Research Article

Equivalent Conditions of Complete p th Moment Convergence for Weighted Sums of I. I. D. Random Variables under Sublinear Expectations

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We investigate the complete p th moment convergence for weighted sums of independent, identically distributed random variables under sublinear expectations space. Using moment inequality and truncation methods, we prove the equivalent conditions of complete p th moment convergence of weighted sums of independent, identically distributed random variables under sublinear expectations space, which complement the corresponding results obtained in Guo and Shan (2020).

1. Introduction

Peng [1, 2] presented the concept of the sublinear expectation space to study the uncertainty of probability and distribution. The seminal work of Peng [1, 2] encourages people to study limit theorems under sublinear expectations space. Zhang [3–5] proved results including exponential inequalities, Rosenthal's inequalities, and Donsker's invariance principle under sublinear expectations. Wu [6] obtained precise asymptotics for complete integral convergence. Xu and Cheng [7] studied precise asymptotics in the law of iterated logarithm under sublinear expectations. The interested reader could refer to Xu and Zhang [8, 9], Chen [10], Gao and Xu [11], Fang et al. [12], Hu et al. [13], Hu and Yang [14], Huang and Wu [15], Kucmaszewska [16], Ma and Wu [17], Wang and Wu [18], Wu and Jiang [19], Yu and Wu [20], Zhang [21], Zhong and Wu [22], and references therein for more limit theorems under sublinear expectations.

Recently, Guo and Shan [23] studied equivalent conditions of complete q th moment convergence for weighted sums of sequences of negatively orthant dependent random variables. Xu and Cheng [24] obtained equivalent conditions of complete convergence for weighted sums of sequences of

i. i. d. random variables under sublinear expectations. Motivated by the work of Guo and Shan [23] and Xu and Cheng [24], here we try to prove the equivalent conditions of complete p th moment convergence of weighted sums of independent, identically distributed random variables under sublinear expectations space, which complement the corresponding results obtained in Guo and Shan [23], also extend results in Xu and Cheng [24] from complete convergence to complete p th moment convergence.

We organized the rest of this paper as follows. In Section 2, we give necessary basic notions, concepts, and relevant properties and present necessary lemmas under sublinear expectations. In Section 3, we give our main results, Theorems 1–4, whose proofs are presented in Section 4.

2. Preliminaries

As in Xu and Cheng [24], we adopt similar notations as in the work by Peng [2] and Chen [10]. Suppose that $(\Omega, \mathcal{F})$ is a given measurable space. We assume that $\mathcal{H}$ is a subset of all random variables on $(\Omega, \mathcal{F})$ such that $I_A \in \mathcal{H}$ (cf. [10]), where $A \in \mathcal{F}$, and $X_1, \dots, X_n \in \mathcal{H}$ implies $\varphi(X_1, \dots, X_n) \in \mathcal{H}$, for each $\varphi \in \mathcal{C}_{l, \text{Lip}}(\mathbb{R}^n)$, where $\mathcal{C}_{l, \text{Lip}}(\mathbb{R}^n)$ denotes the linear space of (local Lipschitz) function φ satisfying

$$|\varphi(\mathbf{x}) - \varphi(\mathbf{y})| \leq C(1 + |\mathbf{x}|^m + |\mathbf{y}|^m)(|\mathbf{x} - \mathbf{y}|), \quad \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n, \quad (1)$$

for some $C > 0$ and $m \in \mathbb{N}$ depending on φ .

Definition 1. A sublinear expectation $\mathbb{E}$ on $\mathcal{H}$ is a functional $\mathbb{E}: \mathcal{H} \rightarrow \overline{\mathbb{R}} := [-\infty, \infty]$ satisfying the following properties: for all $X, Y \in \mathcal{H}$, we have

- (a) Monotonicity: if $X \geq Y$, then $\mathbb{E}[X] \geq \mathbb{E}[Y]$
- (b) Constant preserving: $\mathbb{E}[c] = c, \forall c \in \mathbb{R}$
- (c) Positive homogeneity: $\mathbb{E}[\lambda X] = \lambda \mathbb{E}[X], \forall \lambda \geq 0$
- (d) Subadditivity: $\mathbb{E}[X + Y] \leq \mathbb{E}[X] + \mathbb{E}[Y]$ whenever $\mathbb{E}[X] + \mathbb{E}[Y]$ is not of the form $\infty - \infty$ or $-\infty + \infty$

A set function $V: \mathcal{F} \rightarrow [0, 1]$ is said to be a capacity if it obeys

- (a) $V(\emptyset) = 0$ and $V(\Omega) = 1$.
- (b) $V(A) \leq V(B)$, $A \subset B$ and $A, B \in \mathcal{F}$.

Moreover, if V is continuous, then V should satisfy

- (c) $V(A_n) \uparrow V(A)$ if $A_n \uparrow A$.
- (d) $V(A_n) \downarrow V(A)$ if $A_n \downarrow A$.

A capacity V is called subadditive if $V(A + B) \leq V(A) + V(B)$, $A, B \in \mathcal{F}$.

In this paper, given a sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$, set $\mathbb{V}(A) := \inf\{\mathbb{E}[\xi]: I_A \leq \xi, \xi \in \mathcal{H}\} = \mathbb{E}[I_A]$, $\forall A \in \mathcal{F}$ (see (8) and the definitions of $\mathbb{V}$ above (8) in [4]). Clearly, $\mathbb{V}$ is a subadditive capacity. Denote the Choquet expectations $C_{\mathbb{V}}$ by

$$C_{\mathbb{V}}(X) := \int_0^\infty \mathbb{V}(X > x) dx + \int_{-\infty}^0 (\mathbb{V}(X > x) - 1) dx. \quad (2)$$

Suppose that $\mathbf{X} = (X_1, \dots, X_m)$, $X_i \in \mathcal{H}$, and $\mathbf{Y} = (Y_1, \dots, Y_n)$, $Y_i \in \mathcal{H}$, are two random vectors on $(\Omega, \mathcal{H}, \mathbb{E})$. $\mathbf{Y}$ is called to be independent of $\mathbf{X}$ if, for each Borel-measurable function ψ on $\mathbb{R}^m \times \mathbb{R}^n$ with $\psi(\mathbf{X}, \mathbf{Y}), \psi(\mathbf{x}, \mathbf{Y}) \in \mathcal{H}$; for each $\mathbf{x} \in \mathbb{R}^m$, we have $\mathbb{E}[\psi(\mathbf{X}, \mathbf{Y})] = \mathbb{E}[\mathbb{E}[\psi(\mathbf{x}, \mathbf{Y})|_{\mathbf{x}=\mathbf{X}}]]$ whenever $\bar{\psi}(\mathbf{x}) := \mathbb{E}[\psi(\mathbf{x}, \mathbf{Y})] < \infty$ for each $\mathbf{x}$ and $\mathbb{E}[\bar{\psi}(\mathbf{X})] < \infty$ (see Definition 2.5 in [10]). $\{X_n\}_{n=1}^\infty$ is called a sequence of independent random variables if X_{n+1} is independent of $(X_1, \dots, X_n)$, for each $n \geq 1$.

Assume that $\mathbf{X}_1$ and $\mathbf{X}_2$ are two n -dimensional random vectors defined, respectively, in sublinear expectation spaces $(\Omega_1, \mathcal{H}_1, \mathbb{E}_1)$ and $(\Omega_2, \mathcal{H}_2, \mathbb{E}_2)$. They are called identically distributed if, for every Borel-measurable function ψ such that $\psi(\mathbf{X}_1) \in \mathcal{H}_1, \psi(\mathbf{X}_2) \in \mathcal{H}_2$,

$$\mathbb{E}_1[\psi(\mathbf{X}_1)] = \mathbb{E}_2[\psi(\mathbf{X}_2)], \quad (3)$$

whenever the sublinear expectations are finite. $\{X_n\}_{n=1}^\infty$ is called to be identically distributed if, for each $i \geq 1$, X_i and X_1 are identically distributed.

In the sequel, we suppose that $\mathbb{E}$ is countably subadditive, i.e., $\mathbb{E}(X) \leq \sum_{n=1}^\infty \mathbb{E}(X_n)$, whenever $X \leq \sum_{n=1}^\infty X_n$, $X, X_n \in \mathcal{H}$, and $X \geq 0, X_n \geq 0, n = 1, 2, \dots$. Let C denote a positive

constant which may differ from line to line. $I(A)$ or I_A represents the indicator function of A , $a_n \ll b_n$ means that there exists a constant $C > 0$ such that $a_n \leq Cb_n$ for n large sufficiently, and $a_n \approx b_n$ means that $a_n \ll b_n$ and $b_n \ll a_n$. We use $\log x$ for $\ln \max\{e, x\}$.

We first present several necessary lemmas to prove our main results. By using Corollary 2.2, Theorem 2.3 in [5], the proofs of Theorem 3 in [25], and Minkowski's inequality under sublinear expectations, we see that the following lemma holds.

Lemma 1 (cf. Lemma 2.3 in [24]). Suppose that $\{X_i, 1 \leq i \leq n\}$ is a sequence of independent random variables under sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$ with $\mathbb{E}[X_i] \leq 0$, $\mathbb{E}|X_i|^M < \infty, 1 \leq i \leq n$, and $M \geq 2$. Then,

$$\mathbb{E} \max_{1 \leq j \leq n} \left| \sum_{i=1}^j X_i \right|^M \leq C \log^M n \left(\sum_{i=1}^n \mathbb{E}|X_i|^M + \left(\sum_{i=1}^n \mathbb{E}|X_i|^2 \right)^{M/2} \right), \quad (4)$$

where C depends on M only.

Proof. For readers' convenience, we give complete proofs here. First, for $b \geq 0$ and $n \geq 1$, set $S_{b,n} = \sum_{k=b+1}^{b+n} X_i$ and $L_{b,n} = \max_{1 \leq k \leq n} |S_{b,k}|$, and

$$g(F_{b,n}) = C_M \left(\sum_{i=b+1}^{b+n} \mathbb{E}|X_i|^M + \left(\sum_{i=b+1}^{b+n} \mathbb{E}|X_i|^2 \right)^{M/2} \right), \quad (5)$$

where C_M depends on M determined as in (2.8) of Corollary 2.2 (b) in [5]. By Corollary 2.2 in [5], we know that, for all $b \geq 0$ and $n \geq 1$,

$$\mathbb{E}(|S_{b,n}|^M) \leq g(F_{b,n}). \quad (6)$$

Obviously, for $b \geq 0$ and $1 \leq k \leq k+l$,

$$g(F_{b,k}) + g(F_{b+k,l}) \leq g(F_{b,k+l}). \quad (7)$$

Set $\Lambda(1) = 1$, and for $n \geq 2$, $\Lambda(n) = 1 + \Lambda(m-1)$, where m is the integer part of $(1/2)(n+2)$. $1 + \log(2(m-1)) \leq \log(2n)$ implies that $\Lambda(n) \leq \log(2n)$. Now, it is enough to prove that, for $b \geq 0$ and $n \geq 1$,

$$\mathbb{E}(L_{b,n}^M) \leq (\Lambda(n))^M g(F_{b,n}). \quad (8)$$

As in the proof of Theorem 4 of Móricz [25], let $n > 1$ be given. Obviously, $n = 2m-1$ or $2m-2$. For $b \geq 0$ and $m \leq k \leq n$, we see that

$$|S_{b,k}| \leq |S_{b,m}| + |S_{b+m,k-m}|, \quad (9)$$

whence, for such k 's,

$$|S_{b,k}| \leq |S_{b,m}| + L_{b+m,n-m}. \quad (10)$$

Since, for $1 \leq k < m$, we have $|S_{b,k}| \leq L_{b,m-1}$, hence, for $1 \leq k \leq n$,

$$|S_{b,k}| \leq |S_{b,m}| + (L_{b,m-1}^M + L_{b+m,n-m}^M)^{1/M}. \quad (11)$$

Thus,

$$L_{b,n} \leq |S_{b,n}| + (L_{b,m-1}^M + L_{b+m,n-m}^M)^{1/M}, \quad (12)$$

and by Minkowski's inequality under sublinear expectations (see (4.10) in Proposition 4.2 of Chapter I of [2]),

$$(\mathbb{E}(L_{b,n}^M))^{1/M} \leq (\mathbb{E}(|S_{b,n}|))^{1/M} + (\mathbb{E}(L_{b,m-1}^M) + \mathbb{E}(L_{b+m,n-m}^M))^{1/M}. \quad (13)$$

Suppose now that (8) holds for $k < n$. Then, by the choice of m , we see that

$$\begin{aligned} \mathbb{E}(L_{b,m-1}^M) &\leq \Lambda^M (m-1)g(F_{b,m-1}), \\ \mathbb{E}(L_{b+m,n-m}^M) &\leq \Lambda^M (n-m)g(F_{b+m,n-m}) \\ &\leq \Lambda^M (m-1)g(F_{b+m,n-m}). \end{aligned} \quad (14)$$

By these two inequalities above and (7), we see that

$$\mathbb{E}(L_{b,m-1}^M) + \mathbb{E}(L_{b+m,n-m}^M) \leq \Lambda^M (m-1)g(F_{b,n}). \quad (15)$$

Finally, (6) implies

$$\mathbb{E}(|S_{b,n}|^M) \leq g(F_{b,n}) \leq g(F_{b,n}). \quad (16)$$

By (13)–(16), we conclude that

$$(\mathbb{E}(L_{b,n}^M))^{1/M} \leq (1 + \Lambda(m-1))g^{1/M}(F_{b,n}) = \Lambda(n)g^{1/M}(F_{b,n}), \quad (17)$$

which implies the result. Hence, by (6), the conclusion of (8) is true for $n = 1$. By induction, (8) holds for all $n = 1, 2, \dots$. The proof is complete. $\square$

Lemma 2 (see Lemma 2.4 in [24, 26]). *Let $\{X_n; n \geq 1\}$ be a sequence of independent random variables under sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$. Then, for all $n \geq 1$ and $x > 0$,*

$$\left[1 - \mathbb{V}\left(\max_{1 \leq j \leq n} |X_j| > x\right)\right]^2 \sum_{j=1}^n \mathbb{V}(|X_j| > x) \leq 4\mathbb{V}\left(\max_{1 \leq j \leq n} |X_j| > x\right). \quad (18)$$

Remark 1. In the proofs of Lemma 2.4 in [24, 26], by independence of $I(A_k) := I(|X_k| > x)$, $k = 1, \dots, n$, $\mathbb{E}(X + c) = \mathbb{E}(c + X) = \mathbb{E}(X) + c$ for constant c and $X \in \mathcal{H}$ (Definition 2.5 in [10]), we see that

$$\begin{aligned} \sum_{k=1}^n \mathbb{E}[I(A_k)] &= \sum_{k=1}^{n-2} \mathbb{E}[I(A_k)] + \mathbb{E}[I(A_{n-1}) + \mathbb{E}[I(A_n)]] \\ &= \sum_{k=1}^{n-2} \mathbb{E}[I(A_k)] + \mathbb{E}\left[\left[x + \mathbb{E}[I(A_n)]\right] \middle|_{x=I(A_{n-1})}\right] = \sum_{k=1}^{n-2} \mathbb{E}[I(A_k)] + \mathbb{E}\left[\mathbb{E}[x + I(A_n)] \middle|_{x=I(A_{n-1})}\right] \\ &= \sum_{k=1}^{n-2} \mathbb{E}[I(A_k)] + \mathbb{E}[I(A_{n-1}) + I(A_n)] = \dots = \mathbb{E}\left[I(A_1) + \mathbb{E}\left[\sum_{k=2}^n I(A_k)\right]\right] \\ &= \mathbb{E}\left[\sum_{k=1}^n I(A_k)\right], \end{aligned} \quad (19)$$

which implies that Lemma 2 is valid. The difference between the sublinear expectations and linear expectations could be implied by the subtle observation that

$$\begin{aligned} \mathbb{E}[I(A_1) - I(A_2)] &= \mathbb{E}\left[\mathbb{E}[x - I(A_2)] \middle|_{x=I(A_1)}\right] \\ &= \mathbb{E}\left[\left[x + \mathbb{E}[-I(A_2)]\right] \middle|_{x=I(A_1)}\right] = \mathbb{E}[I(A_1)] + \mathbb{E}[-I(A_2)], \end{aligned} \quad (20)$$

which is not necessarily equal to $\mathbb{E}[I(A_1)] - \mathbb{E}[I(A_2)]$.

Lemma 3. *Let X be a random variable under sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$, $q > 0$, $r > 0$, and $p > 0$. Then, the following are equivalent:*

$$(i) \quad \begin{cases} C_{\mathbb{V}}(|X|^p) < \infty, & \text{for } p > \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q} \ln(|X|)) < \infty, & \text{for } p = \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q}) < \infty, & \text{for } p < \frac{r}{q}. \end{cases} \quad (21)$$

(ii)

$$\int_1^\infty dy \int_1^\infty y^{r-1} \mathbb{V}(|X| > x^{1/p} y^q) dx < \infty. \quad (22)$$

Proof

$$\begin{aligned} \int_1^\infty dy \int_1^\infty y^{r-1} \mathbb{V}(|X| > x^{1/p} y^q) dx &= \int_1^\infty dt \int_{t^q}^\infty t^{r-1} \mathbb{V}(|X| > s) s^{p-1} t^{-qp} ds \quad (\text{Settings} = x^{1/p} y^q, t = y) \\ &\approx \begin{cases} \int_1^\infty \mathbb{V}(|X| > s) s^{p-1} ds \approx \int_1^\infty \mathbb{V}(|X| > s) s^{p-1} ds \approx C_{\mathbb{V}}(|X|^p), & \text{for } p > \frac{r}{q}, \\ \int_1^\infty \mathbb{V}(|X| > s) s^{r/q-1} \ln(s) ds \approx C_{\mathbb{V}}(|X|^{r/q} \ln(|X|)), & \text{for } p = \frac{r}{q}, \\ \int_1^\infty \mathbb{V}(|X| > s) s^{p-1+(r-pq/q)} ds \approx C_{\mathbb{V}}(|X|^{r/q}), & \text{for } p < \frac{r}{q}. \end{cases} \end{aligned} \quad (23)$$

The proof is finished. $\square$

(ii)

Lemma 4. Let X be a random variable under sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$, $q > 0$, $r > 0$, and $p > 0$. Then, the following is equivalent:

$$\int_1^\infty dy \int_1^\infty y^{r-1} \ln(y) \mathbb{V}(|X| > x^{1/p} y^q) dx < \infty. \quad (25)$$

(i)

Proof

$$\begin{cases} C_{\mathbb{V}}(|X|^p) < \infty, & \text{for } p > \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q} \ln^2 |X|) < \infty, & \text{for } p = \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q} \ln |X|) < \infty, & \text{for } p < \frac{r}{q}. \end{cases} \quad (24)$$

$$\begin{aligned} \int_1^\infty dy \int_1^\infty y^{r-1} \ln(y) \mathbb{V}(|X| > x^{1/p} y^q) dx &\approx \int_1^\infty dt \int_{t^q}^\infty t^{r-1} \ln(t) \mathbb{V}(|X| > s) s^{p-1} t^{-qp} ds \quad (\text{Settings} = x^{1/p} y^q, t = y) \\ &\approx \int_1^\infty \mathbb{V}(|X| > s) s^{p-1} ds \int_1^{s^{1/q}} t^{r-1-qp} \ln(t) dt \\ &\approx \begin{cases} \int_1^\infty \mathbb{V}(|X| > s) s^{p-1} ds \approx C_{\mathbb{V}}(|X|^p), & \text{for } p > \frac{r}{q}, \\ \int_1^\infty \mathbb{V}(|X| > s) s^{r/q-1} \ln^2(s) ds \approx C_{\mathbb{V}}(|X|^{r/q} \ln^2 |X|), & \text{for } p = \frac{r}{q}, \\ \int_1^\infty \mathbb{V}(|X| > s) s^{p-1+(r-pq/q)} \ln(s) ds \approx C_{\mathbb{V}}(|X|^{r/q} \ln |X|), & \text{for } p < \frac{r}{q}. \end{cases} \end{aligned} \quad (26)$$

This finishes the proof.

3. Main Results

We state our main results, the proofs of which will be given in Section 4.

Theorem 1. Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, identically distributed as X under sub-linear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$. Assume that $r > 1$, $q > (1/2)$, and $\beta > -q/r$, and suppose that $\mathbb{E}X = -\mathbb{E}(-X) = 0$ for $(1/2) < q \leq 1$. Suppose that $\{a_{ni} \approx (i/n)^\beta (1/n^q), 1 \leq i \leq n, n \geq 1\}$ is a triangular array of real numbers. Then, the following is equivalent:

(i)

$$\begin{cases} C_{\mathbb{V}}(|X|^p) < \infty, & \text{for } p > \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q} \ln |X|) < \infty, & \text{for } p = \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q}) < \infty, & \text{for } p < \frac{r}{q}. \end{cases} \quad (27)$$

(ii)

$$\sum_{n=1}^{\infty} n^{r-2} C_{\mathbb{V}} \left(\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p - \varepsilon \right)^+ \right) < \infty, \quad \forall \varepsilon > 0. \quad (28)$$

Remark 2. If $(\Omega, \mathcal{H}, \mathbb{E})$ is classic probability space, then Theorem 1 recovers Theorem 7 in [23] in case in which $\{X_n, n \geq 1\}$ is a sequence of independent random variables, identically distributed as X . As pointed in Hossein and Nezakati [27], why we need p th moment convergence under sublinear expectations, the complete moment convergence is a more general expression than complete convergence in theory and practice, and the interested reader also could refer to the first study in complete moment convergence by Chow [28].

Remark 3. Under the same conditions in Theorem 1, if (28) holds, by the same proof of Remark 3.2 of Hossein and Nezakati [27], we see that, for all $\varepsilon > 0$,

$$\sum_{n=1}^{\infty} n^{r-2} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p > \varepsilon \right) < \infty. \quad (29)$$

Moreover, if $\mathbb{V}$ is continuous, by (29), we see that

$$\sum_{i=1}^n a_{ni} X_i \longrightarrow 0, \quad \text{a.s. } \mathbb{V}, \text{ as } n \longrightarrow \infty. \quad (30)$$

In (30), if $(1/2) < q \leq 1$, $\beta = 0$, we conclude that $\sum_{i=1}^n X_i/n^q \longrightarrow 0$ a.s. $\mathbb{V}$, as $n \longrightarrow \infty$, which is similar to Theorem 1 of Zhang and Lin [29].

Theorem 2. Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, identically distributed as X under sub-linear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$. Assume that $r > 1$, $q > (1/2)$, and $\beta = -q/r < 0$, and suppose that $\mathbb{E}X = -\mathbb{E}(-X) = 0$ for $(1/2) < q \leq 1$. Suppose that $\{a_{ni} \approx (i/n)^\beta (1/n^q), 1 \leq i \leq n, n \geq 1\}$ is a triangular array of real numbers. Then, (28) is equivalent to

$$\begin{cases} C_{\mathbb{V}}(|X|^p) < \infty, & \text{for } p > \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q} \ln^2 |X|) < \infty, & \text{for } p = \frac{r}{q}, \\ C_{\mathbb{V}}(|X|^{r/q} \ln |X|) < \infty, & \text{for } p < \frac{r}{q}. \end{cases} \quad (31)$$

Theorem 3. Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, identically distributed as X under sub-linear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$. Assume that $r > 1$, $q > (1/2)$, and $-q < \beta < -q/r < 0$, and suppose that $\mathbb{E}X = -\mathbb{E}(-X) = 0$ for $(1/2) < q \leq 1$. Suppose that $\{a_{ni} \approx (i/n)^\beta (1/n^q), 1 \leq i \leq n, n \geq 1\}$ is a triangular array of real numbers. Then, (28) is equivalent to

$$\begin{cases} C_{\mathbb{V}}(|X|^p) < \infty, & \text{for } p > \frac{(r-1)}{(q+\beta)}, \\ C_{\mathbb{V}}(|X|^{(r-1)/(q+\beta)} \ln |X|) < \infty, & \text{for } p = \frac{(r-1)}{(q+\beta)}, \\ C_{\mathbb{V}}(|X|^{(r-1)/(q+\beta)}) < \infty, & \text{for } p < \frac{(r-1)}{(q+\beta)}. \end{cases} \quad (32)$$

As in the proofs of Theorems 1–3, we can get the following corollary.

Corollary 1. Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, identically distributed as X under sub-linear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$. Assume that $r > 1$, $q > (1/2)$, and $\beta > -q$, and suppose that $\mathbb{E}X = -\mathbb{E}(-X) = 0$, for $(1/2) < q \leq 1$. Suppose that $\{a_{ni} \approx ((n-i)/n)^\beta (1/n^q), 0 \leq i \leq n-1, n \geq 1\}$ is a triangular array of real numbers. Then,

(i) (27) is equivalent to

$$\sum_{n=1}^{\infty} n^{r-2} \mathbb{E} \left(\left(\max_{0 \leq k \leq n-1} \left| \sum_{i=0}^k a_{ni} X_i \right|^p - \varepsilon \right)^+ \right) < \infty, \quad \beta > \frac{-q}{r}. \quad (33)$$

(ii) (31) is equivalent to (33) when $\beta = -q/r$.

(iii) (32) is equivalent to (33) when $-q < \beta < -q/r$.

The following theorem is complete p th moment convergence on Cesàro summation of independent, identically distributed random variables under $(\Omega, \mathcal{H}, \mathbb{E})$.

Theorem 4. Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, identically distributed as X under sub-linear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$. Assume that $r > 1$,

$q > (1/2)$, $0 < \alpha \leq 1$, and $p > 0$, and suppose that $\mathbb{E}X = -\mathbb{E}(-X) = 0$, for $(1/2) < q \leq 1$. Suppose $A_n^\alpha = [(\alpha + 1)(\alpha + 2), \dots, (\alpha + n)]/n!$, $n = 1, 2, \dots$ and $A_0^\alpha = 1$. Then,

(i) (27) is equivalent to

$$\sum_{n=1}^{\infty} n^{r-2} (A_n^\alpha)^{-p} \mathbb{E} \left(\left(\max_{0 \leq k \leq n-1} \left| \sum_{i=0}^k A_{n-i}^{\alpha-1} X_i \right|^p - \epsilon (A_n^\alpha)^p \right)^+ \right) < \infty, \quad \forall \epsilon > 0, \quad (34)$$

when $1 - 1/r < \alpha \leq 1$.

(ii) (31) is equivalent to (34) when $\alpha = 1 - 1/r$.

(iii) (34) is equivalent to

$$\begin{cases} C_{\mathbb{V}}(|X|^p) < \infty, & \text{for } p > \frac{(r-1)}{(q\alpha)}, \\ C_{\mathbb{V}}(|X|^{(r-1)/(q\alpha)} \ln|X|) < \infty, & \text{for } p = \frac{(r-1)}{(q\alpha)}, \\ C_{\mathbb{V}}(|X|^{(r-1)/(q\alpha)}) < \infty, & \text{for } p < \frac{(r-1)}{(q\alpha)}, \end{cases} \quad (35)$$

when $0 < \alpha < 1 - 1/r$.

Remark 4. If $(\Omega, \mathcal{H}, \mathbb{E})$ is classic probability space, then Theorems 1–3 and Corollary 1 recover, respectively, Theorems 10, 11, and 14 and Corollary 13 in Guo and Shan [23] in case in which $\{X_n, n \geq 1\}$ is a sequence of independent random variables, identically distributed as X .

4. Proofs of the Main Results

Proof of Theorem 1. We first prove that (27) implies (28). Notice that

$$\begin{aligned} & \sum_{n=1}^{\infty} n^{r-2} C_{\mathbb{V}} \left(\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p - \epsilon \right)^+ \right) \\ &= \sum_{n=1}^{\infty} n^{r-2} \int_0^{\infty} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p > \epsilon + x \right) dx \\ &= \sum_{n=1}^{\infty} n^{r-2} \int_{\epsilon}^1 \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p > x \right) dx + \sum_{n=1}^{\infty} n^{r-2} \int_1^{\infty} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p > x \right) dx \\ &\leq \sum_{n=1}^{\infty} n^{r-2} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right| > \epsilon^{1/p} \right) + \sum_{n=1}^{\infty} n^{r-2} \int_1^{\infty} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right| > x^{1/p} \right) dx \\ &:= \text{I} + \text{II}. \end{aligned} \quad (36)$$

From Xu and Cheng [24] and the fact that (27) implies $C_{\mathbb{V}}(|X|^{r/q}) < \infty$, we see that $\text{I} < \infty$. We next establish $\text{II} < \infty$. Choose $0 < \alpha < 1/p$, and $\delta > 0$ is small sufficiently and integer K is large enough. For every $1 \leq i \leq n$, $n \geq 1$, we note the fact that n

is large sufficiently to guarantee $x^{\alpha} n^{-\delta} < (x^{1/p}/4K)$. Since the first finite terms of the series do not affect the convergence of the series, without loss of restrictions, we could assume the definitions of $X_{ni}^{(j)}$ below are meaningful. Write

$$\begin{aligned}
X_{ni}^{(1)} &= -x^\alpha n^{-\delta} I(a_{ni} X_i < -x^\alpha n^{-\delta}) + a_{ni} X_i I(|a_{ni} X_i| \leq x^\alpha n^{-\delta}) + x^\alpha n^{-\delta} I(a_{ni} X_i > x^\alpha n^{-\delta}), \\
X_{ni}^{(2)} &= (a_{ni} X_i - x^\alpha n^{-\delta}) I\left(x^\alpha n^{-\delta} < a_{ni} X_i < \frac{x^{1/p}}{4K}\right), \\
X_{ni}^{(3)} &= (a_{ni} X_i + x^\alpha n^{-\delta}) I\left(-\frac{x^{1/p}}{4K} < a_{ni} X_i < -x^\alpha n^{-\delta}\right), \\
X_{ni}^{(4)} &= (a_{ni} X_i + x^\alpha n^{-\delta}) I\left(a_{ni} X_i \leq -\frac{x^{1/p}}{4K}\right) + (a_{ni} X_i - x^\alpha n^{-\delta}) I\left(a_{ni} X_i \geq \frac{x^{1/p}}{4K}\right).
\end{aligned} \tag{37}$$

Observe that $\sum_{i=1}^k a_{ni} X_i = \sum_{i=1}^k X_{ni}^{(1)} + \sum_{i=1}^k X_{ni}^{(2)} + \sum_{i=1}^k X_{ni}^{(3)} + \sum_{i=1}^k X_{ni}^{(4)}$. Notice that

Consequently, to prove (28), we only need to prove that

$$\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right| > x^{1/p} \right) \subset \bigcup_{j=1}^4 \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_{ni}^{(j)} \right| > \frac{x^{1/p}}{4} \right). \tag{38}$$

$$J_j := \sum_{n=1}^{\infty} n^{r-2} \int_1^{\infty} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_{ni}^{(j)} \right| > \frac{x^{1/p}}{4} \right) dx < \infty, \quad j = 1, 2, 3, 4. \tag{39}$$

From the definition of $X_{ni}^{(4)}$, we deduce that

$$\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_{ni}^{(4)} \right| > \frac{x^{1/p}}{4} \right) \subset \left(\max_{1 \leq i \leq n} |a_{ni} X_i| > \frac{x^{1/p}}{4K} \right). \tag{40}$$

$$\int_1^{s^{1/q}} t^{\beta(r-1)/(q+\beta)} dt \approx s^{(1/q)+(\beta(r-1)/q(q+\beta))}. \tag{41}$$

From $a_{ni} \approx (i/n)^\beta (1/n^q)$, we see that

Observe that $\beta > -q/r$ implies $\beta(r-1)/(q+\beta) > -1$. Therefore,

$$\begin{aligned}
J_4 &\leq \sum_{n=1}^{\infty} n^{r-2} \sum_{i=1}^n \int_1^{\infty} \mathbb{V} \left(|a_{ni} X_i| > \frac{x^{1/p}}{4K} \right) dx \\
&\approx \sum_{n=1}^{\infty} n^{r-2} \sum_{i=1}^n \int_1^{\infty} \mathbb{V} \left(|X_i| > \frac{x^{1/p}}{4CK} n^{q+\beta} i^{-\beta} \right) dx \\
&\approx \int_1^{\infty} dx \int_1^{\infty} u^{r-2} du \int_1^u \mathbb{V} \left(|X_i| > \frac{x^{1/p}}{4CK} u^{q+\beta} v^{-\beta} \right) dv \quad (\text{Setting } s = u^{q+\beta} v^{-\beta}, t = v) \\
&\approx \int_1^{\infty} dx \int_1^{\infty} ds \int_1^{s^{1/q}} s^{(r-1)/(q+\beta)-1} t^{\beta(r-1)/(q+\beta)} \mathbb{V} \left(|X_i| > \frac{x^{1/p}}{4CK} s \right) dt \\
&\approx \int_1^{\infty} dx \int_1^{\infty} s^{r/q-1} \mathbb{V} \left(|X_i| > \frac{x^{1/p}}{4CK} s \right) ds.
\end{aligned} \tag{42}$$

Hence, from Lemma 3 and (27), we obtain $J_4 < \infty$. From the definition of $X_{ni}^{(2)}$, we conclude that $X_{ni}^{(2)} \geq 0$. From the

subadditivity of capacity and Definition 2.5 in [10], it follows that

$$\begin{aligned}
& \mathbb{V}\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_{ni}^{(2)} \right| > \frac{x^{1/p}}{4}\right) = \mathbb{V}\left(\sum_{i=1}^n X_{ni}^{(2)} > \frac{x^{1/p}}{4}\right) \\
& \leq \mathbb{V}(\text{there are at least } K \text{ indices } i \in [1, n] \text{ such that } a_{ni} X_{ni} > x^\alpha n^{-\delta}) \\
& \leq \sum_{1 \leq i_1 < i_2 < \dots < i_K \leq n} \mathbb{V}(a_{ni_j} X_{i_j} > x^\alpha n^{-\delta}, \text{ for all } 1 \leq j \leq K) \\
& \quad (\text{In Definition 2.5 in Chen [10], we set } X = (a_{ni_1} X_{i_1}, \dots, a_{ni_{K-1}} X_{i_{K-1}}), Y = a_{ni_K} X_{i_K}) \\
& \quad (\varphi(X, Y_n) = I(a_{ni_1} X_{i_1} > x^\alpha n^{-\delta}, \dots, a_{ni_{K-1}} X_{i_{K-1}} > x^\alpha n^{-\delta}, a_{ni_K} X_{i_K} > x^\alpha n^{-\delta})) \\
& = \sum_{1 \leq i_1 < i_2 < \dots < i_K \leq n} \mathbb{E}[I(a_{ni_1} X_{i_1} > x^\alpha n^{-\delta}, \dots, a_{ni_{K-1}} X_{i_{K-1}} > x^\alpha n^{-\delta})] \mathbb{V}(a_{ni_K} X_{i_K} > x^\alpha n^{-\delta}) \\
& = \dots = \sum_{1 \leq i_1 < i_2 < \dots < i_K \leq n} \prod_{j=1}^K \mathbb{V}(a_{ni_j} X_{i_j} > x^\alpha n^{-\delta}) \leq \left[\sum_{j=1}^n \mathbb{V}(a_{nj} X > x^\alpha n^{-\delta}) \right]^K.
\end{aligned} \tag{43}$$

From $\beta > -q/r$, we obtain $\sum_{i=1}^n a_{ni}^{r/q} \approx \sum_{i=1}^n n^{-r(q+\beta)/q\beta r/q} \approx n^{1-r}$. Since (27) implies $\mathbb{E}|X|^{r/q} < \infty$, by Markov's inequality under sublinear expectations (cf. (9) in Hu and Wu [30]) and (43), we conclude that

$$\begin{aligned}
J_2 & \leq \sum_{n=1}^{\infty} n^{r-2} \int_1^{\infty} \left[\sum_{j=1}^n \mathbb{V}(a_{nj} X > x^\alpha n^{-\delta}) \right]^K dx \\
& \leq \sum_{n=1}^{\infty} n^{r-2} \int_1^{\infty} \left(\sum_{j=1}^n x^{-r\alpha/q} n^{r\delta/q} a_{nj}^{r/q} \mathbb{E}|X|^{r/q} \right)^K dx \\
& \approx \sum_{n=1}^{\infty} n^{r-2-K(r-1-r\delta/q)} \int_1^{\infty} x^{-rK\alpha/q} dx.
\end{aligned} \tag{44}$$

Notice that $r > 1$ and $\alpha > 0$; we could choose δ small sufficiently and integer K large enough such that $r - 2 - K(r - 1 - r\delta/q) < -1$ and $-rK\alpha/q < -1$. Hence, from (44), we obtain $J_2 < \infty$. Similarly, we can get $J_3 < \infty$. In order to estimate J_1 , we first prove that

$$\sup_{x \geq 1} \frac{1}{x^{1/p}} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k \mathbb{E} X_{ni}^{(1)} \right| \longrightarrow 0, \quad \text{as } n \longrightarrow \infty. \tag{45}$$

Observe that (27), Lemma 4.5 in Zhang [4], and Hölder's inequality under sublinear expectations imply that $\mathbb{E}|X|^{r/q} < \infty$ and $\mathbb{E}|X|^{1/p} < \infty$. When $q > 1$, observe that $|X_{ni}^{(1)}| \leq x^\alpha n^{-\delta}$ and $|X_{ni}^{(1)}| \leq |a_{ni} X_i|$, by Hölder's inequality, we see that

$$\begin{aligned}
\max_{1 \leq k \leq n} \left| \sum_{i=1}^k \mathbb{E} X_{ni}^{(1)} \right| & \leq \sum_{i=1}^n \mathbb{E} |X_{ni}^{(1)}| \leq x^{\alpha(1-1/q)} n^{-\delta(1-1/q)} \sum_{i=1}^n \mathbb{E} |a_{ni} X_i|^{1/q} \\
& \ll x^{\alpha(1-1/q)} n^{-\delta(1-1/q)} \sum_{i=1}^n |a_{ni}|^{1/q} \\
& \leq x^{\alpha(1-1/q)} n^{-\delta(1-1/q)} n^{(r-1)/r} \left(\sum_{i=1}^n |a_{ni}|^{r/q} \right)^{1/r} \\
& \approx x^{\alpha(1-1/q)} n^{-\delta(1-1/q)}.
\end{aligned} \tag{46}$$

Observing that $\alpha(1-1/q) < \alpha < 1/p$, by (46), for any $x \geq 1$, we obtain

$$\sup_{x \geq 1} \frac{1}{x^{1/p}} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k \mathbb{E} X_{ni}^{(1)} \right| \ll n^{-\delta(1-1/q)} \longrightarrow 0, \quad \text{as } n \longrightarrow \infty. \tag{47}$$

When $1/2 < q \leq 1$, noticing that $\mathbb{E}(X) = \mathbb{E}(-X) = 0$ and by choosing δ small sufficiently such that $-\delta(1-r/q) + 1 - r < 0$, we see that

$$\begin{aligned} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k \mathbb{E} X_{ni}^{(1)} \right| &\leq 2 \sum_{i=1}^n \mathbb{E} |a_{ni} X_i| I(|a_{ni} X_i| > x^\alpha n^{-\delta}) \\ &\leq 2x^{\alpha(1-r/q)} n^{-\delta(1-r/q)} \sum_{i=1}^n \mathbb{E} |a_{ni} X_i|^{r/q} \\ &\leq 2x^{\alpha(1-r/q)} n^{-\delta(1-r/q)} \sum_{i=1}^n |a_{ni}|^{r/q} \approx x^{\alpha(1-r/q)} n^{-\delta(1-r/q)+1-r}. \end{aligned} \quad (48)$$

Observing that $1 - r/q < 0$, we have

$$\sup_{x \geq 1} \frac{1}{x^{1/p}} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k \mathbb{E} X_{ni}^{(1)} \right| \ll n^{-\delta(1-r/q)+1-r} \rightarrow 0, \quad \text{as } n \rightarrow \infty. \quad (49)$$

Thus, to establish $J_1 < \infty$, we only need to prove that

$$J_1^* := \sum_{n=1}^{\infty} n^{r-2} \int_1^{\infty} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k (X_{ni}^{(1)} - \mathbb{E} X_{ni}^{(1)}) \right| > \frac{x^{1/p}}{8} \right) dx < \infty. \quad (50)$$

Observe that $\{X_{ni}^{(1)}, 1 \leq i \leq n, n \geq 1\}$ is independent, identically distributed under $(\Omega, \mathcal{H}, \mathbb{E})$. By Markov's inequality under sublinear expectations, C_r 's inequality, and

Lemma 1, we conclude that, for a suitably large M , which will be determined later:

$$\begin{aligned} &\mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k (X_{ni}^{(1)} - \mathbb{E} X_{ni}^{(1)}) \right| > \frac{x^{1/p}}{8} \right) \\ &\ll x^{-M/p} (\ln n)^M \left\{ \sum_{i=1}^n \mathbb{E} |X_{ni}^{(1)}|^M + \left[\sum_{i=1}^n \mathbb{E} |X_{ni}^{(1)}|^2 \right]^{M/2} \right\}. \end{aligned} \quad (51)$$

Choosing large sufficiently M such that $-M(1/p - \alpha) - r\alpha/q < -1$ and $-1 - (M - r/q)\delta < -1$, we have

$$\begin{aligned} &\sum_{n=1}^{\infty} n^{r-2} (\ln n)^M \sum_{i=1}^n \int_1^{\infty} x^{-M/p} \mathbb{E} |X_{ni}^{(1)}|^M dx \\ &\ll \sum_{n=1}^{\infty} n^{r-2} n^{-\delta(M-r/q)} (\ln n)^M \sum_{i=1}^n |a_{ni}|^{r/q} \int_1^{\infty} x^{-M(1/p-\alpha)-r\alpha/q} dx \\ &\ll \sum_{n=1}^{\infty} n^{-1-\delta(M-r/q)} (\ln n)^M < \infty. \end{aligned} \quad (52)$$

When $r/q \geq 2$, (27) implies that $\mathbb{E}|X|^2 < \infty$. Observing that $q > 1/2$, we could choose M large enough such that

$-M/p < -1$, $r - 2 - (2q - 1)M/2 < -1$. Then, by $\beta > -q/r \geq -(1/2)$,

$$\begin{aligned} &\sum_{n=1}^{\infty} n^{r-2} (\ln n)^M \int_1^{\infty} x^{-M/p} \left[\sum_{i=1}^n \mathbb{E} |X_{ni}^{(1)}|^2 \right]^{M/2} dx \\ &\ll \sum_{n=1}^{\infty} n^{r-2} (\ln n)^M \left(\sum_{i=1}^n a_{ni}^2 \right)^{M/2} \int_1^{\infty} x^{-M/p} dx \\ &\ll \sum_{n=1}^{\infty} n^{r-2} (\ln n)^M \left(\sum_{i=1}^n i^{2\beta} n^{-2(q+\beta)} \right)^{M/2} \\ &\ll \sum_{n=1}^{\infty} n^{r-2-(2q-1)M/2} (\ln n)^M < \infty. \end{aligned} \quad (53)$$

When $r/q < 2$, choosing M large enough such that $-M[1/p - \alpha + ar/(2q)] < -1$ and $r - 2 - [\delta(2 - r/q) + r - 1]M/2 < -1$, we have

$$\begin{aligned} & \sum_{n=1}^{\infty} n^{r-2} (\ln n)^M \int_1^{\infty} x^{-M/p} \left[\sum_{i=1}^n \mathbb{E} |X_{ni}^{(1)}|^2 \right]^{M/2} dx \\ & \ll \sum_{n=1}^{\infty} n^{r-2} n^{-\delta(2-r/q)M/2} (\ln n)^M \left(\sum_{i=1}^n a_{ni}^{r/q} \right)^{M/2} \int_1^{\infty} x^{-M[1/p - \alpha + ar/(2q)]} dx \\ & \ll \sum_{n=1}^{\infty} n^{r-2 - [\delta(2-r/q) + r - 1]M/2} (\ln n)^M < \infty. \end{aligned} \quad (54)$$

Consequently, by (51)–(54), we get $J_1^* < \infty$. Now, we prove that (28) implies (27). By Markov's inequality under

sublinear expectations (see (8) of Wu [6]), (28), and Lemma 4.5 (iii) in [4],

$$\begin{aligned} & \sum_{n=1}^{\infty} n^{r-2} \mathbb{V} \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right| > \epsilon \right) \\ & = \sum_{n=1}^{\infty} n^{r-2} \mathbb{E} \left(1 \cdot I \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right| > \epsilon \right) \right) \\ & \leq \sum_{n=1}^{\infty} n^{r-2} \mathbb{E} \left(\left(\frac{\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p - (\epsilon/2)^p \right)^+}{(\epsilon/2)^p} \right) I \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right| > \epsilon \right) \right) \\ & \leq \sum_{n=1}^{\infty} n^{r-2} \mathbb{E} \left(\left(\frac{\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p - (\epsilon/2)^p \right)^+}{(\epsilon/2)^p} \right) \right) \\ & \leq \sum_{n=1}^{\infty} n^{r-2} \mathbb{E} \frac{\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|^p - (\epsilon/2)^p \right)^+}{(\epsilon/2)^p} < \infty. \end{aligned} \quad (55)$$

Since

$$\max_{1 \leq k \leq n} |a_{nk} X_k| \leq 2 \max_{1 \leq k \leq n} \left| \sum_{i=1}^k a_{ni} X_i \right|, \quad (56)$$

by (55) and similar proofs of (3.17) in [31], we conclude that

$$\mathbb{V} \left(\max_{1 \leq k \leq n} |a_{nk} X_k| > \epsilon \right) \longrightarrow 0, \quad \text{as } n \longrightarrow \infty. \quad (57)$$

Therefore, by (57) and Lemma 2, we obtain

$$\sum_{i=1}^n \mathbb{V}(|a_{ni} X_i| > \epsilon) \ll \mathbb{V} \left(\max_{1 \leq k \leq n} |a_{nk} X_k| > \epsilon \right), \quad \text{for all } \epsilon > 0. \quad (58)$$

Now, combining (58) with (28) gives

$$\sum_{n=1}^{\infty} n^{r-2} \int_{\epsilon}^{\infty} \sum_{i=1}^n \mathbb{V}(|a_{ni} X_i| > x^{1/p}) dx < \infty. \quad (59)$$

From the proof of (42), we conclude that (59) is equivalent to (27). $\square$

Proof. of Theorem 2. As in the proof of Theorem 1, with Lemma 4 in place of Lemma 3, and using

$$\begin{aligned} \sum_{i=1}^n a_{ni}^{r/q} & \approx \sum_{i=1}^n n^{-r(q+\beta)/q} t^{\beta r/q} \approx n^{-r(q+\beta)/q} \ln n, \\ \int_1^{s^{1/q}} t^{\beta(r-1)/(q+\beta)} dt & = \int_1^{s^{1/q}} t^{-1} dt \approx \ln(s), \end{aligned} \quad (60)$$

when $\beta = -q/r$, we could prove Theorem 2. Hence, the proof is omitted. $\square$

Proof. of Theorem 3. As in the proof of Theorem 1, using

$$\sum_{i=1}^n a_{ni}^{r/q} \approx \sum_{i=1}^n n^{-r(q+\beta)/q} t^{\beta r/q} \approx n^{-r(q+\beta)/q}, \quad \text{if } \beta < \frac{-q}{r}, \quad (61)$$

and with

$$\int_1^{s^{1/q}} t^{\beta(r-1)/(q+\beta)} dt \approx C, \quad \text{if } \beta < \frac{-q}{r}, \quad (62)$$

in place of (41), we could prove Theorem 3. Thus, the proof is omitted. $\square$

Proof. of Theorem 4. Setting $a_{ni} = (A_{n-i}^{\alpha-1}/A_n^\alpha)^q$, $0 \leq i \leq n$ and $n \geq 1$, we observe that $a_{ni} \approx (n-i)^{q(\alpha-1)} n^{-q\alpha}$, $0 \leq i < n$, $n \geq 1$, and $a_{nn} \approx n^{-q\alpha}$. As in the proof of Theorem 14 in [23], letting $\beta = q(\alpha-1)$ in Corollary 1, we finish the proof of Theorem 4. $\square$

Data Availability

No data were used to support this study.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

Authors' Contributions

All authors contributed equally and read and approved the final manuscript.

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