|X_i| < 1+x) + 3 \sum_{i=1}^n |EX_i I(|X_i| < 1+x)| \\ &\quad + 3 \sum_{i=1}^n E|X_i| |EX_i I(|X_i| < 1+x)| |EX_i I(|X_i| < 1+x)| \\ &\quad + \sum_{i=1}^n E|X_i|^2 I(|X_i| < 1+x) |EX_i I(|X_i| < 1+x)| \\ &\leq \beta_x + 3 \sum_{i=1}^n |EX_i I(|X_i| \geq 1+x)| + 3 \sum_{i=1}^n E|X_i|^2 |EX_i I(|X_i| \geq 1+x)| \\ &\quad + \sum_{i=1}^n |EX_i I(|X_i| \geq 1+x)| \\ &\leq \beta_x + 3 \sum_{i=1}^n E|X_i| I(|X_i| \geq 1+x) + 3 \sum_{i=1}^n E|X_i| I(|X_i| \geq 1+x) \\ &\quad + \sum_{i=1}^n E|X_i| I(|X_i| \geq 1+x) \\ &= \beta_x + 7 \sum_{i=1}^n E|X_i| I(|X_i| \geq 1+x) \\ &\leq \beta_x + 7 \sum_{i=1}^n \frac{E|X_i|^2 I(|X_i| \geq 1+x)}{(1+x)} = \beta_x + \frac{7\alpha_x}{(1+x)}. \end{aligned}$$

2. By (2.1), we note that

$$\begin{aligned} \text{Var } S_x &= \sum_{i=1}^n \text{Var } Y_{i,x} = \sum_{i=1}^n (EY_{i,x}^2 - E^2 Y_{i,x}) \\ &= \sum_{i=1}^n EX_i^2 I(|X_i| < 1+x) - \sum_{i=1}^n E^2 X_i I(|X_i| < 1+x) \end{aligned}$$

$$\begin{aligned}
&= 1 - \sum_{i=1}^n E X_i^2 I(|X_i| \geq 1+x) - \sum_{i=1}^n E^2 X_i I(|X_i| \geq 1+x) \\
(2.2) \quad &= 1 - \alpha_x - \sum_{i=1}^n E^2 X_i I(|X_i| \geq 1+x).
\end{aligned}$$

From this and the fact that $\alpha_x \geq 0$, we have $\text{Var } S_x \leq 1$.

By (2.2), we have

$$\begin{aligned}
\text{Var } S_x &= 1 - \alpha_x - \sum_{i=1}^n E^2 X_i I(|X_i| \geq 1+x) \\
&\geq 1 - \alpha_x - \sum_{i=1}^n E X_i^2 I(|X_i| \geq 1+x) \\
&= 1 - 2\alpha_x.
\end{aligned}$$

Hence, $1 - 2\alpha_x \leq \text{Var } S_x \leq 1$.

3. For $0 < t \leq 0.11$, by using Taylor's formula, we have

$$\begin{aligned}
\frac{1}{\sqrt{1-2t}} &= 1 + \frac{t}{(1-2c)^{\frac{3}{2}}} \text{ for some } c \in (0, 0.11] \\
&\leq 1 + \frac{t}{(1-2(0.11))^{\frac{3}{2}}} \\
&\leq 1 + 1.452t.
\end{aligned}$$

From this fact and 2., we have

$$0 < \frac{1}{\sqrt{\text{Var } S_x}} \leq \frac{1}{\sqrt{1-2\alpha_x}} \leq 1 + 1.452\alpha_x$$

for $\alpha_x \leq 0.11$. □

Proposition 2.2. For each $x > 0$, let $\bar{Y}_{i,x} = \frac{Y_{i,x} - EY_{i,x}}{\sqrt{\text{Var } S_x}}$ and $\bar{S}_x = \sum_{i=1}^n \bar{Y}_{i,x}$.

(1) If $\alpha_x \leq 0.099$ and $1.3 \leq x \leq 2$, then

$$\left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right| \leq \frac{54.513\alpha_x}{(1+x)^2} + \frac{41.195\beta_x}{(1+x)^3}.$$

(2) If $(1+x)^2\alpha_x < \frac{1}{5}$, then

$$\left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right| \leq \frac{C_1\alpha_x}{(1+x)^2} + \frac{C_2\beta_x}{(1+x)^3}$$

where $C_1 = 57.186$ $C_2 = 73.515$ for $2 \leq x < 3$,

$C_1 = 33.318$ $C_2 = 76.17$ for $3 \leq x < 7.98$,

$C_1 = 3.976$ $C_2 = 45.8$ for $7.98 \leq x < 14$, and

$C_1 = 1.226$ $C_2 = 39.382$ for $x \geq 14$.

Proof. 1. By Proposition 2.1(1) of ([14]) and Proposition 2.1(2), we have

$$(2.3) \quad |ES_x| \leq \frac{\alpha_x}{1+x} \leq 0.043 \text{ and } 1 \geq \text{Var } S_x \geq 0.802$$

which imply

$$(2.4) \quad 0 \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}} \leq \frac{2 + 0.043}{\sqrt{0.802}} = 2.2813.$$

By Proposition 2.1(1) and (2.3),

$$\begin{aligned}
 \sum_{i=1}^n E|\bar{Y}_{i,x}|^3 &= \sum_{i=1}^n E \left| \frac{Y_{i,x} - EY_{i,x}}{\sqrt{\text{Var } S_x}} \right|^3 \\
 &= \frac{1}{(\text{Var } S_x)^{\frac{3}{2}}} \sum_{i=1}^n E|Y_{i,x} - EY_{i,x}|^3 \\
 &\leq \frac{1}{(\text{Var } S_x)^{\frac{3}{2}}} \left(\beta_x + \frac{7\alpha_x}{1+x} \right) \\
 (2.5) \quad &= 1.3923\beta_x + 4.2375\alpha_x.
 \end{aligned}$$

Note that $\bar{S}_x = \sum_{i=1}^n \bar{Y}_{i,x}$ is the sum of independent random variables whose

$$E\bar{Y}_{i,x} = 0 \text{ and } \text{Var } \bar{S}_x = 1.$$

By (2.5) and Theorem 1.1,

$$\begin{aligned}
 |P(\bar{S}_x \leq z) - \Phi(z)| &\leq 0.7915 \sum_{i=1}^n E|\bar{Y}_{i,x}|^3 \\
 &\leq 0.7915(1.3923\beta_x + 4.2375\alpha_x) \\
 &\leq 1.102\beta_x + 3.354\alpha_x
 \end{aligned}$$

for all $z \in \mathbb{R}$. From this fact, (2.3) and (2.4), we have

$$\begin{aligned}
 &\left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right| \\
 &\leq \frac{\left(1 + \left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right)\right)^3 (1.102\beta_x + 3.354\alpha_x)}{\left(1 + \left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right)\right)^3} \\
 &\leq \frac{(3.2813)^3 (1.102\beta_x + 3.354\alpha_x)}{\left(1 + \left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right)\right)^3} \\
 &\leq \frac{38.933\beta_x + 118.495\alpha_x}{(0.957 + x)^3} \\
 &\leq \frac{41.195\beta_x}{(1+x)^3} + \frac{125.379\alpha_x}{(1+x)^3} \\
 &\leq \frac{41.195\beta_x}{(1+x)^3} + \frac{54.513\alpha_x}{(1+x)^2}
 \end{aligned}$$

where we use the fact that

$$\frac{1+x}{0.957+x} \leq 1.019 \text{ for all } 1.3 < x < 2$$

in the fourth inequality.

2. **Case 2** $2 \leq x < 3$.

We can prove the result of this case by using the same argument as 1.

Case 3 $3 \leq x < 7.98$.

To bound $|P(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right)|$ in 1., we used Theorem 1.1.

But in this case, we will use Theorem 1.2.

We note that

$$(2.6) \quad 0 \leq \alpha_x \leq 0.0125, \quad 1 \geq \text{Var } S_x \geq 0.975,$$

and, by Proposition 2.1(1) of ([14]), $|ES_x| \leq 0.00313$.

Then, for $3 \leq x \leq 7.98$,

$$\frac{1}{1 + \left(\frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right)^3} \leq \frac{2.29}{\left(1 + \frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right)^3} \quad \text{and} \quad \sum_{i=1}^n E|\bar{Y}_{i,x}|^3 \leq 1.039\beta_x + 1.819\alpha_x.$$

From these facts, (2.6) and Theorem 1.2, we have

$$\begin{aligned} & \left| P\left(\bar{S}_x \leq \frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right) \right| \\ & \leq \frac{(31.935) \sum_{i=1}^n E|\bar{Y}_{i,x}|^3}{1 + \left(\frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right)^3} \leq \frac{(31.935)(2.29) \sum_{i=1}^n E|\bar{Y}_{i,x}|^3}{\left(1 + \frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right)^3} \\ & \leq \frac{73.131(1.039\beta_x + 1.819\alpha_x)}{(0.99687 + x)^3} \leq \frac{(1.0008)^3(75.983\beta_x + 132.952\alpha_x)}{(1+x)^3} \\ & \leq \frac{76.17\beta_x}{(1+x)^3} + \frac{133.27\alpha_x}{(1+x)^3} \leq \frac{76.17\beta_x}{(1+x)^3} + \frac{33.318\alpha_x}{(1+x)^2}, \end{aligned}$$

where we use the fact that

$$\frac{1+x}{0.99687+x} \leq 1.0008 \quad \text{for all } 3 \leq x < 7.98$$

in the fourth inequality.

Case $x \geq 7.98$.

We can prove the result of this case by using the same argument as the case $3 \leq x < 7.98$. \square

We are now ready to prove Theorem 1.3.

Proof of Theorem 1.3. It suffices to consider only $x \geq 0$ as we can simply apply the results to $-W_n$ when $x < 0$.

Case 1. $0 \leq x < 1.3$.

Note that for $x \geq 0$,

$$EX_i^2 I(|X_i| \geq 1) + E|X_i|^3 I(|X_i| < 1) \leq EX_i^2 I(|X_i| \geq 1+x) + E|X_i|^3 I(|X_i| < 1+x)$$

and for $0 \leq x \leq 1.3$, $(1+x)^3 \leq 12.167$.

From these facts and (1.3), we have

$$\begin{aligned} & |P(W_n \leq x) - \Phi(x)| \\ & \leq 4.1 \sum_{i=1}^n \left\{ EX_i^2 I(|X_i| \geq 1) + E|X_i|^3 I(|X_i| < 1) \right\} \\ & \leq 4.1 \sum_{i=1}^n \left\{ EX_i^2 I(|X_i| \geq 1+x) + E|X_i|^3 I(|X_i| < 1+x) \right\} \\ & \leq \frac{4.1(12.167)}{(1+x)^3} \sum_{i=1}^n \left\{ EX_i^2 I(|X_i| \geq 1+x) + E|X_i|^3 I(|X_i| < 1+x) \right\} \end{aligned}$$

$$\leq 49.89 \sum_{i=1}^n \left\{ \frac{EX_i^2 I(|X_i| \geq 1+x)}{(1+x)^2} + \frac{E|X_i|^3 I(|X_i| < 1+x)}{(1+x)^3} \right\}.$$

Before proving another case, we need the equation

$$(2.7) \quad |P(W_n \leq x) - \Phi(x)| \leq \frac{4.931\alpha_x}{(1+x)^2} + \left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right|$$

for $\alpha_x \leq 0.11$ and $x \geq 1.3$.

By (2.9) of ([14]), it suffices to show that for $\alpha_x \leq 0.11$ and $x \geq 1.3$,

$$(2.8) \quad |P(S_x \leq x) - \Phi(x)| \leq \frac{3.319\alpha_x}{(1+x)^2} + \left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right|.$$

By Proposition 2.1(1) and Proposition 2.1(2), we have

$$\frac{x - ES_x}{\sqrt{\text{Var } S_x}} \geq x - ES_x \geq x - \frac{\alpha_x}{1+x},$$

which implies

$$\min \left\{ x, \frac{x - ES_x}{\sqrt{\text{Var } S_x}} \right\} \geq x - \frac{\alpha_x}{1+x}.$$

From this and the fact that

$$\Phi(b) - \Phi(a) = \frac{1}{\sqrt{2\pi}} \int_a^b e^{-t^2/2} dt \leq \frac{1}{\sqrt{2\pi} e^{a^2/2}} \int_a^b 1 dt = \frac{(b-a)}{\sqrt{2\pi} e^{a^2/2}}$$

for $0 < a < b$, we have

$$(2.9) \quad \begin{aligned} & |P(S_x \leq x) - \Phi(x)| \\ & \leq \left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right| + \left| \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi(x) \right| \\ & \leq \left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right| \\ & \quad + \frac{1}{\sqrt{2\pi} e^{\frac{1}{2} \left[\min \left(x, \frac{x - ES_x}{\sqrt{\text{Var } S_x}} \right) \right]^2}} \left| \frac{x}{\sqrt{\text{Var } S_x}} - x - \frac{ES_x}{\sqrt{\text{Var } S_x}} \right| \\ & \leq \left| P\left(\bar{S}_x \leq \frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x - ES_x}{\sqrt{\text{Var } S_x}}\right) \right| \\ & \quad + \frac{1}{\sqrt{2\pi} e^{\frac{1}{2} \left(x - \frac{\alpha_x}{1+x} \right)^2}} \left| \frac{x}{\sqrt{\text{Var } S_x}} - x - \frac{ES_x}{\sqrt{\text{Var } S_x}} \right|. \end{aligned}$$

Note that for $x \geq 1.3$

$$e^{\frac{x^2}{2}} \geq 0.933(1+x), \quad e^{\frac{x^2}{2}} \geq 0.193(1+x)^3$$

and

$$e^{\frac{1}{2} \left(x - \frac{\alpha_x}{1+x} \right)^2} \geq e^{\frac{x^2}{2} - \left(\frac{x}{1+x} \right) \alpha_x} \geq 0.89 e^{\frac{x^2}{2}}.$$

From these facts, Proposition 2.1(1), Proposition 2.1(3) and $\alpha_x \leq 0.11$, we have

$$\frac{1}{\sqrt{2\pi} e^{\frac{1}{2} \left(x - \frac{\alpha_x}{1+x} \right)^2}} \left| \frac{x}{\sqrt{\text{Var } S_x}} - x - \frac{ES_x}{\sqrt{\text{Var } S_x}} \right|$$

$$\begin{aligned}
&\leq \frac{1}{\sqrt{2\pi} \left(0.89e^{\frac{x^2}{2}}\right)} \left| \frac{x}{\sqrt{\text{Var } S_x}} - x \right| + \frac{1}{\sqrt{2\pi} \left(0.89e^{\frac{x^2}{2}}\right)} \left| \frac{ES_x}{\sqrt{\text{Var } S_x}} \right| \\
&\leq \frac{1.452\alpha_x x}{\sqrt{2\pi}(0.89)(0.193)(1+x)^3} + \frac{\alpha_x}{(1+x)} \frac{(1+1.452\alpha_x)}{\sqrt{2\pi}(0.89)(0.933)(1+x)} \\
&\leq \frac{3.373\alpha_x x}{(1+x)^3} + \frac{0.558\alpha_x}{(1+x)^2} \leq \frac{3.931\alpha_x}{(1+x)^2}.
\end{aligned}$$

From this fact, (2.8) and (2.9), we have (2.7)

Case 2. $1.3 \leq x < 2$.

By the fact that $|P(W_n \leq x) - \Phi(x)| \leq 0.55$ ([3, pp. 246]), we can assume $\frac{\alpha_x}{(1+x)^2} \leq 0.011$, i.e. $\alpha_x \leq 0.099$.

From this fact, (2.7) and Proposition 2.2(1), we have

$$\begin{aligned}
|P(W_n \leq x) - \Phi(x)| &\leq \frac{4.931\alpha_x}{(1+x)^2} + \left| P\left(\bar{S}_x \leq \frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right) - \Phi\left(\frac{x-ES_x}{\sqrt{\text{Var } S_x}}\right) \right| \\
&\leq \frac{4.931\alpha_x}{(1+x)^2} + \frac{41.195\beta_x}{(1+x)^3} + \frac{54.513\alpha_x}{(1+x)^2} \\
&= \frac{59.444\alpha_x}{(1+x)^2} + \frac{41.195\beta_x}{(1+x)^3} \leq 59.444\delta_x.
\end{aligned}$$

Case 3. $2 \leq x \leq 14$.

Subcase 3.1. $(1+x)^2\alpha_x \geq \frac{1}{5}$.

Using the same argument of subcase 1.1 in Theorem 1.2 of ([14]) and the facts that

$$(2.10) \quad \frac{1+x}{x} = 1 + \frac{1}{x} \leq 1.5 \text{ and } e^{\frac{x^2}{2}} \geq 0.92x^3 \text{ for } 2 \leq x \leq 14,$$

we can show that

$$|P(W_n \leq x) - \Phi(x)| \leq 37.408\delta_x.$$

Subcase 3.2. $(1+x)^2\alpha_x < \frac{1}{5}$.

Note that for $x \geq 2$, we have

$$0 \leq \alpha_x \leq \frac{1}{5(1+x)^2} \leq 0.023 \leq 0.11.$$

By (2.7) and Proposition 2.2(2), we obtain the required bounds.

Case 4. $x > 14$.

Follows the argument of case 3 on replacing the inequalities

$$e^{\frac{x^2}{2}} \geq 60x^3 \text{ and } \frac{1+x}{x} = 1 + \frac{1}{x} \leq 1.071$$

in (2.10). □

Proof of Corollary 1.4. If $0 \leq x < 1.3$.

We used the same argument as case 1 of Theorem 1.3 and the fact that $(1 + \frac{x}{4})^3 \leq 2.327$ to get $C = 9.54$.

Suppose that $x \geq 1.3$. By the fact that

$$\delta_x \leq \left(\frac{1 + \frac{x}{4}}{1+x} \right)^2 \delta_{\frac{x}{4}},$$

we have

$$\delta_x \leq \begin{cases} 0.332\delta_{\frac{x}{4}} & \text{if } 1.3 \leq x < 2, \\ 0.250\delta_{\frac{x}{4}} & \text{if } 2 \leq x < 3, \\ 0.192\delta_{\frac{x}{4}} & \text{if } 3 \leq x < 7.98, \\ 0.112\delta_{\frac{x}{4}} & \text{if } 7.98 \leq x < 14, \\ 0.090\delta_{\frac{x}{4}} & \text{if } x \geq 14. \end{cases}$$

Then Corollary 1.4 follows from this fact and Theorem 1.3. □

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