

ON THE CONVERGENCE OF SERIES OF MARTINGALE DIFFERENCES WITH MULTIDIMENSIONAL INDICES

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ABSTRACT. Let $\{X_{\mathbf{n}}; \mathbf{n} \succeq \mathbf{1}\}$ be a field of martingale differences taking values in a p -uniformly smooth Banach space. The paper provides conditions under which the series $\sum_{\mathbf{i} \preceq \mathbf{n}} X_{\mathbf{i}}$ converges almost surely and the tail series $\{T_{\mathbf{n}} = \sum_{\mathbf{i} \succ \mathbf{n}} X_{\mathbf{i}}; \mathbf{n} \succeq \mathbf{1}\}$ satisfies $\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_{\mathbf{k}}\| = \mathcal{O}_P(b_{\mathbf{n}})$ and $\frac{\sup_{\mathbf{k} \succ \mathbf{n}} \|T_{\mathbf{k}}\|}{B_{\mathbf{n}}} \xrightarrow{P} 0$ for given fields of positive numbers $\{b_{\mathbf{n}}\}$ and $\{B_{\mathbf{n}}\}$. This result generalizes results of A. Rosalsky, J. Rosenblatt [7], [8] and S. H. Sung, A. I. Volodin [11].

1. Introduction

Let $\{X_n; n \geq 1\}$ be a sequence of random variables taking values in a real separable Banach space \mathbb{E} with norm $\|\cdot\|$. If the series $\sum_{i=1}^{\infty} X_i$ converges a.s., then the tail series

$$T_n = \sum_{i=n}^{\infty} X_i, \quad n \geq 1$$

is well-defined and $\sup_{k \geq n} \|T_k\| \xrightarrow{P} 0$ as $n \rightarrow \infty$.

A. Rosalsky, J. Rosenblatt [7], [8] and S. H. Sung, A. I. Volodin [11] investigated the rate in which $\sup_{k \geq n} \|T_k\| \xrightarrow{P} 0$ as $n \rightarrow \infty$. Namely, they provided conditions under which $\sup_{k \geq n} \|T_k\| = \mathcal{O}_P(b_n)$ and $\frac{\sup_{k \geq n} \|T_k\|}{B_n} \xrightarrow{P} 0$, where (b_n) and (B_n) are given sequences of positive numbers.

The aim of this paper is to extend these results to the case where $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}; \mathbf{n} \in \mathbb{N}^d\}$ is a field of \mathbb{E} -valued martingale differences. Under the assumption that \mathbb{E} is a p -uniformly smooth Banach space for some $1 \leq p \leq 2$, we will provide sufficient conditions ensuring that $\sum_{\mathbf{n} \succeq \mathbf{1}} X_{\mathbf{n}}$ converges, $\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_{\mathbf{k}}\| = \mathcal{O}_P(b_{\mathbf{n}})$, $\frac{\sup_{\mathbf{k} \succ \mathbf{n}} \|T_{\mathbf{k}}\|}{B_{\mathbf{n}}} \xrightarrow{P} 0$ and $\sum_{\mathbf{n} \succeq \mathbf{1}} \frac{1}{|\mathbf{n}|} P(\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_{\mathbf{k}}\| > \varepsilon a_{\mathbf{n}}) < \infty$ for every $\varepsilon > 0$, where $T_{\mathbf{n}} = S - \sum_{\mathbf{i} \preceq \mathbf{n}} X_{\mathbf{i}}$, $\{b_{\mathbf{n}}\}$, $\{B_{\mathbf{n}}\}$, $\{a_{\mathbf{n}}\}$ are given fields of positive numbers.

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2. Preliminaries and some useful lemmas

Throughout this paper, the symbol C will denote a generic constant ($0 < C < \infty$) which is not necessarily the same one in each appearance.

Let \mathbb{E} be a real separable Banach space. For a \mathbb{E} -valued random variable X and sub σ -algebra \mathcal{G} of \mathcal{F} , the conditional expectation $E(X|\mathcal{G})$ is defined analogously to that in the real-random variable case and enjoys similar properties (see [10]).

$(\mathbb{E}, \|\cdot\|)$ is said to be p -uniformly smooth ($1 \leq p \leq 2$) if there exists a finite positive constant C such that for all \mathbb{E} -valued martingales $\{S_n; 1 \leq n \leq m\}$

$$(2.1) \quad E\|S_m\|^p \leq C \sum_{n=1}^m E\|S_n - S_{n-1}\|^p.$$

Clearly, every separable Banach space is of 1-uniformly smooth. If a real separable Banach space is of p -uniformly smooth for some $1 < p \leq 2$, then it is of r -uniformly smooth for all $r \in [1, p)$. A Hilbert space is 2-uniformly smooth and the space L_p is $\min\{p, 2\}$ -uniformly smooth (see [5], [6]).

Let d be a positive integer, the set of all nonnegative integer d -dimensional lattice points will be denoted by \mathbb{N}_0^d and the set of all positive integer d -dimensional lattice points will be denoted by \mathbb{N}^d . For $\mathbf{m} = (m_1, \dots, m_d) \in \mathbb{N}_0^d$, $\mathbf{n} = (n_1, \dots, n_d) \in \mathbb{N}_0^d$, $\alpha = (\alpha_1, \dots, \alpha_d) \in \mathbb{N}_0^d$, denote $\mathbf{m} + \mathbf{n} = (m_1 + n_1, \dots, m_d + n_d)$, $\mathbf{m} - \mathbf{n} = (m_1 - n_1, \dots, m_d - n_d)$, $|\mathbf{n}| = n_1 \cdot n_2 \cdots n_d$, $\|\mathbf{n}\| = \min\{n_1, \dots, n_d\}$, $\mathbf{1} = (1, \dots, 1) \in \mathbb{N}_0^d$, $\bigvee_{i=1}^d (m_i < n_i)$ means that there is at least one of $m_1 < n_1, m_2 < n_2, \dots, m_d < n_d$ holds. We write $\mathbf{m} \preceq \mathbf{n}$ (or $\mathbf{n} \succeq \mathbf{m}$) if $m_i \leq n_i, 1 \leq i \leq d$; $\mathbf{m} \prec \mathbf{n}$ if $\mathbf{m} \preceq \mathbf{n}$ and $\mathbf{m} \neq \mathbf{n}$; $\mathbf{m} \ll \mathbf{n}$ (or $\mathbf{n} \gg \mathbf{m}$) if $\bigvee_{i=1}^d (m_i < n_i)$.

Let (Ω, \mathcal{F}, P) be a probability space, \mathbb{E} be a real separable Banach space, $\mathcal{B}(\mathbb{E})$ be the σ -algebra of all Borel sets in \mathbb{E} . Let $\{X_{\mathbf{n}}, \mathbf{1} \preceq \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ be a (d -dimensional) field of \mathbb{E} -valued random variables and $\{\mathcal{F}_{\mathbf{n}}, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ be a (d -dimensional) field of nondecreasing sub- σ -algebras of \mathcal{F} with respect to the partial order \preceq on \mathbb{N}^d such that $X_{\mathbf{n}}$ is $\mathcal{F}_{\mathbf{n}}$ -measurable for all $\mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}$, then $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is said to be an *adapted field*.

Let $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ be an adapted field. For $\mathbf{n} \in \mathbb{N}_0^d$ ($\mathbf{m} - \mathbf{1} \preceq \mathbf{n} \preceq \mathbf{M} - \mathbf{1}$), we adopt the convention that $\mathcal{F}_{\mathbf{n}} = \{\emptyset, \Omega\}$ if there exists a positive i ($1 \leq i \leq d$) such that $n_i = m_i - 1$ and set

$$\mathcal{F}_{\mathbf{n}}^{(i)} = \sigma\{\mathcal{F}_{\mathbf{k}} : \mathbf{k} = (k_1, \dots, k_d), m_j \leq k_j \leq M_j (j \neq i), \text{ and } k_i = n_i\}$$

for all $1 \leq i \leq d$, and $\mathcal{F}_{\mathbf{n}}^* = \sigma\{\mathcal{F}_{\mathbf{n}}^{(i)} : 1 \leq i \leq d\}$.

The adapted field $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is said to be a *field of martingale differences* if $E(X_{\mathbf{n}}|\mathcal{F}_{\mathbf{n}-1}^{(i)}) = 0$ for all $\mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}, 1 \leq i \leq d$ (see [3]).

The adapted field $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is said to be *strong adapted* if $E(X_{\mathbf{n}}|\mathcal{F}_{\mathbf{n}-1}^*)$ is $\mathcal{F}_{\mathbf{n}}^{(i)}$ -measurable for all $\mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}, 1 \leq i \leq d$.

Remark 2.1. Let $\{X_n, \mathcal{F}_n : \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ be a field of martingale differences. Then it is strong adapted. Conversely, let $\{X_n, \mathcal{F}_n : \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ be strong adapted, when $d = 1$ then $\{X_n - E(X_n|\mathcal{F}_{n-1}^*), \mathcal{F}_n : m \leq n \leq M\}$ is a sequences of martingale differences, but when $d > 1$ then $\{X_n - E(X_n|\mathcal{F}_{n-1}^*), \mathcal{F}_n : \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is not necessarily a field of martingale differences, because $X_n - E(X_n|\mathcal{F}_{n-1}^*)$ may not be \mathcal{F}_n -measurable.

The adapted field $\{X_n, \mathcal{F}_n, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is said to be *strong* adapted* if $\{X_n I_A, \mathcal{F}_n, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is strong adapted for all $A \in \sigma(X_n)$.

Clearly, when $\{X_n, \mathcal{F}_n, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is a field of martingale differences, then it is not necessarily a strong* adapted field.

The adapted field $\{X_n, \mathcal{F}_n, \mathbf{m} \preceq \mathbf{n} \preceq \mathbf{M}\}$ is said to be a field of *strong* martingale differences* if it is a strong* adapted and a field of martingale differences.

Remark 2.2. Let $\{X_n, \mathcal{F}_n : m \leq n \leq M\}$ be a sequence of martingale differences, by $E(X_n I_A | \mathcal{F}_{n-1}^*) = E(X_n I_A | \mathcal{F}_{n-1}) \in \mathcal{F}_{n-1} = \mathcal{F}_{n-1}^{(1)}$. Then $\{X_n, \mathcal{F}_n : m \leq n \leq M\}$ is a strong* martingale differences.

Example 1. Let $\{X_n, \mathbf{n} \in \mathbb{N}^d\}$ be a field of independent random variables with mean 0. Put $\mathcal{F}_n = \sigma(X_k, \mathbf{k} \preceq \mathbf{n})$, then $E(X_n | \mathcal{F}_{n-1}^i) = 0$, $E(X_n I_A | \mathcal{F}_{n-1}^*) = 0$ for all $A \in \sigma(X_n)$ and $\mathbf{n} \in \mathbb{N}^d$, $1 \leq i \leq d$. Therefore, $\{X_n, \mathcal{F}_n, \mathbf{n} \in \mathbb{N}^d\}$ is a field of strong* martingale differences.

Example 2. Let $\{X_n, \mathcal{G}_n : n \geq 1\}$ be a sequence of martingale differences and set

$$X_n = X_n \text{ if } \mathbf{n} = (n, n, \dots, n) \text{ and } X_n = 0 \text{ if } \mathbf{n} \neq (n, n, \dots, n);$$

$$\mathcal{G}_n = \mathcal{G}_n \text{ if } \mathbf{n} = (n, n, \dots, n) \text{ and } \mathcal{G}_n = \{\emptyset, \Omega\} \text{ if } \mathbf{n} \neq (n, n, \dots, n).$$

Let $\mathcal{F}_n = \sigma\{\mathcal{G}_k, \mathbf{k} \preceq \mathbf{n}\}$ for all $\mathbf{n} \succeq \mathbf{1}$. Then $\{X_n, \mathcal{F}_n : \mathbf{n} \succeq \mathbf{1}\}$ is a field of martingale differences. Moreover, for all $\mathbf{n} \succeq \mathbf{1}$, then

$$E(X_n I_A | \mathcal{F}_{n-1}^*) = E(X_n I_A | \mathcal{G}_{n-1}) \in \mathcal{G}_n = \mathcal{F}_n^i$$

if $\mathbf{n} = (n, n, \dots, n)$ and $E(X_n I_A | \mathcal{F}_{n-1}^*) = 0 \in \mathcal{F}_n^i$ if otherwise, for all $A \in \sigma(X_n)$, so $\{X_n, \mathcal{F}_n : \mathbf{n} \succeq \mathbf{1}\}$ is a field of strong* martingale differences.

Example 3. Let $\{Y_n, \mathbf{n} \in \mathbb{N}^d\}$ be a field of independent random variables with mean 0. Put $\mathcal{F}_n = \sigma(Y_k, \mathbf{k} \preceq \mathbf{n})$ and $X_n = \prod_{\mathbf{k} \preceq \mathbf{n}} Y_k$, so $\{X_n, \mathbf{n} \in \mathbb{N}^d\}$ is not a field of independent random variables. If $E X_n < \infty$ for all $\mathbf{n} \succeq \mathbf{1}$, then $E(X_n | \mathcal{F}_n^i) = 0$, $E(X_n I_A | \mathcal{F}_{n-1}^*) = E(X_n I_A) \in \mathcal{F}_n^i$ for all $A \in \sigma(X_n)$, $\mathbf{n} \succeq \mathbf{1}$, $1 \leq i \leq d$. Therefore, $\{X_n, \mathcal{F}_n, \mathbf{n} \in \mathbb{N}^d\}$ is a field of strong* martingale differences.

In the sequel the following lemmas are useful.

Lemma 2.3. *Let \mathbb{E} be a real separable p -uniformly smooth Banach space for some $1 \leq p \leq 2$. Then there exists a positive constant C such that for all strong*

adapted fields of \mathbb{E} -valued random variables $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}} : \mathbf{1} \preceq \mathbf{n} \preceq \mathbf{m}\}$, we have

$$(2.2) \quad E \max_{\mathbf{1} \preceq \mathbf{k} \preceq \mathbf{m}} \left\| \sum_{\mathbf{1} \preceq \mathbf{i} \preceq \mathbf{k}} (X_{\mathbf{i}} - E(X_{\mathbf{i}} | \mathcal{F}_{\mathbf{i}-1}^*)) \right\|^p \leq C \sum_{\mathbf{1} \preceq \mathbf{k} \preceq \mathbf{m}} E \|X_{\mathbf{k}}\|^p.$$

Proof. We will show that (2.2) holds by induction. Set

$$S_{\mathbf{k}} = \sum_{\mathbf{1} \preceq \mathbf{i} \preceq \mathbf{k}} (X_{\mathbf{i}} - E(X_{\mathbf{i}} | \mathcal{F}_{\mathbf{i}-1}^*)).$$

Firstly, for $d = 1$, note that $\{\max_{\mathbf{1} \leq i \leq k} \|S_i\|, \mathcal{F}_k : 1 \leq k \leq m\}$ is a nonnegative sub-martingale. Applying Doob's inequality and by (2.1), we have (2.2). We assume that (2.2) holds for $d - 1$; we will show that it holds for d .

Denote

$$\mathbf{k} = (k_1, \dots, k_{d-1}, k_d) = (\mathbf{k}', k_d); \quad \mathbf{m} = (m_1, \dots, m_{d-1}, m_d) = (\mathbf{m}', m_d);$$

with $\mathbf{k}', \mathbf{m}' \in \mathbb{N}^{d-1}$; set $Y_{k_d} = \max_{\mathbf{1} \preceq \mathbf{k}' \preceq \mathbf{m}'} \|S_{(\mathbf{k}'; k_d)}\|$ for each $1 \leq k_d \leq m_d$, we have

$$\begin{aligned} & E(S_{(\mathbf{k}'; k_d)} | \mathcal{F}_{(\mathbf{k}'; k_d-1)}^d) \\ &= E(S_{(\mathbf{k}'; k_d-1)} | \mathcal{F}_{(\mathbf{k}'-1; k_d-1)}^d) \\ & \quad + \sum_{\mathbf{1} \preceq \mathbf{k}' \preceq \mathbf{m}'} (E(X_{(\mathbf{k}'; k_d)} - E(X_{(\mathbf{k}'; k_d)} | \mathcal{F}_{(\mathbf{k}'-1, k_d-1)}^*)) | \mathcal{F}_{(\mathbf{k}'-1; k_d-1)}^d) \\ &= S_{(\mathbf{k}'; k_d-1)} \end{aligned}$$

and by $\{S_{(\mathbf{k}'; k_d)}; \mathcal{F}_{(\mathbf{k}'; k_d)} : 1 \leq k_d \leq m_d\}$ being a strong adapted random field, it means that for each $\mathbf{1} \preceq \mathbf{k}' \preceq \mathbf{m}'$ then $\{S_{(\mathbf{k}'; k_d)}; \mathcal{F}_{(\mathbf{k}'; k_d)}^d : 1 \leq k_d \leq m_d\}$ is a martingale, and we have that $\{Y_{k_d}; \mathcal{F}_{(\mathbf{k}'; k_d)} : 1 \leq k_d \leq m_d\}$ is a nonnegative sub-martingale sequence. Applying Doob's inequality, we obtain

$$\begin{aligned} E\left(\max_{\mathbf{1} \preceq (\mathbf{k}', k_d) \preceq \mathbf{m}} \|S_{(\mathbf{k}'; k_d)}\|^p\right) &= E\left(\max_{1 \leq k_d \leq m_d} Y_{k_d}^p\right) \leq C \cdot E Y_{m_d}^p \\ &= C \cdot E\left(\max_{\mathbf{1} \preceq \mathbf{k}' \preceq \mathbf{m}'} \|S_{(\mathbf{k}'; m_d)}\|^p\right). \end{aligned}$$

Set

$$X_{\mathbf{k}'}^{d-1} = \sum_{1 \leq k_d \leq m_d} X_{(\mathbf{k}'; k_d)}; \quad \mathcal{F}_{\mathbf{k}'}^{d-1} = \sigma(\mathcal{F}_{(\mathbf{k}'; k_d)}^{d-1} : 1 \leq k_d \leq m_d).$$

Note that $\mathcal{F}_{(\mathbf{k}', k_d)}^i = (\mathcal{F}_{\mathbf{k}'}^{d-1})^i$, $\mathcal{F}_{(\mathbf{k}', k_d)}^* = (\mathcal{F}_{\mathbf{k}'}^{d-1})^*$ for all $1 \leq k_d \leq m_d$, $1 \leq i \leq d - 1$, then $\{X_{\mathbf{k}'}^{d-1}, \mathcal{F}_{\mathbf{k}'}^{d-1} : \mathbf{1} \preceq \mathbf{k}' \preceq \mathbf{m}'\}$ is a strong adapted field. Therefore, by the induction assumption,

$$\begin{aligned} & E \max_{\mathbf{1} \preceq (\mathbf{k}', k_d) \preceq \mathbf{m}} \|S_{(\mathbf{k}'; k_d)}\|^p \\ & \leq C \cdot \sum_{\mathbf{1} \preceq \mathbf{k}' \preceq \mathbf{m}'} E \left\| \sum_{1 \leq k_d \leq m_d} (X_{(\mathbf{k}'; k_d)} - E(X_{(\mathbf{k}'; k_d)} | \mathcal{F}_{(\mathbf{k}'-1, k_d-1)}^*)) \right\|^p \\ & \leq C \sum_{\mathbf{1} \preceq \mathbf{k} \preceq \mathbf{m}} E \|X_{\mathbf{k}} - E(X_{\mathbf{k}} | \mathcal{F}_{\mathbf{k}-1}^*)\|^p \leq C \sum_{\mathbf{1} \preceq \mathbf{k} \preceq \mathbf{m}} E \|X_{\mathbf{k}}\|^p, \quad \text{by (1.1)}. \quad \square \end{aligned}$$

Remark 2.4. If $\{X_{\mathbf{n}}; \mathbf{n} \succeq \mathbf{1}\}$ is a \mathbb{E} -valued martingale difference field, from Lemma 2.3, we obtain Lemma 1.1 in [3] (for $p = q$). Moreover, by Remark 2.1, Lemma 2.3 is stronger than Lemma 1.1 in [3] (for $p = q$).

Lemma 2.5. *Let $\{X_{\mathbf{n}}; \mathbf{n} \succeq \mathbf{1}\}$ be a field of \mathbb{E} -valued random variables. Then*

$$(2.3) \quad P\left(\sup_{\mathbf{k} \succeq \mathbf{m}} \|X_{\mathbf{k}}\| > \epsilon\right) = \lim_{\|\mathbf{n}\| \rightarrow \infty} P\left(\max_{\mathbf{m} \preceq \mathbf{k} \preceq \mathbf{n}} \|X_{\mathbf{k}}\| > \epsilon\right),$$

$$(2.4) \quad P\left(\liminf_{\|\mathbf{n}\| \rightarrow \infty} \|X_{\mathbf{n}}\| > \epsilon\right) \leq \liminf_{\|\mathbf{n}\| \rightarrow \infty} P(\|X_{\mathbf{n}}\| > \epsilon).$$

Proof. 1. Remark that for $d = 1$, by the continuity from below theorem, we have (2.3). Assume that (2.3) holds for $d = D - 1 \geq 1$, we wish to show that for $d = D$. Let $\mathbf{m} = (m_1, m_2, \dots, m_d) = (m_1, \mathbf{m}_1)$, $\mathbf{k} = (k_1, k_2, \dots, k_d) = (k_1, \mathbf{k}_1)$, $\mathbf{n} = (n_1, n_2, \dots, n_d) = (n_1, \mathbf{n}_1)$, by the continuity from below theorem, we have

$$\begin{aligned} P\left(\sup_{\mathbf{k} \succeq \mathbf{m}} \|X_{\mathbf{k}}\| > \epsilon\right) &= P\left(\lim_{n_1 \rightarrow \infty} \sup_{\mathbf{k}_1 \succeq \mathbf{m}_1} \max_{m_1 \leq k_1 \leq n_1} \|X_{\mathbf{k}}\| > \epsilon\right) \\ &= \lim_{n_1 \rightarrow \infty} P\left(\sup_{\mathbf{k}_1 \succeq \mathbf{m}_1} \max_{m_1 \leq k_1 \leq n_1} \|X_{\mathbf{k}}\| > \epsilon\right). \end{aligned}$$

By the induction assumption,

$$\begin{aligned} P\left(\sup_{\mathbf{k} \succeq \mathbf{m}} \|X_{\mathbf{k}}\| > \epsilon\right) &= \lim_{n_1 \rightarrow \infty} \lim_{\|\mathbf{n}_1\| \rightarrow \infty} P\left(\max_{\mathbf{m} \preceq \mathbf{k} \preceq \mathbf{n}} \|X_{\mathbf{k}}\| > \epsilon\right) \\ &= \lim_{\|\mathbf{n}\| \rightarrow \infty} P\left(\max_{\mathbf{m} \preceq \mathbf{k} \preceq \mathbf{n}} \|X_{\mathbf{k}}\| > \epsilon\right). \end{aligned}$$

2. By Theorem 8.1.3 of Chow and Teicher [1] and the same argument as in the proof of (2.3), we have (2.4). \square

Lemma 2.6. *Let $\{X_{\mathbf{n}}; \mathbf{n} \succeq \mathbf{1}\}$ be a field of \mathbb{E} -valued random variables. Then, $X_{\mathbf{n}}$ converges a.s. as $\|\mathbf{n}\| \rightarrow \infty$ if and only if for all $\epsilon > 0$,*

$$(2.5) \quad \lim_{\|\mathbf{n}\| \rightarrow \infty} P\left(\sup_{\mathbf{k} \succeq \mathbf{0}} \|X_{\mathbf{n}+\mathbf{k}} - X_{\mathbf{n}}\| > \epsilon\right) = 0.$$

Proof. Necessity. Suppose that $X_{\mathbf{n}} \rightarrow X$ a.s. as $\|\mathbf{n}\| \rightarrow \infty$. Then (2.5) holds, by the following inequality

$$\sup_{\mathbf{k} \succeq \mathbf{0}} \|X_{\mathbf{n}+\mathbf{k}} - X_{\mathbf{n}}\| \leq \sup_{\mathbf{m} \succeq \mathbf{n}} \|X_{\mathbf{m}} - X\| + \|X_{\mathbf{n}} - X\|.$$

Sufficiency. Suppose (2.5) holds, let $\mathbf{n}' = (n, n, \dots, n)$, $\mathbf{k}' = (k, k, \dots, k)$, we have $n \rightarrow \infty$ if and only if $\|\mathbf{n}'\| \rightarrow \infty$. Set $Y_n = X_{\mathbf{n}'}$ for all $n \geq 1$. Then for an arbitrary $\epsilon > 0$,

$$\lim_{n \rightarrow \infty} P\left(\sup_{k \geq 0} \|Y_{n+k} - Y_n\| > \epsilon\right) = \lim_{\|\mathbf{n}'\| \rightarrow \infty} P\left(\sup_{\mathbf{k}' \succeq \mathbf{0}} \|X_{\mathbf{n}'+\mathbf{k}'} - X_{\mathbf{n}'}\| > \epsilon\right) = 0$$

which implies that Y_n converges a.s. to a certain random variable X as $n \rightarrow \infty$, i.e., $X_{\mathbf{n}'}$ converges a.s. to X as $\|\mathbf{n}'\| \rightarrow \infty$. Now we prove $X_{\mathbf{n}} \rightarrow X$ a.s. as $\|\mathbf{n}\| \rightarrow \infty$.

For an arbitrary $\varepsilon > 0$,

$$P\left(\sup_{\mathbf{n} \succeq \mathbf{n}'} \|X_{\mathbf{n}} - X\| > \varepsilon\right) \leq P\left(\sup_{\mathbf{n} \succeq \mathbf{n}'} \|X_{\mathbf{n}} - X_{\mathbf{n}'}\| > \varepsilon/2\right) + P(\|X_{\mathbf{n}'} - X\| > \varepsilon/2) \rightarrow 0 \text{ as } n \rightarrow \infty,$$

so $X_{\mathbf{n}} \rightarrow X$ a.s. as $\|\mathbf{n}\| \rightarrow \infty$. □

3. Main results

Let $\{X_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ be a field of random variables in Banach space \mathbb{E} . Put $S_{\mathbf{n}} = \sum_{\mathbf{k} \preceq \mathbf{n}} X_{\mathbf{k}}$ for all $\mathbf{n} \succeq \mathbf{1}$. The series $\sum_{\mathbf{n} \succeq \mathbf{1}} X_{\mathbf{n}}$ is said to *converge a.s.* if the field of \mathbb{E} -valued random variables $\{S_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ converges a.s.. In this case, put

$$S = \lim_{\|\mathbf{n}\| \rightarrow \infty} S_{\mathbf{n}}$$

and

$$T_{\mathbf{n}} = S - S_{\mathbf{n}} = \sum_{\mathbf{k} \succ \mathbf{n}} X_{\mathbf{k}}$$

(set $S_{\mathbf{0}} = 0$). We have

$$\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_{\mathbf{k}}\| \xrightarrow{P} 0 \quad \text{as} \quad \|\mathbf{n}\| \rightarrow \infty.$$

The following theorems provide sufficient conditions a.s. for the convergence of $\sum_{\mathbf{n} \succeq \mathbf{1}} X_{\mathbf{n}}$ as well as the rate of convergence to 0 of $\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_{\mathbf{k}}\|$.

Theorem 3.1. *Let \mathbb{E} be a p -uniformly smooth Banach space for some $1 \leq p \leq 2$, $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}; \mathbf{n} \in \mathbb{N}^d\}$ be a field of \mathbb{E} -valued martingale differences. Let $\{b_{\mathbf{n}}\}, \{B_{\mathbf{n}}\}$ be fields of positive constants, such that $b_{\mathbf{n}} = o(1), B_{\mathbf{n}} = o(1)$ as $\|\mathbf{n}\| \rightarrow \infty$.*

(1) *If*

$$(3.1) \quad \sum_{\mathbf{k} \succ \mathbf{n}} E\|X_{\mathbf{k}}\|^p = \mathcal{O}(b_{\mathbf{n}}^p),$$

then $\sum_{\mathbf{n} \succeq \mathbf{1}} X_{\mathbf{n}}$ converges a.s. and

$$(3.2) \quad \sup_{\mathbf{k} \succeq \mathbf{n}} \|T_{\mathbf{k}}\| = \mathcal{O}_P(b_{\mathbf{n}}).$$

(2) *If*

$$(3.3) \quad \sum_{\mathbf{k} \succ \mathbf{n}} E\|X_{\mathbf{k}}\|^p = o(B_{\mathbf{n}}^p)$$

as $\|\mathbf{n}\| \rightarrow \infty$, then $\sum_{\mathbf{n} \geq \mathbf{1}} X_{\mathbf{n}}$ converges a.s. and

$$(3.4) \quad \frac{\sup_{\mathbf{k} \geq \mathbf{n}} \|T_{\mathbf{k}}\|}{B_{\mathbf{n}}} \xrightarrow{P} 0$$

as $\|\mathbf{n}\| \rightarrow \infty$.

Proof. (1) Set $S_{\mathbf{n}} = \sum_{\mathbf{k} \leq \mathbf{n}} X_{\mathbf{k}}$. For an arbitrary $\varepsilon > 0$, set $\mathbf{n} = (n_1, \dots, n_i, \dots, n_d) = (\mathbf{n}_i, n_i, \mathbf{n}'_i)$, $\mathbf{k} = (k_1, \dots, k_i, \dots, k_d) = (\mathbf{k}_i, k_i, \mathbf{k}'_i)$, $\mathbf{j} = (j_1, \dots, j_i, \dots, j_d) = (\mathbf{j}_i, j_i, \mathbf{j}'_i)$ for all $1 \leq i \leq d$. We have that

$$\begin{aligned} & P \left(\sup_{\mathbf{k} \geq \mathbf{0}} \|S_{\mathbf{n}+\mathbf{k}} - S_{\mathbf{n}}\| > \varepsilon \right) \\ &= P \left(\sup_{\mathbf{k} \geq \mathbf{0}} \left\| \sum_{\mathbf{n} \ll \mathbf{j} \leq \mathbf{n}+\mathbf{k}} X_{\mathbf{j}} \right\| > \varepsilon \right) \\ &\leq \sum_{i=1}^d P \left(\sup_{\mathbf{k} \geq \mathbf{0}} \left\| \sum_{\mathbf{1} \leq \mathbf{j}_i \leq \mathbf{n}_i} \sum_{j_i=n_i}^{n_i+k_i} \sum_{\mathbf{1} \leq \mathbf{j}'_i \leq \mathbf{n}'_i+\mathbf{k}'_i} X_{(\mathbf{j}_i, j_i, \mathbf{j}'_i)} \right\| > \varepsilon/d \right). \end{aligned}$$

Applying the Markov inequality and Lemma 2.3, we obtain

$$\begin{aligned} & P \left(\sup_{\mathbf{k} \geq \mathbf{0}} \left\| \sum_{\mathbf{1} \leq \mathbf{j}_i \leq \mathbf{n}_i} \sum_{j_i=n_i}^{n_i+k_i} \sum_{\mathbf{1} \leq \mathbf{j}'_i \leq \mathbf{n}'_i+\mathbf{k}'_i} X_{(\mathbf{j}_i, j_i, \mathbf{j}'_i)} \right\| > \varepsilon/d \right) \\ &\leq \frac{d^p}{\varepsilon^p} E \left(\sup_{\mathbf{k} \geq \mathbf{0}} \left\| \sum_{\mathbf{1} \leq \mathbf{j}_i \leq \mathbf{n}_i} \sum_{j_i=n_i}^{n_i+k_i} \sum_{\mathbf{1} \leq \mathbf{j}'_i \leq \mathbf{n}'_i+\mathbf{k}'_i} X_{(\mathbf{j}_i, j_i, \mathbf{j}'_i)} \right\|^p \right) \\ &\leq C \sum_{\mathbf{1} \leq \mathbf{j}_i \leq \mathbf{n}_i} \sum_{j_i=n_i}^{\infty} \sum_{\mathbf{j}'_i \geq \mathbf{1}} E \|X_{(\mathbf{j}_i, j_i, \mathbf{j}'_i)}\|^p. \end{aligned}$$

Then, using (3.1) or (3.3), we have that

$$\begin{aligned} P \left(\sup_{\mathbf{k} \geq \mathbf{0}} \|S_{\mathbf{n}+\mathbf{k}} - S_{\mathbf{n}}\| > \varepsilon \right) &\leq C \sum_{i=1}^d \sum_{\mathbf{1} \leq \mathbf{j}_i \leq \mathbf{n}_i} \sum_{j_i=n_i}^{\infty} \sum_{\mathbf{j}'_i \geq \mathbf{1}} E \|X_{(\mathbf{j}_i, j_i, \mathbf{j}'_i)}\|^p \\ &= C \sum_{\mathbf{j} \gg \mathbf{n}} E \|X_{\mathbf{j}}\|^p = o(1) \text{ as } \|\mathbf{n}\| \rightarrow \infty, \end{aligned}$$

which implies that $S_{\mathbf{n}}$ converges a.s as $\|\mathbf{n}\| \rightarrow \infty$ (by Lemma 2.6). Then $\sum_{\mathbf{n} \geq \mathbf{1}} X_{\mathbf{n}}$ converges a.s. Thus, the tail series $\{T_{\mathbf{n}} = \sum_{\mathbf{k} \gg \mathbf{n}} X_{\mathbf{k}}\}$ is a well-defined field of random variables.

Next, to prove that (3.1) implies (3.2), observe that for $K > 0$

$$\sup_{\mathbf{n} \geq \mathbf{1}} P \left(\frac{\sup_{\mathbf{k} \geq \mathbf{n}} \|T_{\mathbf{k}}\|}{b_{\mathbf{n}}} > K \right)$$

$$\begin{aligned}
 &= \sup_{\mathbf{n} \succeq \mathbf{1}} P \left(\sup_{\mathbf{k} \succeq \mathbf{n}} \left\| \sum_{\mathbf{i} \succeq \mathbf{k}} X_{\mathbf{i}} \right\| > K.b_{\mathbf{n}} \right) \\
 &= \sup_{\mathbf{n} \succeq \mathbf{1}} \lim_{\|\mathbf{N}\| \rightarrow \infty} P \left(\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{N}} \lim_{\|\mathbf{M}\| \rightarrow \infty} \left\| \sum_{\mathbf{k} \ll \mathbf{i} \preceq \mathbf{M}} X_{\mathbf{i}} \right\| > K.b_{\mathbf{n}} \right) \quad (\text{by Lemma 2.5}) \\
 &\leq \sup_{\mathbf{n} \succeq \mathbf{1}} \lim_{\|\mathbf{N}\| \rightarrow \infty} P \left(\lim_{\|\mathbf{M}\| \rightarrow \infty} \max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{N}} \left\| \sum_{\mathbf{k} \ll \mathbf{i} \preceq \mathbf{M}} X_{\mathbf{i}} \right\| > K.b_{\mathbf{n}} \right) \\
 &\leq \sup_{\mathbf{n} \succeq \mathbf{1}} \lim_{\|\mathbf{M}\| \rightarrow \infty} \liminf P \left(\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{N}} \left\| \sum_{\mathbf{k} \ll \mathbf{i} \preceq \mathbf{M}} X_{\mathbf{i}} \right\| > K.b_{\mathbf{n}} \right) \quad (\text{by Lemma 2.5}) \\
 &\leq \sup_{\mathbf{n} \succeq \mathbf{1}} \liminf_{\|\mathbf{M}\| \rightarrow \infty} P \left(\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{M}} \left\| \sum_{\mathbf{k} \ll \mathbf{i} \preceq \mathbf{M}} X_{\mathbf{i}} \right\| > K.b_{\mathbf{n}} \right) \\
 &\leq \sup_{\mathbf{n} \succeq \mathbf{1}} \liminf_{\|\mathbf{M}\| \rightarrow \infty} \left(P \left(\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{M}} \left\| \sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{k}} X_{\mathbf{i}} \right\| > \frac{K.b_{\mathbf{n}}}{2} \right) + P \left(\left\| \sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{M}} X_{\mathbf{i}} \right\| > \frac{K.b_{\mathbf{n}}}{2} \right) \right) \\
 &\leq 2 \sup_{\mathbf{n} \succeq \mathbf{1}} \lim_{\|\mathbf{M}\| \rightarrow \infty} P \left(\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{M}} \left\| \sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{k}} X_{\mathbf{i}} \right\| > \frac{K.b_{\mathbf{n}}}{2} \right) \\
 &\leq \sup_{\mathbf{n} \succeq \mathbf{1}} \frac{2^{p+1}}{K^p b_{\mathbf{n}}^p} \lim_{\|\mathbf{M}\| \rightarrow \infty} E \max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{M}} \left\| \sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{k}} X_{\mathbf{i}} \right\|^p \quad (\text{by the Markov inequality}) \\
 &\leq \sup_{\mathbf{n} \succeq \mathbf{1}} \frac{2^{p+1}}{K^p b_{\mathbf{n}}^p} \lim_{\|\mathbf{M}\| \rightarrow \infty} \sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{M}} E \|X_{\mathbf{i}}\|^p \quad (\text{by Lemma 2.3}) \\
 &= \sup_{\mathbf{n} \succeq \mathbf{1}} \frac{C}{K^p b_{\mathbf{n}}^p} \sum_{\mathbf{n} \ll \mathbf{i}} E \|X_{\mathbf{i}}\|^p \leq \frac{C}{K^p} \quad (\text{by (3.1)}) \rightarrow 0 \text{ as } K \rightarrow \infty.
 \end{aligned}$$

(2) The proof that (3.3) implies (3.4) is the same as that in (1). □

Remark 3.2. It should be noted that:

- In the case $d = 1$, Theorem 3.1 reduces to Corollary 2 in [9].
- The proof of Theorem 3.1 closely follows the pattern of Theorem 1 in [8].
- The primary mode of the convergence given by (3.4) of Theorem 3.1 was introduced in [4] in the case $d = 1$ for the tail series of a convergent series of random variables.

Example 4. Let $\{V_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ be a field independent, identically distributed, mean 0 random variables in a p -uniformly smooth Banach space \mathbb{E} ($1 \leq p \leq 2$) such that $E\|V_{\mathbf{1}}\|^p < \infty$, let $\{a_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ be a field of nonzero constants such that $\sum_{\mathbf{n} \succeq \mathbf{1}} |a_{\mathbf{n}}|^p < \infty$, set $X_{\mathbf{n}} = a_{\mathbf{n}}V_{\mathbf{n}}$ for $\mathbf{n} \succeq \mathbf{1}$. Then $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}; \mathbf{n} \succeq \mathbf{1}\}$ is

a field of martingale difference. By taking $b_n = (\sum_{i \gg n} |a_i|^p)^{1/p} \alpha_n^{-1}$, where $\{\alpha_n, n \succeq 1\}$ is any field of positive numbers, we have that

$$\frac{\sum_{i \gg n} E \|X_i\|^p}{b_n} = \alpha_n E \|X_1\|^p \rightarrow 0.$$

If $\sup_{n \succeq 1} \alpha_n < \infty$, then (3.1) holds, by Theorem 3.1, we have that

$$\sup_{k \succeq n} \|T_k\| = \mathcal{O}_P(b_n).$$

If $\alpha_n \rightarrow 0$ as $\|n\| \rightarrow \infty$, then (3.3) holds, by Theorem 3.1, we have that

$$\frac{\sup_{k \succeq n} \|T_k\|}{b_n} \xrightarrow{P} 0 \text{ as } \|n\| \rightarrow \infty.$$

Next, we establish the rate of convergence of series of strong* martingale difference fields, with the field of positive Borel function $\{\phi_n, n \succeq 1\}$ which have a property similar to that of the sequence of functions in [2] of Hong and Tsay, i.e.,

$$(3.5) \quad C_n \frac{u^{\lambda_n}}{v^{\lambda_n}} \leq \frac{\Phi_n(u)}{\Phi_n(v)} \leq D_n \frac{u^{\mu_n}}{v^{\mu_n}} \quad \text{for all } u \geq v > 0,$$

where $C_n \geq 1, D_n \geq 1, \lambda_n \geq 1, 0 < \mu_n \leq p$.

Note that the array of functions $\{\Phi_n, n \succeq 1\}$ with $\Phi_n(x) = x^p, p \geq 1$ satisfies the condition (3.5).

Theorem 3.3. *Let \mathbb{E} be a p -uniformly smooth Banach space for some $1 \leq p \leq 2$, let $\{X_n, \mathcal{F}_n; n \in \mathbb{N}^d\}$ be a field of \mathbb{E} -valued strong* martingale differences. Let $\{\Phi_n; n \succeq 1\}$ be a field of positive Borel functions which satisfies the conditions (3.5) and $\Phi_n(u) \leq \Phi_m(u)$ for all $n \ll m$. Let $\{b_n\}, \{B_n\}$ be fields of positive constants, such that $\Phi_n(b_n) = o(1), \Phi_n(B_n) = o(1)$ as $\|n\| \rightarrow \infty$.*

(1) *If*

$$(3.6) \quad \sum_{k \gg n} A_k E \Phi_k(\|X_k\|) = \mathcal{O}(\Phi_n(b_n)),$$

where $A_n = \max\{\frac{1}{C_n}, D_n\}$, then $\sum_{n \succeq 1} X_n$ converges a.s. and the series $\{T_n = \sum_{k \gg n} X_k\}$ satisfies the relation

$$(3.7) \quad \sup_{k \succeq n} \|T_k\| = \mathcal{O}_P(b_n).$$

(2) *If*

$$(3.8) \quad \sum_{k \gg n} A_k E \Phi_k(\|X_k\|) = o((\Phi_n(B_n)))$$

as $\|n\| \rightarrow \infty$ (where $A_n = \max\{\frac{1}{C_n}, D_n\}$), then $\sum_{n \succeq 1} X_n$ converges a.s. and the series $\{T_n = \sum_{k \gg n} X_k\}$ obeys the limit law

$$(3.9) \quad \frac{\sup_{k \succeq n} \|T_k\|}{B_n} \xrightarrow{P} 0 \text{ as } \|n\| \rightarrow \infty.$$

Proof. (1) For each $\mathbf{n} \succeq \mathbf{1}$, set $Y_{\mathbf{n}} = X_{\mathbf{n}}I(\|X_{\mathbf{n}}\| \leq 1)$, $Z_{\mathbf{n}} = X_{\mathbf{n}}I(\|X_{\mathbf{n}}\| > 1)$, $U_{\mathbf{n}} = Y_{\mathbf{n}} - E(Y_{\mathbf{n}}|\mathcal{F}_{\mathbf{n}}^*)$, $V_{\mathbf{n}} = Z_{\mathbf{n}} - E(Z_{\mathbf{n}}|\mathcal{F}_{\mathbf{n}}^*)$. $S_{\mathbf{n}}^1 = \sum_{\mathbf{k} \preceq \mathbf{n}} U_{\mathbf{k}}$, and $S_{\mathbf{n}}^2 = \sum_{\mathbf{k} \preceq \mathbf{n}} V_{\mathbf{k}}$. Then $X_{\mathbf{n}} = U_{\mathbf{n}} + V_{\mathbf{n}}$ and $S_{\mathbf{n}} = S_{\mathbf{n}}^1 + S_{\mathbf{n}}^2$. Moreover, since $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ is a field of strong* martingale differences, it is clear that $\{U_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ and $\{V_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ are strong adapted fields.

By the proof of Theorem 3.1, we have

$$\begin{aligned} P\left(\sup_{\mathbf{k} \succeq \mathbf{0}} \|S_{\mathbf{n}+\mathbf{k}}^1 - S_{\mathbf{n}}^1\| > \varepsilon\right) &= C \sum_{\mathbf{i} \gg \mathbf{n}} E\|Y_{\mathbf{i}}\|^p \leq C \sum_{\mathbf{i} \gg \mathbf{n}} E\|Y_{\mathbf{i}}\|^{\mu_{\mathbf{n}}} \\ &\leq C \sum_{\mathbf{i} \gg \mathbf{n}} D_{\mathbf{i}} \cdot E \frac{\Phi_{\mathbf{i}}(\|Y_{\mathbf{i}}\|)}{\Phi_{\mathbf{i}}(1)} \leq C \sum_{\mathbf{i} \gg \mathbf{n}} A_{\mathbf{i}} \frac{E\Phi_{\mathbf{i}}(\|X_{\mathbf{i}}\|)}{\Phi_{\mathbf{i}}(1)} \\ &\leq C \frac{1}{\Phi_{\mathbf{1}}(1)} \sum_{\mathbf{i} \gg \mathbf{n}} A_{\mathbf{i}} E\Phi_{\mathbf{n}}(\|X_{\mathbf{i}}\|) < o(1) \text{ as } \|\mathbf{n}\| \rightarrow \infty. \end{aligned}$$

Then $S_{\mathbf{n}}^1$ converges a.s. as $\|\mathbf{n}\| \rightarrow \infty$ (by Lemma 2.6). Next, by the proof of Theorem 3.1, we have

$$\begin{aligned} P\left(\sup_{\mathbf{k} \succeq \mathbf{0}} \|S_{\mathbf{n}+\mathbf{k}}^2 - S_{\mathbf{n}}^2\| > \varepsilon\right) &= C \sum_{\mathbf{i} \gg \mathbf{n}} E\|Z_{\mathbf{i}}\| \leq C \sum_{\mathbf{i} \gg \mathbf{n}} E\|Z_{\mathbf{i}}\|^{\lambda_{\mathbf{n}}} \\ &\leq C \sum_{\mathbf{i} \gg \mathbf{n}} \frac{1}{C_{\mathbf{i}}} \cdot E \frac{\Phi_{\mathbf{i}}(\|Z_{\mathbf{i}}\|)}{\Phi_{\mathbf{i}}(1)} \leq C \sum_{\mathbf{i} \gg \mathbf{n}} A_{\mathbf{i}} \frac{E\Phi_{\mathbf{i}}(\|X_{\mathbf{i}}\|)}{\Phi_{\mathbf{i}}(1)} \\ &\leq C \frac{1}{\Phi_{\mathbf{1}}(1)} \sum_{\mathbf{i} \gg \mathbf{n}} A_{\mathbf{i}} E\Phi_{\mathbf{n}}(\|X_{\mathbf{i}}\|) < o(1) \text{ as } \|\mathbf{n}\| \rightarrow \infty. \end{aligned}$$

Then $S_{\mathbf{n}}^2$ converges a.s. as $\|\mathbf{n}\| \rightarrow \infty$ (by Lemma 2.6). By $S_{\mathbf{n}} = S_{\mathbf{n}}^1 + S_{\mathbf{n}}^2$, which implies $S_{\mathbf{n}}$ converges a.s as $\|\mathbf{n}\| \rightarrow \infty$, then $\sum_{\mathbf{n} \succeq \mathbf{1}} X_{\mathbf{n}}$ converges a.s. Thus, the tail series $\{T_{\mathbf{n}} = \sum_{\mathbf{k} \gg \mathbf{n}} X_{\mathbf{k}}; \mathbf{n} \succeq \mathbf{1}\}$ is a well-defined field of random variables.

Next, to prove that (3.6) implies (3.7), for each $\mathbf{n} \succeq \mathbf{1}$, and $\mathbf{n} \succeq \mathbf{i}$, set $Y'_{\mathbf{i}} = X_{\mathbf{i}}I(\|X_{\mathbf{i}}\| \leq b_{\mathbf{n}})$, $Z'_{\mathbf{i}} = X_{\mathbf{i}}I(\|X_{\mathbf{i}}\| > b_{\mathbf{n}})$, $U'_{\mathbf{n}} = Y'_{\mathbf{n}} - E(Y'_{\mathbf{n}}|\mathcal{F}_{\mathbf{n}}^*)$, $V'_{\mathbf{n}} = Z'_{\mathbf{n}} - E(Z'_{\mathbf{n}}|\mathcal{F}_{\mathbf{n}}^*)$. Then $X_{\mathbf{n}} = U'_{\mathbf{n}} + V'_{\mathbf{n}}$. Moreover, by $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ being a strong* martingale difference field, then it is clear that $\{U'_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ and $\{V'_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}, \mathbf{n} \succeq \mathbf{1}\}$ are strong adapted fields.

By the proof of Theorem 3.1, we observe that for $K > 0$,

$$\begin{aligned} &\sup_{\mathbf{n} \succeq \mathbf{1}} P\left(\frac{\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_{\mathbf{k}}\|}{b_{\mathbf{n}}} > K\right) \\ &\leq 2 \sup_{\mathbf{n} \succeq \mathbf{1}} \liminf_{\|\mathbf{M}\| \rightarrow \infty} P\left(\frac{\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{M}} \|\sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{k}} X_{\mathbf{i}}\|}{b_{\mathbf{n}}} > \frac{K}{2}\right) \\ &\leq 2 \sup_{\mathbf{n} \succeq \mathbf{1}} \lim_{\|\mathbf{M}\| \rightarrow \infty} \left(P\left(\frac{\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{M}} \|\sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{k}} U'_{\mathbf{i}}\|}{b_{\mathbf{n}}} > \frac{K}{4}\right) \right. \\ &\quad \left. + P\left(\frac{\max_{\mathbf{n} \preceq \mathbf{k} \preceq \mathbf{M}} \|\sum_{\mathbf{n} \ll \mathbf{i} \preceq \mathbf{k}} V'_{\mathbf{i}}\|}{b_{\mathbf{n}}} > \frac{K}{4}\right) \right) \end{aligned}$$

$$\begin{aligned}
 &\leq 2 \sup_{n \geq 1} \lim_{M \rightarrow \infty} \left(\frac{4^p}{K^p b_n^p} E \left\{ \max_{n \leq k \leq M} \left\| \sum_{n \ll i \leq k} U'_i \right\|^p \right\} \right. \\
 &\quad \left. + \frac{4}{K b_n} E \left\{ \max_{n \leq k \leq M} \left\| \sum_{n \ll i \leq k} V'_i \right\| \right\} \right) \\
 &\leq 2 \sup_{n \geq 1} \lim_{\|M\| \rightarrow \infty} \left(\frac{4^p}{K^p b_n^p} \sum_{n \ll i \leq M} E \|U'_i\|^p \right. \\
 &\quad \left. + \frac{4}{K b_n} \sum_{n \ll i \leq M} E \|V'_i\| \right) \quad (\text{by Lemma 2.3}) \\
 &\leq C \sup_{n \geq 1} \left(\frac{1}{K^p} \sum_{n \ll i} \frac{E \|Y'_i\|^p}{b_n^p} + \frac{1}{K} \sum_{n \ll i} \frac{E \|Z'_i\|}{b_n} \right) \\
 &\leq C \sup_{n \geq 1} \left(\frac{1}{K^p} \sum_{n \ll i} \frac{E \|Y'_i\|^{\mu_n}}{b_n^{\mu_n}} + \frac{1}{K} \sum_{n \ll i} \frac{E \|Z'_i\|^{\lambda_n}}{b_n^{\lambda_n}} \right) \\
 &\leq C \sup_{n \geq 1} \left(\frac{1}{K^p} \sum_{i \gg n} D_i \cdot E \frac{\Phi_i(\|Y'_i\|)}{\Phi_i(b_n)} + \frac{1}{K} \sum_{i \gg n} \frac{1}{C_i} \cdot E \frac{\Phi_i(\|Z'_i\|)}{\Phi_i(b_n)} \right) \\
 &\leq C \sup_{n \geq 1} \frac{1}{\Phi_n(b_n)} \left(\frac{1}{K^p} + \frac{1}{K} \right) \sum_{i \gg n} A_i E \Phi_n(\|X_i\|) < o(1) \text{ as } K \rightarrow \infty.
 \end{aligned}$$

(2) The proof that (3.8) implies (3.9) is the same as that in (1). □

When $d = 1$, by Remark 2.1, we have the following corollary.

Corollary 3.3.1. *Let \mathbb{E} be a p -uniformly smooth Banach space for some $1 \leq p \leq 2$, $\{X_n, \mathcal{F}_n; n \in \mathbb{N}\}$ be a sequence of \mathbb{E} -valued martingale differences. Let $\{\Phi_n; n \geq 1\}$ be a sequence of positive Borel functions which satisfies the following two conditions*

$$C_n \frac{u^{\lambda_n}}{v^{\lambda_n}} \leq \frac{\Phi_n(u)}{\Phi_n(v)} \leq D_n \frac{u^{\mu_n}}{v^{\mu_n}} \text{ for all } u \geq v > 0,$$

where $C_n \geq 1, D_n \geq 1, \lambda_n \geq 1, 0 < \mu_n \leq p$,

$$\Phi_n(u) \leq \Phi_m(u) \text{ for all } n > m.$$

Let $\{b_n\}, \{B_n\}$ be sequences of positive constants, such that $\Phi_n(b_n) = o(1), \Phi_n(B_n) = o(1)$.

(1) If

$$\sum_{k \geq n+1} A_k E \Phi_k(\|X_k\|) = \mathcal{O}(\Phi_n(b_n)),$$

then $\sum_{n \geq 1} X_n$ converges a.s. and

$$\sup_{k \geq n+1} \|T_k\| = \mathcal{O}_P(b_n),$$

where $T_n = \sum_{k \geq n+1} X_k$.

(2) If

$$\sum_{k \geq n+1} A_k E\Phi_k(\|X_k\|) = o((\Phi_n(B_n))),$$

then $\sum_{n \geq 1} X_n$ converges a.s. and

$$\frac{\sup_{k \geq n+1} \|T_k\|}{B_n} \xrightarrow{P} 0,$$

where $T_n = \sum_{k \geq n+1} X_k$.

Remark 3.4. Let $\{X_n, \mathcal{F}_n, n \geq 1\}$ be a sequence of real-valued independent random variables with $EX_n = 0, n \geq 1$. Let $\{g_n(x), n \geq 1\}$ be a sequence of functions defined on $[0, \infty)$ such that

$$0 \leq g_n(0) \leq g_n(x), 0 < g_n(x) \uparrow \infty \text{ as } n \uparrow \infty \text{ for each } x > 0$$

and

$$\frac{g_n(x)}{x} \uparrow, \frac{g_n(x)}{x^p} \downarrow \text{ on } (0, \infty), n \geq 1, \text{ for some } 1 < p \leq 2.$$

In Corollary 3.3.1, taking $\Phi_n = g_n$ for all $n \geq 1$, with $\lambda_n = 1, \mu_n = p, C_n = 1, D_n = 1, n \geq 1$, we obtain Theorem 2 in [11].

Finally, we establish the rate of complete convergence of the tail series of martingale difference fields.

Theorem 3.5. Let \mathbb{E} be a p -uniformly smooth Banach space for some $1 \leq p \leq 2$, $\{X_n, \mathcal{F}_n; \mathbf{n} \in \mathbb{N}^d\}$ be a field of \mathbb{E} -valued martingale differences. Let $\{a_n\}$ be field of positive constants, such that either $a_n \leq a_m$ for all $\mathbf{n} \preceq \mathbf{m}$ or $a_n \geq a_m$ for all $\mathbf{n} < \mathbf{m}$ and $\sup_n a_{2^n}/a_{2^{n+1}} \leq M < \infty$. If

$$(3.10) \quad \sum_{\mathbf{n} \succeq \mathbf{1}} \varphi(\mathbf{n}) E\|X_n\|^p < \infty,$$

where $\varphi(\mathbf{n}) = \sum_{2^k \ll \mathbf{n}} \frac{1}{a_{2^k}}$, then for all $\varepsilon > 0$,

$$(3.11) \quad \sum_{\mathbf{n} \succeq \mathbf{1}} \frac{1}{|\mathbf{n}|} P(\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_k\| > \varepsilon a_n) < \infty.$$

Proof. For all $\mathbf{n} \succeq \mathbf{2}$ then $\varphi(\mathbf{n}) \geq \frac{1}{b_2} > 0$, so $\sum_{\mathbf{k} \gg \mathbf{n}} E\|X_k\|^p = o(1)$ as $\|\mathbf{n}\| \rightarrow \infty$. By proof of Theorem 3.1, we have $\sum_{\mathbf{n} \succeq \mathbf{1}} X_n$ converges a.s. Thus, the tail series $\{T_n = \sum_{\mathbf{k} \gg \mathbf{n}} X_k; \mathbf{n} \succeq \mathbf{1}\}$ is a well-defined field of random variables.

Next, to prove that (3.10) implies (3.11), we note that

$$\sum_{\mathbf{n} \succeq \mathbf{1}} \frac{1}{|\mathbf{n}|} P(\sup_{\mathbf{k} \succeq \mathbf{n}} \|T_k\| > \varepsilon a_n) = \sum_{\mathbf{n} \succeq \mathbf{0}} \sum_{2^i \preceq \mathbf{n} \preceq 2^{i+1}} \frac{1}{|\mathbf{i}|} P(\sup_{\mathbf{k} \succeq \mathbf{i}} \|T_k\| > \varepsilon a_i)$$

$$\leq \sum_{\mathbf{n} \geq \mathbf{0}} P(\sup_{\mathbf{k} \geq 2\mathbf{n}} \|T_{\mathbf{k}}\| > \frac{1}{M} \cdot \varepsilon a_{2\mathbf{n}}).$$

Applying Lemma 2.3, Lemma 2.5, the Markov inequality and the same argument as the proof of Theorem 3.1, we see

$$\begin{aligned} \sum_{\mathbf{n} \geq \mathbf{1}} \frac{1}{|\mathbf{n}|} P(\sup_{\mathbf{k} \geq \mathbf{n}} \|T_{\mathbf{k}}\| > \varepsilon b_{\mathbf{n}}) &= \sum_{\mathbf{n} \geq \mathbf{0}} \frac{M^p}{\varepsilon^p \cdot a_{2\mathbf{n}}^p} \sum_{\mathbf{i} \gg 2\mathbf{n}} E\|X_{\mathbf{i}}\|^p \\ &\leq C \sum_{\mathbf{n} \geq \mathbf{1}} \varphi(\mathbf{n}) \|X_{\mathbf{n}}\|^p < \infty. \end{aligned} \quad \square$$

Corollary 3.5.1. *Let \mathbb{E} be a p -uniformly smooth Banach space for some $1 \leq p \leq 2$, $\{X_{\mathbf{n}}, \mathcal{F}_{\mathbf{n}}; \mathbf{n} \in \mathbb{N}^d\}$ be a field of \mathbb{E} -valued martingale differences. If*

$$(3.12) \quad \sum_{\mathbf{n} \geq \mathbf{1}} E\|X_{\mathbf{n}}\|^p < \infty,$$

then for all $\alpha > 0, \varepsilon > 0$,

$$(3.13) \quad \sum_{\mathbf{n} \geq \mathbf{1}} \frac{1}{|\mathbf{n}|} P(\sup_{\mathbf{k} \geq \mathbf{n}} \|T_{\mathbf{k}}\| > \varepsilon |\mathbf{n}|^\alpha) < \infty.$$

Proof. Put $a_{\mathbf{n}} = |\mathbf{n}|^\alpha$ then $\varphi(\mathbf{n}) \leq \sum_{\mathbf{n} \geq \mathbf{1}} \frac{1}{|2\alpha\mathbf{n}|} = (\sum_{n=1}^\infty \frac{1}{2^{an}})^d < \infty$ so (3.12) implies (3.10). By Theorem 3.5 we get (3.13). \square

Corollary 3.5.2. *Let \mathbb{E} be a p -uniformly smooth Banach space for some $1 \leq p \leq 2$, $\{X_n, \mathcal{F}_n; n \in \mathbb{N}\}$ be a sequence of \mathbb{E} -valued martingale differences. If*

$$(3.14) \quad \sum_{n=1}^\infty E\|X_n\|^p \log_2 n < \infty,$$

then for all $\alpha > 0, \varepsilon > 0$,

$$(3.15) \quad \sum_{n=1}^\infty \frac{1}{n} P(\sup_{k \geq n} \|T_k\| > \varepsilon) < \infty.$$

Proof. Put $a_n = 1$ then for $d = 1$ we have $\varphi(n) \leq \log_2 n$. Hence (3.14) implies (3.10). By Theorem 3.5 we obtain (3.15). \square

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