

ON BERRY-ESSEEN BOUNDS OF SUMMABILITY TRANSFORMS

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ABSTRACT. Let $Y_{n,k}$, $k = 0, 1, 2, \dots$, $n \geq 1$, be a collection of random variables, where for each n , $Y_{n,k}$, $k = 0, 1, 2, \dots$, are independent. Let $A = [p_{n,k}]$ be a regular summability method. We provide some rates of convergence (Berry-Esseen type bounds) for the weak convergence of summability transform (AY) . We show that when $A = [p_{n,k}]$ is the classical Cesàro summability method, the rate of convergence of the resulting central limit theorem is best possible among all regular triangular summability methods with rows adding up to one. We further provide some summability results concerning ℓ^2 -negligibility. An application of these results characterizes the rate of convergence of Schnabl operators while approximating Lipschitz continuous functions.

1. INTRODUCTION

Let $Y := [Y_{k,m}]$ be a matrix of random variables, where $(Y_{k,m}, k = 0, 1, 2, \dots)$ is the m -th column vector consisting of mutually independent random variables with finite variances, $m = 0, 1, 2, \dots$. Let $A := [p_{n,k}]$ be a summability matrix. Consider

$$S_n := \sum_{k=0}^{\infty} \frac{p_{n,k}(Y_{n,k} - E(Y_{n,k}))}{\|\sigma_n p_n\|_2} =: \sum_{k=0}^{\infty} X_{n,k},$$

where

$$X_{n,k} := \frac{p_{n,k}(Y_{n,k} - E(Y_{n,k}))}{\|\sigma_n p_n\|_2}, \quad \|\sigma_n p_n\|_2^2 := \sum_{k=0}^{\infty} p_{n,k}^2 \text{Var}(Y_{n,k}) < \infty.$$

We will present some results concerning the weak convergence of the sequence S_n , $n \geq 1$. Since different summability methods have different convergence fields, one expects to see the dependence of rates of convergence of S_n on the choice of A . Consequently, a natural question is to ask if there is a summability transform that leads to the fastest rate of convergence. We will show that the classical Cesàro transform provides the fastest rate among all regular triangular methods whose row sums equal one.

Convergence, in probability and almost sure sense, of such transforms have already been settled by various authors ([3], [4], [13], [16]). In this paper we will show

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that through summability theory we may unify the weak convergence part with the classical central limit theorem and the classical Berry-Esseen (BE) bound. The Cesàro method gives the classical result and the Abel method, for instance, gives the discounted central limit theorem of actuarial sciences ([10], [2]). The interesting part is that the tools remain the same as found in the standard books of probability (see [5], [8]).

The next section collects some summability results that are relevant for our later developments, and perhaps may have some independent interest in summability theory. Then we present results concerning the optimal rate of BE bounds among triangular summability methods. The last section provides a link to approximation theory.

2. SOME SUMMABILITY RESULTS

Let $A = [a_{n,k}]$, $n, k \geq 0$, be an infinite matrix of complex numbers. We say that a sequence $x := (x_0, x_1, \dots)$ is in the domain of A if the series

$$(Ax)_n := \sum_{k=0}^{\infty} x_k a_{n,k}$$

is convergent for each $n = 0, 1, 2, \dots$. When x is in the domain of A , the transformed sequence $(Ax)_n$, $n = 0, 1, 2, \dots$, is denoted by (Ax) . If c denotes the set of all convergent complex sequences, then $c_A := A^{-1}(c)$ denotes the set of all sequences x such that (Ax) is convergent. We say that B includes A if $c_A \subseteq c_B$. If it obeys the strict inclusion, i.e., if $c_A \subset c_B$, then B is said to be stronger than A . Let I denote the identity matrix. We say A is regular if $c_I \subseteq c_A$ ([11]).

One of the most well-known regular summability methods is the Cesàro method, denoted by $(C, 1)$, in which $a_{n,k} = \frac{1}{n+1}$ for $k = 0, 1, \dots, n$, and zero otherwise. If we define $a_{n,k} = \binom{n}{k} r^k (1-r)^{n-k}$ for $0 \leq k \leq n$ and $n \geq 1$, $a_{0,0} = 1$, we get the Euler method, E_r , which is regular if and only if $r \in (0, 1]$ (cf. [15]). A large class of regular summability methods can be obtained by using two nonnegative integer valued random variables as follows.

Definition 2.1. Let U and V be two nonnegative integer valued random variables. Let V_1, V_2, \dots be mutually independent random variables identically distributed as V . Let $[p_{n,k}]$ be a matrix whose n -th row consists of the discrete probability density of the random variable $U + V_1 + V_2 + \dots + V_n$. We will call $[p_{n,k}]$ the *convolution method* generated by U and V .

It is easy to see that the Euler method E_r is a convolution method. When we take U and V as Poisson(1) random variables, the resulting convolution method is known as the Borel matrix method. By using U and V as geometric, or shifted geometric random variables, one gets the Taylor and Meyer-König summability methods. By using the Silverman-Teoplitz theorem one can easily show that the convolution method is regular if and only if $P(V = 0) < 1$.

Definition 2.2. We say $A = [a_{n,k}]$ is ℓ^2 -negligible if

$$\|a_n\|_2^2 := \sum_{k=0}^{\infty} |a_{n,k}|^2 \rightarrow 0, \quad \text{as } n \rightarrow \infty.$$

The ℓ^2 -negligibility concept is closely tied to the asymptotic negligibility concept that applies in the central limit theorem. It happens to be the case that most of the standard regular summability methods are ℓ^2 -negligible. For instance, the Euler method is ℓ^2 -negligible as long as the parameter $r \in (0, 1)$. We will show that the regular convolution methods (different from the identity matrix) are ℓ^2 -negligible. Further, we will show that any method which includes an ℓ^2 -negligible regular summability method is also ℓ^2 -negligible. The following proposition contains a result which, for the most part, is well known in probability literature (see for instance [5]). The last part is a relevant addition to it, using the Pringsheim double limit concept.

Proposition 2.1. *Let $T = [t_{n,k}]$ be a complex matrix. Then*

$$\lim_{n \rightarrow \infty} \sum_k |t_{n,k}|^2 = 0 \quad \Rightarrow \quad \lim_{n \rightarrow \infty} \max_k |t_{n,k}| = 0.$$

Also, if T satisfies the condition

$$(2.1) \quad \sum_k |t_{n,k}| \leq M < \infty, \quad \text{for all } n \geq 0,$$

then the following are equivalent:

- (1) $\lim_{n \rightarrow \infty} \max_k |t_{n,k}| = 0$.
- (2) T is ℓ^2 -negligible.

Furthermore, if $[t_{n,k}]$ satisfies the uniform boundedness condition (2.1) and its columns tend to zero, then the above two statements are equivalent to

- (3) $\lim_{n,k \rightarrow \infty} |t_{n,k}| = 0$,

where $\lim_{n,k}$ represents the Pringsheim double limit in the sense that for any $\epsilon > 0$ there exists an N so that $|t_{n,k}| < \epsilon$ for both $n, k > N$.

Proof. Since $\max_k |t_{n,k}|^2 \leq \sum_k |t_{n,k}|^2 \rightarrow 0$, the first part follows trivially. To show the converse after assuming (2.1), note that

$$\sum_k |t_{n,k}|^2 \leq \left(\max_k |t_{n,k}| \right) \sum_k |t_{n,k}| \leq \left(\max_k |t_{n,k}| \right) M \rightarrow 0.$$

Now to prove that (3) implies (1), let $\epsilon > 0$. Then by (3), there exists an N_1 such that $|t_{n,k}| < \epsilon$ for all $k, n \geq N_1$. Since the columns tend to zero, we can make the tail entries of the first N_1 columns less than ϵ for large n . That is, there is an N_2 such that $|t_{n,k}| < \epsilon$ for $0 \leq k < N_1$ for all $n \geq N_2$. Thus, $\max_k |t_{n,k}| < \epsilon$ for all $n \geq \max\{N_1, N_2\}$. To prove the converse, assume that (2) is true but (3) is false. Then, there exists an $\epsilon > 0$ and infinitely many n_j and $k_j = k(n_j)$ such that both are increasing to infinity and

$$|t_{n_j, k_j}| > \epsilon, \quad j = 1, 2, \dots$$

But then

$$\sum_k |t_{n_j, k}|^2 \geq |t_{n_j, k_j}|^2 > \epsilon^2$$

for infinitely many j contradicting (2). □

The following proposition is well known in probability literature when the matrix is row finite ([5], [9]). The proof remains similar; therefore, it is omitted.

Proposition 2.2. *Let $[t_{n,k}]$ be a complex matrix which satisfies the following conditions:*

- (1) $\sum_k |t_{n,k}| \leq M < \infty$, for all $n \geq 0$,
- (2) $\sum_k t_{n,k} \rightarrow 1$ as $n \rightarrow \infty$ and
- (3) the matrix $[t_{n,k}]$ is ℓ^2 -negligible.

Then for any complex number z we have

$$\lim_{n \rightarrow \infty} \prod_{k=1}^{\infty} (1 + t_{n,k}z) = e^z.$$

The following result shows that for two regular methods A and B , if B includes A and A is ℓ^2 -negligible, then so is B .

Theorem 2.3. *If $A = [a_{n,k}]$ and $B = [b_{n,k}]$ are regular matrices such that*

$$\lim_{n,k} a_{n,k} = 0 \quad \text{and} \quad \lim_{n,k} b_{n,k} \neq 0,$$

then B cannot include A .

Proof. We will use the classical sliding-hump argument for the proof. First note that since the rows of A are null sequences, the fact that $\lim_{n,k} a_{n,k} = 0$, we have

$$(2.2) \quad \lim_{k \rightarrow \infty} \max_n |a_{n,k}| = 0.$$

Also, since $\lim_{n,k} b_{n,k} \neq 0$, we can choose increasing sequences of row and column indices satisfying

$$|b_{\nu(m),\kappa(m)}| \geq \delta > 0, \quad \text{for all } m.$$

Then use (2.2) to choose a subsequence of these pairs $(\nu(m), \kappa(m))$ such that

$$\max_n |a_{n,\kappa(m)}| < 2^{-m}.$$

Next, the fact that the columns of A and B are null sequences, we choose a further subsequence so that for all m (i.e., for all $\kappa(m)$), $\nu(m)$ is chosen so that $k < \kappa(m)$, $n > \nu(m)$ implies

$$|a_{nk}| < 2^{-m} \quad \text{and} \quad |b_{nk}| < 2^{-m}.$$

Use the fact that the rows of B tend to zero in order to get $|b_{nk}| < 2^{-m}$ for all $k > \kappa(m)$ and $n < \nu(m)$. Define the sequence x by

$$(2.3) \quad x_k := \begin{cases} m+1 & \text{if } k = \kappa(m), \text{ for } m = 0, 1, \dots, \\ 0 & \text{if } k \neq \kappa(m), \text{ for } m = 0, 1, \dots. \end{cases}$$

This yields (for $n > \nu(m)$)

$$\begin{aligned} |(Ax)_n| &= \left| \sum_{j=0}^{\infty} a_{n,\kappa(j)} (j+1) \right| \\ &\leq \left| \sum_{j \leq m} 2^{-m} (j+1) \right| + \left| \sum_{j > m} 2^{-j} (j+1) \right| \\ &= 2^{-m} \frac{(m+1)(m+2)}{2} + R_m, \end{aligned}$$

where $R_m \rightarrow 0$. Therefore, (Ax) converges to zero. Also,

$$\begin{aligned}
 (Bx)_{\nu(m)} &= \sum_{j=0}^{\infty} b_{\nu(m),\kappa(j)} x_{k(j)} \\
 &= b_{\nu(m),\kappa(m)}(m+1) + \sum_{j \neq m} b_{\nu(m),\kappa(j)}(j+1) \\
 &= b_{\nu(m),\kappa(m)}(m+1) + \sum_{j \leq m} 2^{-m}(j+1) + \sum_{j > m} 2^{-j}(j+1) \\
 (2.4) \qquad &= b_{\nu(m),\kappa(m)}(m+1) + 2^{-m-1}m(m+1) + R_m.
 \end{aligned}$$

Since the first term in (2.4) is unbounded and the latter two terms tend to zero, it follows that (Bx) is not even bounded and hence could not be convergent. This shows that B cannot include A . \square

Remark 2.4. It is not necessary to assume the full strength of regularity to have the proof work. It would be sufficient to assume (2.2) instead of assuming that $\lim_{n,k} a_{n,k} = 0$ and that A and B have null columns and rows. Or, it would be sufficient to assume the existence of $\{\kappa(m)\}$ such that $\lim_k \max_m |a_{n,\kappa(m)}| = 0$, and for a corresponding $\{\nu(m)\}$ we have $|b_{\nu(m),\kappa(m)}| \not\rightarrow 0$.

3. BE BOUNDS FOR SUMMABILITY TRANSFORMS

In this section we present the summability analog of the BE bound and show that the Cesàro transform provides, in some sense, the best rate of convergence. By using Propositions 2.1 and 2.2 and the standard results of probability theory one can verify the following version of Lyapunov’s theorem.

Theorem 3.1. *Let $S_n = \sum_{k=0}^{\infty} X_{n,k}$ be a random series, where $X_{n,k}, k = 0, 1, 2, \dots$, are independent. Further, assume that $E|X_{n,k}|^3 = \gamma_{n,k} < \infty$, $E(X_{n,k}) = 0$, and $\sum_{k=0}^{\infty} \text{Var}(X_{n,k}) = 1$. If*

$$\Gamma_n := \|\gamma_n^{1/3}\|_3^3 = \sum_{k=0}^{\infty} \gamma_{n,k} \rightarrow 0, \quad \text{as } n \rightarrow \infty,$$

then S_n converge in distribution to the standard normal random variable.

We now present Berry-Esseen type bounds for S_n . The results are derived from the generalization of the classical Berry-Esseen theorem ([5], [4]) to particular summability methods such as the Abel method and the Zeta method ([6], [10], and [14]). The approach of this proof uses the technique of characteristic functions that is already available in the literature ([4] and [5]), however, with an appropriate modification to incorporate infinitely many terms in the sums. We will omit the details. The central limit theorem for summability methods considered in [7] and [12] deal primarily with non-uniform rates of convergence.

Theorem 3.2. *Let $S_n = \sum_{k=0}^{\infty} X_{n,k}$ be a random series, where $X_{n,k}, k = 0, 1, 2, \dots$, are independent. Further, assume that $E|X_{n,k}|^3 = \gamma_{n,k} < \infty$, $E(X_{n,k}) = 0$, and $\sum_{k=0}^{\infty} \text{Var}(X_{n,k}) = 1$. Then*

$$\left| P(S_n \leq x) - \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-u^2/2} du \right| \leq C\Gamma_n,$$

where C is an absolute constant and $\Gamma_n := \|\gamma_n^{1/3}\|_3^3 = \sum_{k=0}^{\infty} E(|X_{n,k}|^3)$.

We will now examine the bound, Γ_n , when

$$X_{n,k} := \frac{p_{n,k} Y_k}{\|\sigma p_n\|_2}, \quad E(Y_k) = 0, \quad \|\sigma p_n\|_2^2 := \sigma^2 \sum_{k=0}^{\infty} p_{n,k}^2,$$

and Y_0, Y_1, \dots are *iid* random variables with $\text{Var}(Y_0) = \sigma^2$. The aim is to discover which summability transform gives the best rate. The following theorem shows that the class of convolution summability methods cannot be the one that gives the best rate.

Theorem 3.3. *Let $[p_{n,k}]$ be the convolution summability method generated by U and V . If U and V have finite variances, then, $\Gamma_n = O(n^{-1/4})$ as $n \rightarrow \infty$. Furthermore, this rate cannot be improved.*

Proof. Since

$$\|p_n\|_3 \leq \left(\max_k |p_{n,k}| \sum_k |p_{n,k}|^2 \right)^{1/3},$$

we have $\|p_n\|_3 \leq \|p_n\|_{\infty}^{1/3} \|p_n\|_2^{2/3}$. This gives that

$$\frac{\|p_n\|_3}{\|p_n\|_2} \leq \frac{\|p_n\|_{\infty}^{1/3} \|p_n\|_2^{2/3}}{\|p_n\|_2} = \frac{\|p_n\|_{\infty}^{1/3}}{\|p_n\|_2^{1/3}}.$$

Let $S_n = U + V_1 + V_2 + \dots + V_n$ and let S'_n be independent and identically distributed as S_n . Note that

$$\sum_k p_{n,k}^2 = \sum_k P(S_n = k) P(S'_n = k) = P(S_n = S'_n).$$

Let $Z_i := V_i - V'_i$, $i = 1, 2, \dots, n$, where V'_i are independent and identically distributed as V_i . If $\text{Var}(V_1) = \sigma^2$, we have (see, for instance, [5])

$$n^{1/2} P\left(\sum_{k=1}^n Z_k = j\right) \rightarrow \frac{1}{2\sigma\sqrt{\pi}}, \quad p_{n,k} = P(S_n = k) \leq \frac{C}{\sqrt{n}}$$

for any integer j , for all k and some absolute constant C . Hence,

$$\begin{aligned} n^{1/2} P(S_n - S'_n = 0) &= n^{1/2} \sum_{j=-\infty}^{\infty} q_j P\left(\sum_{k=1}^n Z_k = j\right) \\ &\rightarrow \sum_{j=-\infty}^{\infty} q_j \frac{1}{2\sigma\sqrt{\pi}}, \end{aligned}$$

where $\{q_j\}$ is the probability density of $U - U'$. Therefore, $[p_{n,k}]$ is ℓ^2 -negligible, and

$$n^{1/4} \|p_n\|_2 \rightarrow \frac{1}{\sqrt{2\sigma\pi^{1/4}}}.$$

We have

$$\frac{\|p_n\|_3}{\|p_n\|_2} \leq \frac{\|p_n\|_{\infty}^{1/3}}{\|p_n\|_2^{1/3}} \leq C n^{-1/12},$$

for some constant $C > 0$. This gives the required assertion.

To prove that the rate is best possible, take $[p_{n,k}]$ to be the Euler method. If $|k - nr|/\sqrt{nr(1-r)} < K$, then there exists a constant $D > 0$ (which depends only on r and K) such that

$$\left| \binom{n}{k} r^k (1-r)^{n-k} - \frac{1}{\sqrt{2\pi r(1-r)n}} \exp\left\{-\frac{(k - nr)^2}{2nr(1-r)}\right\} \right| \leq \frac{D}{n}.$$

Thus, if $A_{n,K} := \{k : |k - nr|/\sqrt{nr(1-r)} < K\}$ is the set of all non-negative integers k for which the stated inequality holds, we have

$$\begin{aligned} \sum_{k=0}^{\infty} p_{n,k}^3 &\geq \sum_{k \in A_{n,K}} p_{n,k}^3 \\ &\geq \min_{k \in A_{n,K}} p_{n,k}^2 \sum_{k \in A_{n,K}} p_{n,k} \\ &\geq \left\{ \frac{e^{-2K^2}}{(2\pi r(1-r)n)^{2/2}} + O\left(\frac{1}{n^{1.5}}\right) \right\} \sum_{k \in A_{n,K}} p_{n,k}. \end{aligned}$$

By the central limit theorem,

$$\sum_{k \in A_{n,K}} p_{n,k} = P((S_n - nr)/\sqrt{nr(1-r)} < K) \rightarrow P(|Z| < K),$$

where Z is the standard normal random variable. Thus, for any $K > 0$, we have

$$\liminf_n n \sum_{k=0}^{\infty} p_{n,k}^3 \geq \left\{ \frac{e^{-2K^2}}{(2\pi r(1-r))} \right\} P(|Z| < K) > 0.$$

That is, there exists a $\delta > 0$ such that

$$\liminf_n n^{1/3} \|p_n\|_3 \geq \delta > 0.$$

Since

$$\lim_n n^{1/4} \|p_n\|_2 = \frac{1}{\pi^{1/4} \sqrt{2r(1-r)}},$$

we see that

$$\liminf_n n^{1/12} \frac{\|p_n\|_3}{\|p_n\|_2} = \liminf_n \frac{n^{1/3} \|p_n\|_3}{n^{1/4} \|p_n\|_2} \geq C > 0.$$

This proves the theorem. □

In the statistical literature the concept of a minimum variance unbiased estimator is used to identify a best possible estimator (among the class of all unbiased estimators). Since the variance usually goes to zero as the sample size gets large, one only compares the variances of competing unbiased estimators that rely upon the *same* sample size. In a similar context, the comparison of rates of convergence in the central limit theorem will make sense if the competing summability methods use the same “sample size”. Hence, if we identify $X_0, X_1, X_2, \dots, X_{n-1}$ as our sample of size n , then the n -th terms of the competing transforms could be compared if the summability methods are forced to be triangular. In the following theorem we therefore assume that the summability methods are triangular.

The Silverman-Toeplitz theorem states that if the method is regular, then its row sums (even though they may not equal one for any row) must, in the limit,

become one. To avoid the “contamination” in the BE rate that is caused by the row sums not being equal to one, but converging to one, we assume that each row of the summability method adds up to one.

Theorem 3.4. *For the case of iid sequences of random variables, over the class of all triangular real regular summability methods (with rows adding up to one) the rate of convergence, Γ_n , in the Berry-Esseen bound has the fastest rate of decrease when we use the Cesáro summability method.*

Proof. Let $r, s > 0$ so that $\frac{1}{r} + \frac{1}{s} = 1$. For any numbers a_k

$$\begin{aligned} \sum_{k=1}^n |a_k|^p &= \sum_{k=1}^n |a_k|^p \cdot 1 \\ &\leq \left\{ \sum_{k=1}^n |a_k|^{pr} \right\}^{p/(pr)} \cdot n^{1/s} = \|a\|_{pr}^p \cdot n^{1/s}. \end{aligned}$$

This gives that $\|a\|_p^p \leq \|a\|_{pr}^p \cdot n^{\frac{r-1}{r}}$ or $\|a\|_p \leq \|a\|_{pr} \cdot n^{\frac{r-1}{pr}}$. The equality holds if and only if the vector a and the constant vector 1 are parallel. For $p = 2$, and $r = \frac{3}{2}$ we get

$$\|a\|_2 \leq \|a\|_3 \cdot n^{\frac{1}{6}}.$$

In the Hölder inequality the equality takes place if and only if the vectors are parallel. That is, the vector a is a constant vector. Since the rows have to add up to one, the constant vector must be $(1/n, 1/n, \dots, 1/n)$. \square

4. SCHNABL APPROXIMATION OPERATORS

Let X_0, X_1, \dots be a sequence of independent and identically distributed random variables taking values in an interval I . We will assume that the third moment exists and denote $E(X_0) = x$. Let $A = [a_{nk}]$ be a regular summability method with rows adding up to one. The Schnabl approximation operator, [1], is defined by

$$L_n(f, x) := Ef((AX)_n), \quad n = 1, 2, \dots$$

Usually A is assumed to be triangular. Let ω be a modulus of continuity defined over $[0, \infty)$ i.e., $\omega(t)$ is a non-decreasing continuous function with $\omega(0^+) = \omega(0) = 0$ and $\omega(t+s) \leq \omega(t) + \omega(s)$. The space $H^\omega(I)$ consists of those functions $f : I \rightarrow \mathfrak{R}$ so that $|f(u) - f(v)| \leq \omega(|u - v|)$, and is called the Lipschitz space generated by ω . The following result quantifies the rate of convergence of Schnabl operators.

Theorem 4.1. *For the above Schnabl operators, we have*

- (1) *For any $f \in H^\omega(I)$, $|L_n(f, x) - f(x)| = O(\omega(\|a_n\|_2))$.*
- (2) *If $\text{Var}(X_0) > 0$ and $\|a_n\|_3 = o(\|a_n\|_2)$, then the rate cannot be improved in (1).*

Proof. We will use the fact that if ω is a modulus of continuity, then there exists a concave modulus of continuity ω^* so that $\omega \leq \omega^* \leq 2\omega$. When $f \in H^\omega(I)$, we have

$$\begin{aligned} |L_n(f, x) - f(x)| &\leq E|f((AX)_n) - f(x)| \\ &\leq E\omega(|(AX)_n - x|) \\ &\leq 2\omega(E|(AX)_n - x|) \\ &\leq 2\omega(\|a_n\|_2) \left(1 + E \frac{|(AX)_n - x|}{\|a_n\|_2} \right). \end{aligned}$$

The last expression remains bounded. To prove the reverse inequality, take $f(t) = \omega(|t - x|)$ which belongs to $H^\omega(I)$. Since

$$|f(t)| \leq (|t| + 1)\omega(1) + \omega(|x|),$$

we see that $L_n(f, x)$ is well defined. Next let $Z_{n,x} := ((AX)_n - x)/\|a_n\|_2$. Note that

$$\begin{aligned} |L_n(f, x) - f(x)| &= E\omega(\|a_n\|_2|Z_{n,x}|) \\ &\geq \omega(\|a_n\|_2)E\left(\frac{|Z_{n,x}|}{1 + |Z_{n,x}|}\right) \\ &\geq c\omega(\|a_n\|_2), \end{aligned}$$

for a $c > 0$ and all large n , since when $\|a_n\|_3 = o(\|a_n\|_2)$,

$$E\left(\frac{|Z_{n,x}|}{1 + |Z_{n,x}|}\right) \rightarrow E\left(\frac{|Z|}{1 + |Z|}\right) > 0, \quad Z \sim N(0, 1).$$

□

Among the regular triangular methods that have row sums one, Theorem 4.1 suggests that the best rate of convergence for the Schnabl operators is achieved if we use the Cesaro summability method. This class contains the classical approximation operators such as Bernstein, Szasz, Baskakov, Gamma and Weierstrass operators. For each one of them, the rate is $\omega(n^{-1/2})$ ([1]). On the other hand, when A is taken to be the Euler method, the resulting Schnabl operators inherit the rate $\omega(n^{-1/4})$.

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