

VECTOR-VALUED TREE MARTINGALES

TONG-JUN HE, QUAN YI AND YI SHEN

ABSTRACT. In this paper, we first introduce some interesting nontrivial tree martingale examples. Second, we show that, if $1 < a < 2$, then X -valued predictable tree martingale operators $S_t^{(a)}(f)$, $\sigma_t^{(a)}(f)$ can be dominated by X -valued predictable tree martingales f in a certain quasi-norm provided the space X is isomorphic to a $2a$ -uniformly convex Banach space.

1. Preliminaries and definitions.

Definition 1.1. Let $\mathbf{T}$ be a countable, upward directed index set with respect to the partial ordering $\preceq$ satisfying the following two conditions:

- (1) for every $t \in \mathbf{T}$, the set $\mathbf{T}^t := \{u \in \mathbf{T} : u \preceq t\}$ is finite;
- (2) for every $t \in \mathbf{T}$, the set $\mathbf{T}_t := \{u \in \mathbf{T} : t \preceq u\}$ is linearly ordered.

Thus $\mathbf{T}$ is a tree set and every nonempty subset of $\mathbf{T}$ has at least one minimum. The succeeding element of $t \in \mathbf{T}$, namely, the minimum element of the set $\mathbf{T}_t - \{t\}$, is denoted by t^+ . A tree $\mathbf{T}$ is also a special partially ordered set with respect to the partial ordering $\preceq$.

Let $(\Omega, \mathcal{S}, \mu)$ be a measure space $(\Omega, \mathcal{S})$, equipped with a finite measure μ , and let $L^1(\Omega, \mathcal{S}, \mu)$ be the space of integrable functions that are measurable relative to $\mathcal{S}$. With the help of positive contractive projections in $L^1(\Omega, \mathcal{S}, \mu)$ spaces, the definition of tree martingales shall be given as follows:

Definition 1.2. Let $(P_t, t \in \mathbf{T})$ be a family of positive contractive projections in $L^1(\Omega, \mathcal{S}, \mu)$ spaces. Then for every $f \in L^1(\Omega, \mathcal{S}, \mu)$, the

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family $(P_t f, t \in \mathbf{T})$ is called *tree martingales* if $s \preceq t$,

$$(1.1) \quad P_t P_s = P_s P_t = P_s$$

for every pair of comparable elements (s, t) (that is, $s \preceq t$ or $s \succeq t$) in $\mathbf{T}$, and

$$(1.2) \quad P_t P_s = P_s P_t = 0$$

for every pair of non-comparable elements (s, t) (that is, $s \not\preceq t$ or $s \not\succeq t$) in $\mathbf{T}$.

It is clear that Definition 1.2 is equivalent to the definition of tree martingales introduced by Weisz [19] when the value space X of tree martingales is a scalar space. A set $\mathbf{T}_0$ is defined by

$$(1.3) \quad \mathbf{T}_0 \stackrel{\text{def}}{=} \{t_0 \in \mathbf{T} : \text{either } t_0 \preceq t \text{ or } t_0 \text{ is non-comparable with } t \text{ for every } t \in \mathbf{T}\}$$

for every nonempty tree set $\mathbf{T}$, then $\mathbf{T}_0$ is nonempty because of every nonempty subset of the tree set $\mathbf{T}$ having at least one minimum. Next, an element t_0^- is defined for which $t_0^- \prec t_0$ for every $t_0 \in \mathbf{T}_0$, and suppose $P_{t_0^-}$ is a common projection (i.e., $P_{t_0^-} = 0$) for which $P_{t_0} P_{t_0^-} = P_{t_0^-} P_{t_0} = P_{t_0^-}$. In this case, it is clear that $(t_0^-)^+ = t_0$. Let

$$(1.4) \quad \mathbf{T}_0^- \stackrel{\text{def}}{=} \{t_0^- : t_0^- \preceq t_0 \text{ for every } t_0 \in \mathbf{T}_0\} \text{ and } \overline{\mathbf{T}} \stackrel{\text{def}}{=} \mathbf{T} \cup \mathbf{T}_0^-.$$

Obviously, $\overline{\mathbf{T}}$ is still a tree set. That is, the common sigma algebras $(\mathcal{F}_{t_0^-}, t_0^- \in \mathbf{T}_0^-)$ of tree martingales can be defined as well as that of one-parameter martingales. Note that every common σ -algebra $\mathcal{F}_{t_0^-}$ only contains one element $\emptyset$, where every common σ -algebra $\mathcal{F}_{t_0^-}$ is replaced by a corresponding common projection $P_{t_0^-}$.

As one-parameter predictable martingales, we can define predictable tree martingales as follows:

Definition 1.3. We say that an X -valued tree martingale $Pf = (P_t f, t \in \mathbf{T})$ is predictable if there exists a family of $\lambda = (\lambda_t, t \in \mathbf{T})$ of nondecreasing, nonnegative and predictable functions such that

$$\|P_t f\| \leq \lambda_t \quad (t \in \mathbf{T}),$$

where nondecreasing functions mean that, for any comparable elements $s, t \in \mathbf{T}$, if $s \leq t$ then $\lambda_s \leq \lambda_t$; predictable functions mean that, for any $t \in \mathbf{T}$, λ_{t+} is $\mathcal{F}_t$ -measurable. Such a λ is called a prediction belonging to f .

Inspired by the Vilenkin system [6, 12, 18, 20], tree martingales and tree martingale transformation have been introduced by Schipp and Weisz. We refer to Weisz's exposition [19] and the successive papers on tree martingales [7, 9–11]. In the 1980's, Weisz and Schipp [17, 19] showed that Burkholder-Gundy's inequality of tree martingales holds if $2 < p < \infty$. Moreover, Weisz proved that the partial sums of the Vilenkin-Fourier series of an integrable function can be dominated by the maximal function of a suitable tree martingale transforms. However, because of the fact that tree martingale transforms cannot be defined as a one-parameter martingale and stopping times cannot be introduced for tree martingales, the study of tree martingales faces several difficulties. Weisz and Schipp [17, 19] obtained some results by using convexity methods. The question is: Are there efficient ways to overcome these difficulties? Here we study this problem.

By using graph-theoretic tricks, He and Shen [10] established a theorem on the structure of the index set $\mathbf{T}$ of tree martingales. For a tree $\mathbf{T}(V(\mathbf{T}), E(\mathbf{T}))$, the set of the vertex of $\mathbf{T}$ is denoted by $V(\mathbf{T})$, and the set of the arcs of $\mathbf{T}$ is denoted by $E(\mathbf{T})$.

Lemma 1.4. *A tree set $\mathbf{T}$ is isomorphic to a directed infinite locally finite forest. That is,*

$$\bigoplus_{i \in \mathbf{Z}^+} \{\mathbf{T}_i\} = \mathbf{T} \cong \mathbf{T}(V(\mathbf{T}), E(\mathbf{T})) = \bigoplus_{i \in \mathbf{Z}^+} \{\mathbf{T}_i(V(\mathbf{T}_i), E(\mathbf{T}_i))\},$$

$$\mathbf{T}_i \cong \mathbf{T}_i(V(\mathbf{T}_i), E(\mathbf{T}_i)), \quad i \in \mathbf{Z}^+,$$

where every $\mathbf{T}_i$ is a locally finite tree set, and $\mathbf{T}_i(V(\mathbf{T}_i), E(\mathbf{T}_i))$ is a directed infinite locally finite tree with root.

Based on Lemma 1.4, He and Shen [10] further established a decomposition of tree martingales as follows:

Lemma 1.5. *Let $Pf = (P_t f, t \in \mathbf{T})$ be a tree martingale. Then*

$$(1.5) \quad Pf = \bigoplus_{i=0}^{+\infty} P_i f,$$

where $P_i f = (P_t f, t \in \mathbf{T}_i)$ is a locally finite tree martingale and $\{\mathbf{T}_i\}_{i=0}^{\infty}$ is a sequence of locally finite tree sets for which

$$\mathbf{T} = \bigoplus_{i=0}^{+\infty} \mathbf{T}_i.$$

Lemma 1.5 shows that some inequalities for tree martingales can be obtained for locally finite tree martingales. Throughout this paper, we always assume that tree set $\mathbf{T}$ is a locally finite tree with root.

In [7], He characterized the collection of σ -filtrations $(\mathcal{F}_t, t \in \mathbf{T})$ on X -valued tree martingales by the positive contractive projections when the tree set $\mathbf{T}$ is a locally finite tree and X is a scalar-valued space. However, when X is a general Banach space and the tree set $\mathbf{T}$ is a locally finite tree, it is difficult to characterize the collection of σ -filtrations $(\mathcal{F}_t, t \in \mathbf{T})$ on X -valued tree martingales by the positive contractive projections, because the structure of contractive projections on a Lebesgue-Bochner space of X -valued functions $L^1(X)$ is a close connection to the geometric structure of the Banach space X (the structure of contractive projections is considerably complicated without any assumptions of σ -finiteness of $L^p(X)$ -space or separability of the Hilbert space X (see [16])). Here, for simplicity, we assume directly that $(\Omega, \mathcal{F}_{\infty}, P)$ is a complete probability space and

$$(1.6) \quad \mathcal{F}_{\infty} \stackrel{\text{def}}{=} \bigvee_{t \in \mathbf{T}} \mathcal{F}_t = \sigma \left(\bigcup_{t \in \mathbf{T}} \mathcal{F}_t \right),$$

where $\mathcal{F} = (\mathcal{F}_t, t \in \mathbf{T})$ is a family of non-decreasing sub σ -algebras of $\mathcal{F}_{\infty}$ with respect to the partial ordering $\preceq$. Throughout this paper, unless otherwise stated, we let E_t be the conditional expectation operator with respect to $\mathcal{F}_t$, $(X, \|\cdot\|)$ a Banach space, $L_1(X)$ the space of Bochner integrable measurable functions, any $A \in \mathcal{F}_{\infty}$, and the indicator function of a set A is denoted by χ_A .

Let $(\phi_t, t \in \mathbf{T})$ be a family of scalar complex-valued measurable functions with $|\phi| = 1$, and

$$(1.7) \quad P_t f = \phi_t E_t(\overline{f \phi_t}), \quad f \in L_1(X)$$

for each $t \in \mathbf{T}$. Then, from Definition 1.2, we see that $(P_t f, t \in \mathbf{T})$ is a family of tree martingales.

For an X -valued tree martingale, we are going to introduce a quasi-norm $\|\cdot\|_{\mathbf{M}^{pq}}$. Let $f = (f_t, t \in \mathbf{T})$ be a family of $\mathcal{F}_\infty$ -measurable functions (not necessarily a tree martingale) defined on the complete probability space $(\Omega, \mathcal{F}_\infty, P)$. For any $y \geq 0$, we set

$$\nu_y^f = \inf \{t \in \mathbf{T} : \|f_t\| > y\}.$$

Then it is easy to see that

$$\{t \in \nu_y^f\} = \{\omega \in \Omega : \|f_t(\omega)\| > y, \|f_s(\omega)\| \leq y, \text{ for all } s < t\},$$

where $s < t$ means that $s \leq t$ but $s \neq t$. Note that ν_y^f is generally a subset of $\mathbf{T}$, and if $\mathbf{T}$ is a total set, then ν_y^f is a stopping time; but if $\mathbf{T}$ is a partial set, then ν_y^f is not a stopping time since, for a fixed $y > 0$, ν_y^f is not the only one. Now, we may introduce the definition of quasi-norm $\|\cdot\|_{\mathbf{M}^{pq}}$ by ν_y^f . For $0 < p, q < \infty$, let

$$(1.8) \quad \|f\|_{\mathbf{M}^{pq}} = \sup_{y>0} y \left(\int_{\Omega} \left(\sum_{t \in \mathbf{T}} \chi_{\{t \in \nu_y^f\}} \right)^{p/q} \right)^{1/p},$$

$$\mathbf{M}^{pq} = \{f = (f_t, t \in \mathbf{T}) : \|f\|_{\mathbf{M}^{pq}} < \infty\},$$

where $\chi_{\{t \in \nu_y^f\}}$ is the indicator function of the set $\{t \in \nu_y^f\}$.

Remark 1.6. Note that, for each fixed family of $f = (f_t, t \in \mathbf{T})$, the quasi-norm $\|f\|_{\mathbf{M}^{pq}}$ is decreasing with q increasing and increasing with p increasing. Therefore, the limit does exist as $q \rightarrow \infty$ and satisfies

$$\|f\|_{\mathbf{M}^{p\infty}} = \lim_{q \rightarrow \infty} \|f\|_{\mathbf{M}^{pq}} = \sup_{y>0} y P(f^* > y)^{1/p}, \quad 0 < p < \infty.$$

In [19], it is verified that the map $\|\cdot\|_{\mathbf{M}^{pq}}$ is a quasi-norm, namely, for any two families of functions $f = (f_t, t \in \mathbf{T})$ and $g = (g_t, t \in \mathbf{T})$ and for any $\lambda \in C$,

$$\begin{aligned}\|\lambda f\|_{\mathbf{M}^{pq}} &= |\lambda| \|f\|_{\mathbf{M}^{pq}}, \\ \|f + g\|_{\mathbf{M}^{pq}} &\leq K_{pq}(\|f\|_{\mathbf{M}^{pq}} + \|g\|_{\mathbf{M}^{pq}}),\end{aligned}$$

where $0 < p < \infty$, $0 < q \leq \infty$ and K_{pq} depends only on p and q . Moreover, the map $\|\cdot\|_{\mathbf{M}^{pq}}$ is nondecreasing in the following sense: if, for all $t \in \mathbf{T}$, the inequality $\|f_t\| \leq \|g_t\|$ holds, then

$$\|f\|_{\mathbf{M}^{pq}} \leq \|g\|_{\mathbf{M}^{pq}}, \quad 0 < p < \infty, \quad 0 < q \leq \infty.$$

Remark 1.7. Note that if $\mathbf{T}$ is linearly ordered, then the sets $\{t \in \nu_y^f\}$ ($t \in \mathbf{T}$) are pairwise disjoint and $\sum_{t \in \mathbf{T}} \chi_{\{t \in \nu_y^f\}} = \chi_{\{f^* > y\}}$; in this case,

$$\|f\|_{\mathbf{M}^{pq}} = \sup_{y>0} y P(f^* > y)^{1/p}, \quad 0 < p < \infty, \quad 0 < q \leq \infty.$$

Denote by $\mathbf{P}^{pq}$ the space of this kind of X -valued predictable tree martingale and endow it with the following quasi-norm:

$$\|f\|_{\mathbf{P}^{pq}} = \inf \|\lambda\|_{\mathbf{M}^{pq}} \quad (0 < p < \infty, \quad 0 < q \leq \infty),$$

where the infimum is taken over all predictions $\lambda \in \mathbf{M}^{pq}$ belonging to f . For the reader's convenience, throughout this paper, we always let

$$\begin{aligned}\|(S_t^{(a)}(f), t \in T)\|_{\mathbf{M}^{pq}} &= \|S^{(a)}(f)\|_{\mathbf{M}^{pq}}, \\ \|(s_t^{(a)}(f), t \in T)\|_{\mathbf{M}^{pq}} &= \|s^{(a)}(f)\|_{\mathbf{M}^{pq}}.\end{aligned}$$

Next, some interesting tree martingale examples will be introduced in Section 2.

2. Some examples of tree martingales.

Example 2.1. Dyadic intervals are defined by

$$I_n(j) = (j2^{n-1}, (j+1)2^{n-1}], \quad j = 0, 1, 2, \dots, 2^{n-1} - 1, \quad n \geq 1;$$

furthermore, we define

$$\mathcal{I} = \{I_n(j), j = 0, 1, 2, \dots, 2^{n-1} - 1, n \geq 1\}.$$

Then the ordering $\preceq$ in $\mathcal{I}$ is defined by set inclusion and it is clear that the ordering $\preceq$ is a partial ordering. Define

$$\mathcal{F}_{I_n(j)} = \sigma\left(\bigcup_{\substack{I \subseteq I_n(j) \\ I \in \mathcal{I}}} I\right), \quad j = 0, 1, 2, \dots, 2^{n-1} - 1, n \geq 1.$$

Obviously, $(E(f|\mathcal{F}_{I_n(j)}), I_n(j) \in \mathcal{I})$ is a family of tree martingales for any $f \in L^1(X)$.

Example 2.2. Let $m = (m_k, k \in \mathbf{N})$ be a sequence of integers, each of them not less than 2. Let Z_{m_k} denote the m_k th discrete cyclic group. Z_{m_k} can be represented by the set $\{0, 1, \dots, m_k - 1\}$, where the group operation is the mod m_k addition and every subset is open. The measure on Z_{m_k} is defined such that the measure of every singleton is $(k \in \mathbf{N})/m_k$. By means of these cyclic groups, we define the so-called Vilenkin group G_m as the complete direct product of these Z_{m_k} , i.e.,

$$(2.1) \quad G_m = \times_{k=0}^{\infty} Z_{m_k}.$$

Then G_m is a compact Abelian group with Haar measure 1, whose elements are of the form $(x_k, k \in \mathbf{N})$, with $x_k \in Z_{m_k} (k \in \mathbf{N})$. The group operation on G_m (denoted by $\dot{+}$) is the coordinatewise addition (the inverse operation is denoted by $\dot{-}$), the measure (denoted by μ) and the topology are the product measure and topology. Consequently, G_m is a compact Abelian group. If $\sup_{n \in \mathbf{N}} m_n < \infty$, then we call G_m a bounded Vilenkin group. If the generating sequence m is not bounded, then G_m is said to be an unbounded Vilenkin group. A base for the neighborhoods of G_m can be given as follows: for $x \in G_m, n \in \mathbf{N} - \{0\}$,

$$(2.2) \quad \begin{aligned} I_0(x) &= G_m, \\ I_n(x) &= \{y = (y_k, k \in \mathbf{N}) \in G_m : y_k = x_k \text{ for } k < n\}. \end{aligned}$$

The set of intervals on G_m is denoted by

$$(2.3) \quad \mathcal{I} = \{I_n(x) : n \in \mathbf{N}, x \in G_m\}.$$

The Haar measure of $I_n(t)$ is M_n , where the generalized powers $M_n(n \in \mathbf{N})$ are defined in the following way:

$$(2.4) \quad M_0 = 1, \quad M_n = \prod_{j=0}^{n-1} m_j \quad (n \in \mathbf{N} - \{0\}).$$

Then each natural number n can be uniquely expressed as

$$(2.5) \quad n = \sum_{j=0}^{\infty} n_j M_j, \quad 0 \leq n_j < m_j, \quad (n_j \in Z_{m_j}, j \in \mathbf{N}),$$

where only a finite number of n 's may differ from zero. Notice that, under the identification of G_m with $[0, 1)$, the interval $I_n(j)$ corresponds to the interval $[jM_n^{-1}, (j+1)M_n^{-1})$ provided $0 \leq j < M_n$. Let

$$\begin{aligned} \mathcal{J} &= \{I_n(j), 0 \leq j < M_n, n \geq 1\}, \\ \mathcal{F}_{I_n(j)} &= \sigma\left(\bigcup_{\substack{J \subseteq I_n(j) \\ J \in \mathcal{J}}} J\right), \quad 0 \leq j < M_n, \quad n \geq 1. \end{aligned}$$

Obviously, $(E(f|\mathcal{F}_{I_n(j)}), I_n(j) \in \mathcal{J})$ is a family of tree martingales for any $f \in L^1(X)$. When $m_k = 2(k \in \mathbf{N})$, $(E(f|\mathcal{F}_{I_n(j)}), I_n(j) \in \mathcal{J})$ is a family of tree martingales as that in Example 2.1.

Example 2.3. Let $\mathbf{T} = (V(\mathbf{T}), E(\mathbf{T}))$ be a tree. If $e_t \in E(\mathbf{T})$ is an arc of $\mathbf{T}$, $e_t = (t^+, t)$, t^+ is called the trail of e_t and t is called the head of e_t . Now, we put i.i.d. random variables with mean zero on the edges, that is,

$$\xi(e_t) = \xi_t, \quad t \in \mathbf{T}$$

is an independent random variable with mean zero for every arc $e_t \in E(\mathbf{T})$. Therefore, a directed network $N = (\xi(e_t), e_t \in E(\mathbf{T}))$ is obtained. Furthermore, note that

$$\mathbf{T}^{t^+} = (V(\mathbf{T}^{t^+}), E(\mathbf{T}^{t^+}))$$

is a directed locally finite subtree of the tree $\mathbf{T}$; the weight of each vertex in the tree $\mathbf{T}$ is defined by

$$(2.6) \quad f_{t^+} \stackrel{\text{def}}{=} \sum_{e_t \in E(\mathbf{T}^{t^+})} \xi(e_t) = \sum_{\substack{t \in \mathbf{T}^{t^+} \\ t \neq t^+}} \xi_t$$

for every element $t \in \mathbf{T}$. We define

$$\mathcal{F}_{t+} \stackrel{\text{def}}{=} \sigma(f_{t+}), \quad t \in \mathbf{T},$$

which is a family of σ -algebras generated by the weighted function f_{t+} of every vertex in the tree $\mathbf{T}$; then a family $(f_{t+}, t \in \mathbf{T})$ of tree martingales is obtained. Obviously,

$$f_t = (f_{t+} | \mathcal{F}_t) \quad \text{and} \quad f_{t+} = (f_{t+} | \mathcal{F}_{t+}),$$

namely, $(f_t, t \in \mathbf{T})$ is a family of tree martingales with respect to the family $(\mathcal{F}_t, t \in \mathbf{T})$ of σ -algebras.

3. The main results. For any X -valued tree martingales $f = (f_t, t \in \mathbf{T})$, $f = (f_s, s \in \mathbf{T}_t)$ is a sequence of X valued one-parameter martingales. Their maximal functions are defined, respectively, by

$$f_t^* \stackrel{\text{def}}{=} \sup_{s \in \mathbf{T}_t} \|f_s\|, \quad f^* \stackrel{\text{def}}{=} \sup_{t \in \mathbf{T}} \|f_t\|.$$

X -valued tree martingale differences are defined by

$$(3.1) \quad d_t f \stackrel{\text{def}}{=} P_{t+} f - P_t f, \quad t \in \mathbf{T};$$

furthermore, for $1 < a < \infty$, we define

$$(3.2) \quad S_t^{(a)}(f) = \left(\sum_{s \in \mathbf{T}_t} \|d_s f\|^a \right)^{1/a}, \quad s_t^{(a)}(f) = \left(\sum_{s \in \mathbf{T}_t} E_s \|d_s f\|^a \right)^{1/a},$$

$$S^{(a)}(f) \stackrel{\text{def}}{=} \sup_{t \in \mathbf{T}} S_t^{(a)}(f), \quad s^{(a)}(f) \stackrel{\text{def}}{=} \sup_{t \in \mathbf{T}} s_t^{(a)}(f).$$

When $a = 2$, the two operators $S^{(a)}(f)$ and $s^{(a)}(f)$ are, respectively, extensions of quadratic and conditional quadratic variations of one-parameter martingales in the tree martingale case. For the two operators, Weisz [19] established an analogue of Burkholder-Davis-Gundy's inequalities for tree martingales when X is a scalar space. With the help of a family of positive contractive projections in L^1 spaces, He [7] proved the tree martingale Doob's inequality provided X is a scalar space.

Bourgain [1] and Burkholder [2, 3] have shown that there are some relations between X -valued martingale inequalities and the convexity of martingale's valued space X if X is a Banach space. Next, Piser [15] and Kwapien [14] further point out that, if Burkholder-Gundy's inequality for X valued one-parameter martingales holds, provided X is a Banach space, then the Banach space X is a Hilbert space. So the research of quadratic variation and conditional quadratic variation for general Banach space X -valued martingales is trivial. Piser [15] proved the following two theorems for X -valued one-parameter martingales $f = (f_s, s \in \mathbf{T}_t)$.

Lemma 3.1. *Assume that X is a Banach space, $2 \leq a < \infty$, $1 < p < \infty$. Then the following conditions are equivalent:*

- (1) *X is isomorphic to a a -uniformly convex space;*
- (2) *there exists a constant $C > 0$ depending only on a and p such that*

$$(3.3) \quad \|S_t^{(a)}(f)\|_p \leq C\|f_t^*\|_p,$$

$$(3.4) \quad \|s_t^{(a)}(f)\|_p \leq C\|f_t^*\|_p,$$

for any X -valued one-parameter martingales $f = (f_s, s \in \mathbf{T}_t)$.

Lemma 3.2. *Assume that X is a Banach space, $1 \leq a < 2$, $1 < p < \infty$. Then the following conditions are equivalent:*

- (1) *X is isomorphic to a a -uniformly convex space;*
- (2) *there exists a constant $C > 0$ depending only on a and p such that*

$$(3.5) \quad C\|f_t^*\|_p \leq \|S_t^{(a)}(f)\|_p,$$

$$(3.6) \quad C\|f_t^*\|_p \leq \|s_t^{(a)}(f)\|_p,$$

for any X -valued one-parameter martingales $f = (f_s, s \in \mathbf{T}_t)$.

He and Shen [9] tried to investigate the tree martingale transform operator and have shown that the maximal operators of X -valued tree martingale transforms are norm-bounded in $L^p(X)$, provided X is a UMD space. In [8], He and Hou have shown that, assuming that X is isomorphic to an a -uniformly convex space, ($2 \leq a < \infty$). Then,

for an X -valued predictable tree martingale $f = (f_t, t \in \mathbf{T})$ and $p \geq 1, \max(a, p) \leq q < \infty$, we have

$$(3.7) \quad \|(S_t^{(a)}(f), t \in \mathbf{T})\|_{\mathbf{M}^{p\infty}} \leq C_{pq} \|f\|_{\mathbf{P}^{pq}},$$

$$(3.8) \quad \|(s_t^{(a)}(f), t \in \mathbf{T})\|_{\mathbf{M}^{p\infty}} \leq C_{pq} \|f\|_{\mathbf{P}^{pq}},$$

where C_{pq} depends only p and q . It is well known that if the measure space is *granular* in the sense that one has a lower bound $\mu(E) \geq C > 0$ for all sets E of positive measure, then functions are prohibited from being arbitrarily narrow, and lower norms control higher norms:

$$\|f\|_q \leq \|f\|_p c^{(1/q)-(1/p)} \text{ whenever } 0 < p \leq q \leq \infty.$$

Now, suppose a probability space $(\Omega, \mathcal{F}_\infty, P)$ is *granular* and $f \in L^1(X)$ with $f > c > 0$. Then

$$\inf_{t \in \mathbf{T}} E(f | \mathcal{F}_t) \geq c.$$

Since Lemma 3.2 shows that if $1 < a < 2$ then $L^p(X)$ -norms of the operators $S_t^{(a)}(f), s_t^{(a)}(f)$ control $L^p(X)$ -norms of maximal function, but $L^p(X)$ -norms of maximal function do not control the $L^p(X)$ -norms of the operators $S_t^{(a)}(f), s_t^{(a)}(f)$ in the one-parameter martingale case, as some functions in a *granular* probability space, for some X -valued tree martingales $f = (f_t, t \in \mathbf{T})$ with

$$\inf_{t \in \mathbf{T}} \|f_t\| \geq c > 0,$$

do (3.7) and (3.8) still hold when $1 < a \leq 2$? In this paper, we shall investigate this problem.

Theorem 3.3. *Let $f = (f_t, t \in \mathbf{T})$ be a family of X -valued predictable tree martingales for which $E[S^{(a)}(f)] < \infty$, $E[s^{(a)}(f)] < \infty$ and $\inf_{t \in \mathbf{T}} \|f_t\| \geq c > 0$. $\lambda = (\lambda_t, t \in \mathbf{T})$ is a prediction belonging to f , and*

$$\alpha_t \stackrel{\text{def}}{=} \chi_{\{\lambda_t > y, \lambda_s \leq y, s < t\}} \quad (s, t \in \mathbf{T}).$$

Suppose X is isomorphic to a $2a$ -uniformly convex space ($1 < a \leq 2$). If there exists a constant $\alpha_0 \in (1, \infty)$ such that $\sum_{t \in \mathbf{T}} (E\alpha_t)^{1/\alpha_0} < \infty$,

then for any $p \geq 1$ and $\max(a, p) \leq q < \infty$,

(3.9)

$$\|(S_t^{(a)}(f), t \in \mathbf{T})\|_{\mathbf{M}^{p\infty}} \leq C_{pq}(\|f\|_{\mathbf{P}^{qp}} + \|f\|_{\mathbf{P}^{qp}}^{(p/q)^2} + \|f\|_{\mathbf{P}^{(q+p)/p}}^{(p+q)/p}),$$

(3.10)

$$\|(s_t^{(a)}(f), t \in \mathbf{T})\|_{\mathbf{M}^{p\infty}} \leq C_{pq}(\|f\|_{\mathbf{P}^{qp}} + \|f\|_{\mathbf{P}^{qp}}^{(p/q)^2} + \|f\|_{\mathbf{P}^{(q+p)/p}}^{(p+q)/p}),$$

where C_{pq} depends only on p and q .

4. Proof of the main results. Our proof of Theorem 3.3 requires a series of preliminary lemmas. Let $f = (f_t, t \in \mathbf{T})$ be an X -valued tree martingale and $\varepsilon = (\varepsilon_t, t \in \mathbf{T})$ a sequence of functions for which each ε_t is $\mathcal{F}_t$ -measurable. Define

$$(4.1) \quad \begin{aligned} S_{\varepsilon, t, s}^{(a)}(f) &= \left(\sum_{t \leq r < s} \|\varepsilon_r d_r f\|^a \right)^{1/a}, \\ s_{\varepsilon, t, s}^{(a)}(f) &= \left(\sum_{t \leq r < s} E_r \|\varepsilon_r d_r f\|^a \right)^{1/a}, \end{aligned}$$

and

$$(4.2) \quad S_{\varepsilon, t}^{(a)}(f) = \sup_{s \geq t} S_{\varepsilon, t, s}^{(a)}(f), \quad S_{\varepsilon}^{(a)} = \sup_{t \in \mathbf{T}} S_{\varepsilon, t}^{(a)}(f),$$

$$(4.3) \quad s_{\varepsilon, t}^{(a)}(f) = \sup_{s \geq t} s_{\varepsilon, t, s}^{(a)}(f), \quad s_{\varepsilon}^{(a)} = \sup_{t \in \mathbf{T}} s_{\varepsilon, t}^{(a)}(f).$$

He and Hou in [11] extended Weisz's result [19] to the following case.

Lemma 4.1. *Suppose that $f = (f_t, t \in \mathbf{T})$ is an X -valued predictable tree martingale and that $\lambda = (\lambda_t, t \in \mathbf{T})$ is a prediction belonging to f . Let $\varepsilon_t = \chi_{\{x < \lambda_{t+} \leq 2x\}}$ ($t \in \mathbf{T}$) for any real number $x > 0$ and $\varepsilon = (\varepsilon_t, t \in \mathbf{T})$. If $1 \leq a < \infty$, then*

$$(4.4) \quad S_{\varepsilon}^{(a)}(f) \leq \sup_{t \in \mathbf{T}} \alpha_t S_{\varepsilon, t}^{(a)}(f) + 4x \chi_{\{\lambda^* > x\}},$$

$$(4.5) \quad s_{\varepsilon}^{(a)}(f) \leq \sup_{t \in \mathbf{T}} \alpha_t s_{\varepsilon, t}^{(a)}(f) + 4x \chi_{\{\lambda^* > x\}},$$

where $(\alpha_t, t \in \mathbf{T})$ is defined as in Theorem 3.3 and $\lambda^* = \sup_{t \in \mathbf{T}} \lambda_t$ is the maximal function of λ .

Since $\sum_{t \in \mathbf{T}} E\alpha_t$ can be viewed as a nonnegative function on $\mathbf{R}^+$, i.e., $\sum_{t \in \mathbf{T}} E\alpha_t$ is a nonnegative function depending on y , some subsets on $\mathbf{R}^+$ are defined by

$$\begin{aligned}\mathbf{A}_1 &= \left\{ y \in \mathbf{R}^+ : 1 < \sum_{t \in \mathbf{T}} E\alpha_t \leq \infty \right\}, \\ \mathbf{A}_2 &= \left\{ y \in \mathbf{R}^+ : \sum_{t \in \mathbf{T}} E\alpha_t < 1 \right\}, \\ \mathbf{A}_3 &= \left\{ y \in \mathbf{R}^+ : \sum_{t \in \mathbf{T}} E\alpha_t = 1 \right\}.\end{aligned}$$

It is clear that

$$(4.6) \quad \mathbf{A}_i \cap \mathbf{A}_j = \emptyset (i \neq j; i, j = 1, 2, 3) \quad \text{and} \quad \cup_{i=1}^3 \mathbf{A}_i = \mathbf{R}^+.$$

We can also define their indicator functions in the form

$$\chi_{\mathbf{A}_i}(y) = 1, y \in \mathbf{A}_i \quad \text{and} \quad \chi_{\mathbf{A}_i}(y) = 0, y \notin \mathbf{A}_i, \quad i = 1, 2, 3.$$

Lemma 4.2. *Let $f = (f_t, t \in \mathbf{T})$, $\lambda = (\lambda_t, t \in \mathbf{T})$ and $(\alpha_t, t \in \mathbf{T})$ be defined as in Lemma 4.1 and $1 < p \leq q < \infty$. If there exists a constant $\alpha_0 \in (1, \infty)$ for which $\sum_{t \in \mathbf{T}} (E\alpha_t)^{1/\alpha_0} < \infty$, then there also exists a constant $\beta > 1$ such that*

$$(4.7) \quad \sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_1} E\alpha_t)^{1/\beta} \leq \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_1} E\alpha_t \right)^{1+(q/p)},$$

$$(4.8) \quad \sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_2} E\alpha_t)^{1/\beta} \leq \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_2} E\alpha_t \right)^{p/q}.$$

Proof. Using Hölder's inequality, we can derive that, for any $a, b > 1$ with $(1/a) + (1/b) = 1$,

$$(4.9) \quad E\alpha_t \leq (E\alpha_t)^{1/a} \quad (t \in \mathbf{T}).$$

Let $Q(x) = \sum_{t \in \mathbf{T}} (E\alpha_t)^{1/x}$ ($x > 1$); then, by (4.9), we see that

$$(4.10) \quad \sum_{t \in \mathbf{T}} E\alpha_t \leq Q(x) \quad (x > 1).$$

On the other hand, since $0 \leq E\alpha_t \leq 1$ ($t \in \mathbf{T}$), if $1 < x_1 < x_2$ then $(E\alpha_t)^{1/x_1} < (E\alpha_t)^{1/x_2}$ for any $t \in \mathbf{T}$. Thus,

$$(4.11) \quad \sum_{t \in \mathbf{T}} (E\alpha_t)^{1/x_1} \leq \sum_{t \in \mathbf{T}} (E\alpha_t)^{1/x_2}, \text{ i.e., } Q(x_1) \leq Q(x_2).$$

Moreover, it follows from the assumption (there exists a constant $\alpha_0 \in (1, \infty)$ such that $\sum_{t \in \mathbf{T}} (E\alpha_t)^{1/\alpha_0} < \infty$) and (4.10), (4.11) that $Q(x)$ satisfies on interval $[1, \alpha_0]$

$$(4.12) \quad \sum_{t \in \mathbf{T}} E\alpha_t \leq Q(x) \leq Q(\alpha_0) = \sum_{t \in \mathbf{T}} (E\alpha_t)^{1/\alpha_0} < \infty.$$

Therefore, by (4.11) and (4.12), we see that $Q(x)$ is a finite increasing function with respect to x on interval $[1, \alpha_0]$. And, by (4.12), we further see that

$$(4.13) \quad \lim_{x \rightarrow 1^+} Q(x) = \lim_{x \rightarrow 1^+} \sum_{t \in \mathbf{T}} (E\alpha_t)^{1/x} = \sum_{t \in \mathbf{T}} E\alpha_t,$$

that is, the series $Q(x) = \sum_{t \in \mathbf{T}} (E\alpha_t)^{1/x}$ converges to $\sum_{t \in \mathbf{T}} E\alpha_t$ as $x \rightarrow 1^+$ on interval $[1, \alpha_0]$. If $y \in \mathbf{A}_1$, then $1 < \sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_1} E\alpha_t) < \infty$. It is easy to see that

$$(4.14) \quad \sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_1} E\alpha_t) < \left[\sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_1} E\alpha_t) \right]^{1+(q/p)} < \infty.$$

Also, it follows from (4.13) that

$$(4.15) \quad \lim_{x \rightarrow 1^+} \sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_1} E\alpha_t)^{1/x} = \sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_1} E\alpha_t < \infty.$$

Therefore, from (4.14) and (4.15), one can see that there exists a constant $\beta' \in (1, \alpha_0]$ such that

$$\sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_1} E\alpha_t)^{1/\beta'} \leq \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_1} E\alpha_t \right)^{1+(q/p)}.$$

If $y \in \mathbf{A}_2$, then $0 \leq \sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_2} E \alpha_t) < 1$. Furthermore, since $0 < (p/q) < 1$, we have

$$\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_2} E \alpha_t < \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_2} E \alpha_t \right)^{p/q} < 1.$$

In the same way, we can show that there is a constant $\beta'' \in (1, \alpha_0]$ such that

$$\sum_{t \in \mathbf{T}} (\chi_{\mathbf{A}_2} E \alpha_t)^{1/\beta''} \leq \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_2} E \alpha_t \right)^{p/q}.$$

Set $\beta = \min\{\beta', \beta''\}$. Then β is a constant satisfying (4.7) and (4.8). $\square$

A decomposition of martingale differences of an X -valued tree martingale $f = (f_t, t \in \mathbf{T})$ is expressed in terms of

$$(4.16) \quad u^{(k)} = (u_t^{(k)}, t \in \mathbf{T}) = (\chi_{\{2^k < \|d_t f\| \leq 2^{k+1}\}}, t \in \mathbf{T}), \quad k \in \mathbf{Z}.$$

It is clear that, for an X -valued tree martingale $f = (f_t, t \in \mathbf{T})$, we have

$$(4.17) \quad |d_t f|^{2a} = \sum_{k \in \mathbf{Z}} u_t^{(k)} |d_t f|^{2a}, \quad |d_t f|^a = \sum_{k \in \mathbf{Z}} u_t^{(k)} |d_t f|^a.$$

Lemma 4.3. *Let $f = (f_t, t \in \mathbf{T})$, $\lambda = (\lambda_t, t \in \mathbf{T})$, and $\varepsilon = (\varepsilon_t, t \in \mathbf{T})$ be defined as in Lemma 4.1. If $\|S^{(a)}(f)\|_\alpha < \infty$, $\|s^{(a)}(f)\|_\alpha < \infty$, $\inf_{t \in \mathbf{T}} \|f_t\| \geq c > 0$ and $1 < a < p$, then*

$$(4.18) \quad E \left[E_t \left(\sum_{r \geq t} \sum_{k \leq -1} \varepsilon_r u_r^{(k)} \|d_r f\|^a \right)^{p/a} \right] \leq C x^{2a},$$

$$(4.19) \quad E \left[E_t \left(\sum_{r \geq t} \sum_{k \leq -1} u_r^{(k)} E_r \|\varepsilon_r d_r f\|^a \right)^{p/a} \right] \leq C x^{2a}.$$

Proof. Case 1: $x < c/2$. From the definition of a predictable tree martingale $f = (f_t, t \in \mathbf{T})$, we easily see that

$$(4.20) \quad \|f_{r+}\| \leq \lambda_{r+}, \quad \|f_r\| \leq \lambda_r \leq \lambda_{r+} \quad (r \geq t; r, t \in \mathbf{T}),$$

and by $\varepsilon_r = \chi_{\{x < \lambda_{r+} \leq 2x\}}$ and (4.20), one can further derive that

$$(4.21) \quad \|f_{r+}\| + \|f_r\| \leq 4x.$$

Using (4.21) and $\inf_{t \in \mathbf{T}} \|f_t\| \geq c > 0$, we can derive that $x \geq c/2$; this is an antinomy with the assumption $x < c/2$. Therefore,

$$(4.22) \quad \varepsilon_r = \chi_{\{x < \lambda_{r+} \leq 2x\}} \leq \chi_{\{\|f_{r+}\| + \|f_r\| \leq 4x\}} = 0.$$

We easily get the following equality from (4.22)

$$E \left[E_t \left(\sum_{r \geq t} \sum_{k \leq -1} u_r^{(k)} \|\varepsilon_r d_r f\|^a \right)^{p/a} \right] = 0.$$

This implies (4.18) as desired.

Case 2: $x \geq c/2$. By [19, Theorem 2.10, page 22] and $\|s^{(a)}(f)\|_p < \infty$, we can also derive that

$$\begin{aligned} E \left[E_t \left(\sum_{r \geq t} \sum_{k \leq -1} u_r^{(k)} E_r \|\varepsilon_r d_r f\|^a \right)^{p/a} \right] \\ \leq C_q^p E \left[\left(\sum_{r \geq t} \sum_{k \leq -1} \varepsilon_r u_r^{(k)} \|d_r f\|^a \right)^{p/a} \right] \\ \leq C_q^p \|s^{(a)}(f)\|_p^p \\ \leq Cx^{2p}. \end{aligned}$$

In the same way, we can show that (4.19) holds. The proof is now complete. $\square$

Lemma 4.4. *Suppose that the conditions of Theorem 3.3 are met, and let $\varepsilon = (\varepsilon_t, t \in \mathbf{T})$ be defined as in Lemma 4.1. If $p \geq 1$, $\max(a, p) \leq q < \infty$, then for all real numbers $z, x \in \mathbf{R}^+$,*

$$(4.23) \quad P(Y > zx, \mathbf{A}_1) \leq C_{pq} z^{-q} x^{-p} \|\lambda\|_{\mathbf{M}^{(p+q)q}}^{(p+q)},$$

$$(4.24) \quad P(Y > zx, \mathbf{A}_2) \leq C_{pq} z^{-q} x^{q-(p^3/q^2)} \|\lambda\|_{\mathbf{M}^{pq}}^{p^3/q^2},$$

where Y denotes one of the functions $\sup_{t \in \mathbf{T}} \alpha_t S_{\varepsilon, t}^{(a)}(f)$, $\sup_{t \in \mathbf{T}} \alpha_t s_{\varepsilon, t}^{(a)}(f)$.

Proof. Let $Y = \sup_{t \in \mathbf{T}} \alpha_t(y) s_{\varepsilon; t}(f)$. It follows from $|\bar{\phi}| = 1$ that

$$(4.25) \quad \begin{aligned} s_{\varepsilon; t}^{(a)}(f) &= \sup_{s \geq t} \left(\sum_{t \leq r < s} E_r \|\varepsilon_r(d_r f)\|^a \right)^{1/a} \\ &\leq \left(\sum_{r \geq t} \varepsilon_r E_r \|d_r f \bar{\phi}_t\|^a \right)^{1/a}. \end{aligned}$$

Substituting (4.25) into this and by [19, Theorem 2.10, page 22], we have

$$(4.26) \quad \begin{aligned} E_t[(s_{\varepsilon; t}^{(a)}(f))^\theta] &\leq E_t \left[\left(\sum_{r \geq t} \varepsilon_r E_r \|d_r f \bar{\phi}_t\|^a \right)^{\theta/a} \right] \\ &\leq C_\theta E_t \left[\left(\sum_{r \geq t} \varepsilon_r \|d_r f\|^a \right)^{\theta/a} \right] \end{aligned}$$

for any $\theta \geq a$. By the definition of $u_r^{(k)}$, we can derive that

$$(4.27) \quad \begin{aligned} \|d_r f\|^a &= \sum_{k \leq -1} u_r^{(k)} \|d_r f\|^a + \sum_{k \geq 0} u_r^{(k)} \|d_r f\|^a \\ &\leq \sum_{k \leq -1} u_r^{(k)} \|d_r f\|^a + \sum_{k \geq 0} u_r^{(k)} \|d_r f\|^{2a}. \end{aligned}$$

Choose a constant $\alpha > 1$ such that $1/\alpha + 1/\beta = 1$ (where β is required in Lemma 4.2). Since $2 \geq a > 1$, we must have $2a > 2$. Substituting (4.26) and (4.27) into this, and applying Lemma 3.1, [19, Theorem 2.10, page 22] and Jensen's inequality to the one parameter X -valued martingale $(f_r \bar{\phi}_t, r \geq t)$ ($r, t \in \mathbf{T}$), one can further derive that, for any $\theta \geq a$,

$$(4.28) \quad \begin{aligned} (E_t[(s_{\varepsilon; r}^{(a)}(f))^\theta])^\alpha &\leq C_\theta (E_t \left[\left(\sum_{r \geq t} \varepsilon_r \|d_r f\|^a \right)^{\theta/a} \right])^\alpha \\ &\leq C_\theta^\alpha E_t \left[\left(\sum_{r \geq t} \varepsilon_r \|d_r f\|^a \right)^{(\theta\alpha)/a} \right] \end{aligned}$$

$$\begin{aligned}
&\leq C_\theta^\alpha E_t \left[\left(\sum_{r \geq t} \varepsilon_r \left(\sum_{k \leq -1} u_r^{(k)} \|d_r f\|^a \right. \right. \right. \\
&\quad \left. \left. \left. + \sum_{k \geq 0} u_r^{(k)} \|d_r f\|^{2a} \right) \right)^{(2\theta\alpha)/(2a)} \right] \\
&\leq C_\theta^\alpha E_t \left[\left(\sum_{r \geq t} \sum_{k \leq -1} \varepsilon_r u_r^{(k)} \|d_r f\|^a \right)^{(\theta\alpha)/a} \right] \\
&\quad + E_t \left[\left(\sum_{r \geq t} \varepsilon_r \|d_r f\|^{2a} \right)^{(2\theta\alpha)/2a} \right] \\
&\leq C_\theta^\alpha E_t \left[\left(\sum_{r \geq t} \sum_{k \leq -1} \varepsilon_r u_r^{(k)} \|d_r f\|^a \right)^{(\theta\alpha)/a} \right] \\
&\quad + C_\theta^\alpha E_t \left[\left\| \sum_{r \geq t} \varepsilon_r (E_{r+} f - E_r f) \right\|^{2\theta\alpha} \right] \\
&= C_\theta^\alpha E_t \left[\left(\sum_{r \geq t} \sum_{k \leq -1} \varepsilon_r u_r^{(k)} \|d_r f\|^a \right)^{(\theta\alpha)/a} \right] \\
&\quad + C_\theta^\alpha E_t \left[\left\| \sum_{r \geq t} \varepsilon_r (f_{r+} - f_r) \right\|^{2\theta\alpha} \right].
\end{aligned}$$

Set

$$r_0 = \min(r \geq t : \lambda_{r+} > x), \quad \eta = \min(r \geq t : \lambda_{r+} > 2x).$$

By the definition of ε_r and (4.28), we have

$$\begin{aligned}
(4.29) \quad &(E_t[(s_{\varepsilon;r}^{(a)}(f))^\theta])^\alpha \leq C_\theta^\alpha E_t \left[\left(\sum_{r \geq t} \sum_{k \leq -1} \varepsilon_r u_r^{(k)} \|d_r f\|^a \right)^{(\theta\alpha)/a} \right] \\
&\quad + C_\theta^\alpha x^{2\theta\alpha}.
\end{aligned}$$

Taking the expectation and substituting (4.18) of Lemma 4.3 into this yields that

$$\begin{aligned}
(4.30) \quad &E[(E_t(s_{\varepsilon;r}^{(a)}(f))^\theta)^\alpha] \leq C_\theta^\alpha E \left[E_t \left(\sum_{r \geq t} \sum_{k \leq -1} \varepsilon_r u_r^{(k)} \|d_r f\|^a \right)^{(\theta\alpha)/a} \right] \\
&\quad + C_\theta^\alpha x^{2\theta\alpha} \\
&\leq C_\theta x^{2\theta\alpha} + C_\theta^\alpha x^{2\theta\alpha} \leq C_\theta x^{2\theta\alpha}.
\end{aligned}$$

On the other hand, by Chebyshev's inequality and [19, Theorem 2.10, page 22], for any $\theta \geq q$, we have

$$\begin{aligned}
 P(Y > zx) &\leq (zx)^{-q} E \sup_{t \in \mathbf{T}} \alpha_t(s_{\varepsilon;t}^{(a)}(f))^q \\
 (4.31) \qquad &\leq C_q (zx)^{-q} E \left(\sum_{t \in \mathbf{T}} \alpha_t(s_{\varepsilon;t}^{(a)}(f))^\theta \right)^{q/\theta} \\
 &\leq C_q (zx)^{-q} E \left(\sum_{t \in \mathbf{T}} \alpha_t E_t(s_{\varepsilon;t}^{(a)}(f))^\theta \right)^{q/\theta}.
 \end{aligned}$$

Since $0 < q/\theta < 1$, using Jensen's inequality we get

$$\begin{aligned}
 (4.32) \qquad E \left(\sum_{t \in \mathbf{T}} \alpha_t E_t(s_{\varepsilon;t}^{(a)}(f))^\theta \right)^{q/\theta} &\leq \left(E \sum_{t \in \mathbf{T}} \alpha_t E_t(s_{\varepsilon;t}^{(a)}(f))^\theta \right)^{q/\theta} \\
 &\leq \left(\sum_{t \in \mathbf{T}} E[\alpha_t E_t(s_{\varepsilon;t}^{(a)}(f))^\theta] \right)^{q/\theta},
 \end{aligned}$$

and since $1/\alpha + 1/\beta = 1$, using Hölder's inequality and (4.30), we have

$$\begin{aligned}
 (4.33) \qquad E[\alpha_t E_t(s_{\varepsilon;t}^{(a)}(f))^\theta] &\leq (E\alpha_t^\beta)^{1/\beta} (E[(E_t(s_{\varepsilon;t}^{(a)}(f))^\theta)^\alpha])^{1/\alpha} \\
 &\leq C_\theta x^{2\theta} (E\alpha_t)^{1/\beta}.
 \end{aligned}$$

Set $\theta = q^2/p$; substituting (4.32) and (4.33) into this yields that

$$\begin{aligned}
 (4.34) \qquad E \left(\sum_{t \in \mathbf{T}} \alpha_t E_t(s_{\varepsilon;t}^{(a)}(f))^\theta \right)^{q/\theta} &\leq \left(\sum_{t \in \mathbf{T}} C_\theta x^{2\theta} (E\alpha_t)^{1/\beta} \right)^{q/\theta} \\
 &\leq \left(C_\theta x^{2\theta} \sum_{t \in \mathbf{T}} (E\alpha_t)^{1/\beta} \right)^{q/\theta}.
 \end{aligned}$$

Next, since $(p+q)/p > 1$, by Jensen's inequality, (4.7) of Lemma 4.2

and (4.34), one can show that

$$\begin{aligned}
 (4.35) \quad E \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_1} \alpha_t E_t(s_{\varepsilon; t}^{(a)}(f))^\theta \right)^{q/\theta} \\
 \leq C_\theta x^{2q} \left(\left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_1} E \alpha_t \right)^{(p+q)/p} \right)^{q/\theta} \\
 \leq C_\theta x^{2q} \chi_{\mathbf{A}_1} \left[E \left(\sum_{t \in \mathbf{T}} \alpha_t \right) \right]^{(p+q)/p} \\
 \leq C_\theta x^{2q} \chi_{\mathbf{A}_1} E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{(p+q)/q}.
 \end{aligned}$$

By the definition of the quasi-norm $\|\cdot\|_{\mathbf{M}^{(p+q)q}}$, one can derive that

$$\begin{aligned}
 (4.36) \quad E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{(p+q)/q} &= E \left(\sum_{t \in \mathbf{T}} \chi_{\{\lambda_t > x, \lambda_s \leq x, s \leq t\}} \right)^{(p+q)/q} \\
 &\leq x^{-(p+q)} \|\lambda\|_{\mathbf{M}^{(p+q)q}}^{(p+q)}.
 \end{aligned}$$

Using (4.31) and substituting (4.35) and (4.36) into this, one can further show that

$$\begin{aligned}
 (4.37) \quad P(Y > zx, \mathbf{A}_1) &\leq C_\beta (zx)^{-q} E \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_1} \alpha_t E_t(s_{\varepsilon; t}^{(a)}(f))^\theta \right)^{q/\theta} \\
 &\leq C_{pq} z^{-q} x^q \chi_{\mathbf{A}_1} E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{(p+q)/q} \\
 &\leq C_{pq} z^{-q} x^{-p} \|\lambda\|_{\mathbf{M}^{(p+q)q}}^{p+q}.
 \end{aligned}$$

On the other hand, since $q/p > 1$, by using Jensen's inequality, (4.8)

of Lemma 4.2 and (4.34), one can show that

$$\begin{aligned}
 (4.38) \quad E \left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_2} \alpha_t E_t(s_{\varepsilon; t}^{(a)}(f))^\theta \right)^{q/\theta} \\
 \leq C_\theta x^{2q} \left(\left(\sum_{t \in \mathbf{T}} \chi_{\mathbf{A}_2} E \alpha_t \right)^{p/q} \right)^{q/\theta} \\
 \leq C_\theta x^{2q} \left[\chi_{\mathbf{A}_2} E \left(\sum_{t \in \mathbf{T}} \alpha_t \right) \right]^{(p/q)^2} \\
 \leq C_\theta x^{2q} \chi_{\mathbf{A}_2} \left[\left(E \left(\sum_{t \in \mathbf{T}} \alpha_t \right) \right)^{q/p} \right]^{(p/q)^3} \\
 \leq C_\theta x^{2q} \chi_{\mathbf{A}_2} \left[E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{q/p} \right]^{(p/q)^3}.
 \end{aligned}$$

By the definition of the quasi-norm $\|\cdot\|_{\mathbf{M}^{pq}}$ we can also derive that

$$\begin{aligned}
 (4.39) \quad \left[E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{q/p} \right]^{(p/q)^3} &= \left[E \left(\sum_{t \in \mathbf{T}} \chi_{\{\lambda_t > x, \lambda_s \leq x, s \leq t\}} \right)^{q/p} \right]^{(p/q)^3} \\
 &\leq x^{-p^3/q^2} \|\lambda\|_{\mathbf{M}^{pq}}^{p^3/q^2}.
 \end{aligned}$$

Using (4.31) and substituting (4.38) and (4.39) into this, one can further show that

$$\begin{aligned}
 (4.40) \quad P(Y > zx, \mathbf{A}_2) &\leq C_q (zx)^{-q} \chi_{\mathbf{A}_2} E \left(\sum_{t \in \mathbf{T}} \alpha_t E_t(\sigma_{\varepsilon; t}^p(f))^\theta \right)^{q/\theta} \\
 &\leq C_{pq} z^{-q} x^q \chi_{\mathbf{A}_2} \left[E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{q/p} \right]^{(p/q)^3} \\
 &\leq C_{pq} z^{-q} x^{q-(p^3/q^2)} \|\lambda\|_{\mathbf{M}^{pq}}^{p^3/q^2}.
 \end{aligned}$$

The proof is complete. $\square$

Proof of Theorem 3.3.

Proof. To prove Theorem 3.3 we need to verify the following inequality

$$(4.41) \quad y^p P(s^{(a)}(f) > 9y) \leq C_{pq} (\|\lambda\|_{\mathbf{M}^{qp}}^p + \|\lambda\|_{\mathbf{M}^{qp}}^{p^3/q^2} + \|\lambda\|_{\mathbf{M}^{(p+q)q}}^{(p+q)}).$$

We divide the proof of (4.41) into several steps.

Step I. We need to verify that, for any $y \in \mathbf{R}^+$, choose $j \in \mathbf{Z}$ such that $2^j < y \leq 2^{j+1}$. Then

$$(4.42) \quad y^p P(s^{(a)}(f) > 9y) \leq \|\lambda\|_{\mathbf{M}^{p\infty}}^p + y^p P\left(\sum_{k \leq j} Y_k > 2^j\right).$$

Equation (4.42) has been proved in Theorem 2.2 ([11, page 226] or [19, page 151]).

Step II. We shall discuss this step by dividing into the following three cases.

Case 1: $y \in \mathbf{A}_1$. Applying (4.23) of Lemma 4.4 to

$$(4.43) \quad Y_k = \sup_{t \in \mathbf{T}} \alpha_t^k s_{\varepsilon^{(k)}, t}^{(a)}(f), \quad k \in \mathbf{Z},$$

we get for any $k \in \mathbf{Z}$ and $z_k > 0$ with $x = 2^k$

$$(4.44) \quad P(Y_k > z_k 2^k, \mathbf{A}_1) \leq C_{pq} z_k^{-q} 2^{-pk} \|\lambda\|_{\mathbf{M}^{(p+q)q}}^{(p+q)}.$$

By (4.42) and (4.44), we can show that the following inequality holds in the same way as in the proof of Theorem 2.2 ([8, pages 225–227]):

$$(4.45) \quad y^p P(Y > 9y, \mathbf{A}_1) \leq C_{pq} \|\lambda\|_{\mathbf{M}^{(p+q)q}}^{(p+q)}.$$

Case 2: $y \in \mathbf{A}_2$. Applying (4.24) of Lemma 4.4 to

$$(4.46) \quad Y_k = \sup_{t \in \mathbf{T}} \alpha_t^k s_{\varepsilon^{(k)}, t}^{(a)}(f), \quad k \in \mathbf{Z},$$

we get, for any $k \in \mathbf{Z}$ and $z_k > 0$ with $x = 2^k$,

$$(4.47) \quad P(Y_k > z_k 2^k, \mathbf{A}_2) \leq C_{pq} z_k^{-q} 2^{(q-(p^3/q^2))k} \|\lambda\|_{M_{qp}}^{p^3/q^2}.$$

And it follows from (4.42) that

$$(4.48) \quad y^p P(Y > 9y, \mathbf{A}_2) \leq \|\lambda\|_{M^{p\infty}}^p + y^p P\left(\sum_{k \leq j} Y_k > 2^j, \mathbf{A}_2\right).$$

(i) $j > 0$. For any $k \in \mathbf{Z}$, when $k < 0$, we shall consider the following equation

$$(4.49) \quad c_{\theta_1} \sum_{k < \min\{j, 0\}} 2^{\theta_1(k-j)} = \frac{1}{2},$$

where $\theta_1 > 0$ and $c_{\theta_1} = (1 - 2^{-\theta_1})/2$. Set

$$(4.50) \quad z_k 2^k := 2^j c_{\theta_1} 2^{\theta_1(k-j)} = c_{\theta_1} 2^{(\theta_1-1)(k-j)} 2^k.$$

Then

$$(4.51) \quad z_k^{-q} 2^{(q-(p^3/q^2))k} \leq C_{\theta_1} y^{-p} 2^{[q-p(p/q)^2-q(\theta_1-1)]k+[p+q(\theta_1-1)]j}.$$

Next, we shall consider the following inequality

$$\begin{cases} q - p(p/q)^2 - q(\theta_1 - 1) > 0 \\ [q - p(p/q)^2 - q(\theta_1 - 1)]/\gamma = -[p + q(\theta_1 - 1)]. \end{cases}$$

We obtain by solving the inequality

$$\begin{cases} \theta_1 < 2 - (p/q)^3 \\ \theta_1 = \frac{\gamma-2}{\gamma-1} + \frac{1}{\gamma-1} \left(\frac{p}{q}\right)^3 - \frac{\gamma}{\gamma-1} \frac{p}{q} \quad (\gamma \in \mathbf{Z}^+); \end{cases}$$

moreover, there exists a θ_1 such that the inequality above holds, because for a fixed pair p, q ($q > p > 1$), there exists a $\gamma_0 \in \mathbf{Z}^+$ such that

$$2 - \left(\frac{p}{q}\right)^3 > \theta_1 = \frac{\gamma_0 - 2}{\gamma_0 - 1} + \frac{1}{\gamma_0 - 1} \left(\frac{p}{q}\right)^3 - \frac{\gamma_0}{\gamma_0 - 1} \frac{p}{q} > 0,$$

and

$$\begin{aligned}
 (4.52) \quad & \left[q - p \left(\frac{p}{q} \right)^2 - q(\theta_1 - 1) \right] k + [p + q(\theta_1 - 1)]j \\
 & < \frac{1}{\gamma_0} \left[q - p \left(\frac{p}{q} \right)^2 - q(\theta_1 - 1) \right] k \\
 & \quad + [p + q(\theta_1 - 1)]j \\
 & = \frac{1}{\gamma_0 - 1} \left[q + p \left(1 - \left(\frac{p}{q} \right)^2 \right) \right] (k - j).
 \end{aligned}$$

Substituting (4.52) into (4.51) yields that

$$(4.53) \quad z_k^{-q} 2^{(q - (p^3/q^2))k} \leq C_{\theta_1} y^{-p} 2^{1/(\gamma_0 - 1)[q + p(1 - (p/q)^2)](k - j)}.$$

By using (4.53), one can derive that

$$\begin{aligned}
 (4.54) \quad & \sum_{k < \min\{j, 0\}} z_k^{-q} 2^{(q - (p^3/q^2))k} \leq C_{pq} y^{-p} \sum_{k < \min\{j, 0\}} 2^{1/(\gamma_0 - 1)[q + p(1 - (p/q)^2)](k - j)} \\
 & \leq C_{pq} y^{-p}.
 \end{aligned}$$

Substituting (4.53) into (4.47) yields that

$$(4.55) \quad P(Y_k > z_k 2^k, \mathbf{A}_2) \leq C_{pq} y^{-p} 2^{1/(\gamma_0 - 1)[q + p(1 - (p/q)^2)](k - j)} \|\lambda\|_{\mathbf{M}^{qp}}^{p^3/q^2},$$

Combining (4.54) and (4.55) yields that

$$(4.56) \quad \sum_{k < \min\{j, 0\}} P(Y_k > z_k 2^k, \mathbf{A}_2) \leq C_{pq} y^{-p} \|\lambda\|_{\mathbf{M}^{qp}}^{p^3/q^2}.$$

For any $k \in Z$, when $k \geq 0$, we shall consider the following equation

$$(4.57) \quad c_{\theta_2} \sum_{0 \leq k \leq j} 2^{\theta_2(k - j)} = \frac{1}{2},$$

where $\theta_2 > 0$ and $c_{\theta_2} = (1/2)(1 - 2^{-\theta_2})/(1 - 2^{-(j+1)\theta_2})$. Set

$$(4.58) \quad z_k 2^k := 2^j c_{\theta_2} 2^{\theta_2(k - j)} = c_{\theta_2} 2^{(\theta_2 - 1)(k - j)} 2^k;$$

then

$$(4.59) \quad z_k^{-q} 2^{(q-(p^3/q^2))k} \leq C_{\theta_2} y^{-p} 2^{[q-p(p/q)^2-q(\theta_2-1)]k+[p+q(\theta_2-1)]j}.$$

And let $\theta_2 = 1 - (p/q)^3$. Then we can derive that

$$(4.60) \quad \begin{aligned} \left[q - p \left(\frac{p}{q} \right)^2 - q(\theta_2 - 1) \right] k &= qk, \\ [p + q(\theta_2 - 1)]j &= p \left[1 - \left(\frac{p}{q} \right)^2 \right] j, \\ z_k^{-q} 2^{(q-(p^3/q^2))k} &\leq C_{\theta_2} y^{-p} 2^{qk+p[1-(p/q)^2]j}, \end{aligned}$$

and by (4.60) we can further derive that

$$(4.61) \quad \sum_{0 \leq k \leq j} z_k^{-q} 2^{(q-(p^3/q^2))k} \leq C_{pq} y^{-p} \frac{2^{p[1-(p/q)^2]j[2^q(j+1) - 1]}}{2^q - 1}.$$

On the other hand, since $E[s^{(a)}(f)] < \infty$, we must have $s^{(a)}(f) < \infty$ almost everywhere. Then $P(s^{(a)}(f) = \infty) = 0$. Therefore, there exist a constant $y_0 \in R$ and $0 < y_0 < \infty$ such that

$$\|s^{(a)}(f)\|_{\mathbf{M}^{p,\infty}} = \sup_{y>0} yP(Y > y)^{1/p} = y_0 P(Y > y_0)^{1/p}.$$

Furthermore, there is a fixed $j_0 \in \mathbf{Z}$ such that $2^{j_0} < y_0 \leq 2^{j_0+1}$. If $j > j_0$ and $j \in \mathbf{Z}$, then for any y with $2^j < y \leq 2^{j+1}$, we have $y > y_0$ and

$$yP(Y > y, \mathbf{A}_2)^{1/p} \leq y_0 P(Y > y_0, \mathbf{A}_2)^{1/p}.$$

In this case, if we can show that the following inequality holds

$$y_0 P(Y > y_0, \mathbf{A}_2)^{1/p} \leq C_{pq} (\|\lambda\|_{\mathbf{M}^{qp}} + \|\lambda\|_{\mathbf{M}^{qp}}^{(p/q)^2}),$$

then the following inequality must also hold:

$$yP(Y > y, \mathbf{A}_2)^{1/p} \leq C_{pq} (\|\lambda\|_{\mathbf{M}^{qp}} + \|\lambda\|_{\mathbf{M}^{qp}}^{(p/q)^2}).$$

Consequently, we only need to verify that if $j \leq j_0$ and $2^j < y \leq 2^{j+1}$, then the following inequality holds:

$$(4.62) \quad yP(Y > y, \mathbf{A}_2)^{1/p} \leq C_{pq} (\|\lambda\|_{\mathbf{M}^{qp}} + \|\lambda\|_{\mathbf{M}^{qp}}^{(p/q)^2}).$$

In fact, it is not difficult to show that (4.62) holds by the results above. For any $j \leq j_0$, by (4.61) we can derive that

$$(4.63) \quad \sum_{0 \leq k \leq j} z_k^{-q} 2^{(q-(p^3/q^2))k} \leq C_{pq} y^{-p} \frac{2^{p[1-(p/q)^2]j_0} [2^{q(j_0+1)} - 1]}{2^q - 1} \leq C_{pq} y^{-p}.$$

Substituting (4.47) and (4.63) into (4.62) yields that

$$(4.64) \quad \begin{aligned} \sum_{0 \leq k \leq j} P(Y_k > z_k 2^k, \mathbf{A}_2) \\ \leq C_{pq} y^{-p} \sum_{0 \leq k \leq j} z_k^{-q} 2^{(q-(p^3/q^2))k} \|\lambda\|_{\mathbf{M}^{qp}}^{p^3/q^2} \\ \leq C_{pq} y^{-p} \|\lambda\|_{\mathbf{M}^{qp}}^{p^3/q^2}. \end{aligned}$$

On the other hand, it follows from (4.49) and (4.57) that

$$(4.65) \quad c_{\theta_1} \sum_{k < 0} 2^{\theta_1(k-j)} + c_{\theta_2} \sum_{0 \leq k \leq j} 2^{\theta_2(k-j)} = 1.$$

Combining (4.65), (4.50) and (4.58) yields that

$$(4.66) \quad 2^j = \sum_{k < 0} c_{\theta_1} 2^{\theta_1(k-j)} 2^j + \sum_{0 \leq k \leq j} c_{\theta_2} 2^{\theta_2(k-j)} 2^j = \sum_{k \leq j} z_k 2^k.$$

That is, if $\sum_{k \leq j} Y_k > 2^j$, then $\sum_{k \leq j} Y_k > \sum_{k \leq j} z_k 2^k$. In addition, $Y_k \geq 0$ and $z_k 2^k \geq 0$; thus, there exist $k_0 \in \mathbf{Z}$ and $k_0 \leq j$ at least such that $Y_{k_0} > z_{k_0} 2^{k_0}$. Therefore,

$$(4.67) \quad \begin{aligned} P\left(\sum_{k \leq j} Y_k > 2^j, \mathbf{A}_2\right) &\leq P(Y_{k_0} > z_{k_0} 2^{k_0}, \mathbf{A}_2) \\ &\leq \sum_{k \leq j} P(Y_k > z_k 2^k, \mathbf{A}_2). \end{aligned}$$

Substituting (4.56) and (4.64) into (4.67) yields that

$$(4.68) \quad \begin{aligned} P\left(\sum_{k \leq j} Y_k > 2^j, \mathbf{A}_2\right) &\leq \sum_{k < \min\{j, 0\}} P(Y_k > z_k 2^k, \mathbf{A}_2) \\ &\quad + \sum_{0 \leq k \leq j} P(Y_k > z_k 2^k, \mathbf{A}_2) \\ &\leq C_{pq} y^{-p} \|\lambda\|_{\mathbf{M}^{qp}}^{p^3/q^2}. \end{aligned}$$

(ii) $j \leq 0$. For any $k \in \mathbf{Z}$, since $k \leq j$, we must have $k < 0$. In this case, we shall consider the following equation:

$$(4.69) \quad c_\theta \sum_{k \leq j} 2^{\theta(k-j)} = 1,$$

where $\theta > 0$, $c_\theta = 1 - 2^{-\theta}$. By (4.69), we can show that the inequality (4.68) holds by using the method in the proof of Theorem 4.7 [19, pages 150–152]).

Finally, we substitute (4.68) into (4.48) and note that $\|\lambda\|_{\mathbf{M}^{p\infty}}^p \leq \|\lambda\|_{\mathbf{M}^{qp}}^p$ (refer to Remark 1). Therefore,

$$(4.70) \quad y^p P(Y > 9y, \mathbf{A}_2) \leq C_{pq} (\|\lambda\|_{\mathbf{M}^{qp}}^p + \|\lambda\|_{\mathbf{M}^{qp}}^{p^3/q^2}).$$

Case 3. $y \in \mathbf{A}_3$. By Jensen's inequality ($q > p > 1$), we can derive that

$$1 = \sum_{t \in \mathbf{T}} E\alpha_t = \left(\sum_{t \in \mathbf{T}} E\alpha_t \right)^{q/p} \leq E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{q/p}.$$

Thus, it follows from $y \in \mathbf{R}^+$ that

$$(4.71) \quad y \leq y \left[E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{q/p} \right]^{1/q}.$$

By the definition of $\|\lambda\|_{\mathbf{M}^{qp}}$ and (4.71), one can further derive that

$$(4.72) \quad \begin{aligned} \sup_{y \in \mathbf{A}_3} y P(Y > 9y)^{1/p} &\leq \sup_{y \in \mathbf{A}_3} y \\ &\leq \sup_{y \in \mathbf{A}_3} y \left[E \left(\sum_{t \in \mathbf{T}} \alpha_t \right)^{q/p} \right]^{1/q} \leq \|\lambda\|_{\mathbf{M}^{qp}}. \end{aligned}$$

Because of (4.72), one sees that

$$(4.73) \quad y^p P(Y > 9y, \mathbf{A}_3) \leq C_{qp} \|\lambda\|_{\mathbf{M}^{qp}}^p.$$

Step III. By (4.45), (4.70) and (4.73), we can show that (4.41) holds.

Finally, taking the minimum over the quasi-norm $\|\lambda\|_{\mathbf{M}^{qp}}$ of all predictions $\lambda = (\lambda_t, t \in \mathbf{T})$ belonging to f by (4.41), we have

$$\|(s_t^{(p)}(f), t \in \mathbf{T})\|_{\mathbf{M}^{p\infty}} \leq C_{pq}(\|f\|_{\mathbf{P}^{qp}} + \|f\|_{\mathbf{P}^{qp}}^{(p/q)^2}) + \|f\|_{\mathbf{P}^{(q+p)/q}}^{(p+q)/p},$$

as required. Now (3.10) of Theorem 3.3 is proved. We can also show that (3.9) of Theorem 3.3 holds in the same way. The whole proof is now complete. $\square$

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SCHOOL OF AUTOMATION, HUAZHONG UNIVERSITY OF SCIENCE AND TECHNOLOGY, WUHAN 430074, CHINA

Email address: hetongjun@163.com

SCHOOL OF AUTOMATION, HUAZHONG UNIVERSITY OF SCIENCE AND TECHNOLOGY, WUHAN 430074, CHINA

Email address: yinquan@mail.hust.edu.cn

SCHOOL OF AUTOMATION, HUAZHONG UNIVERSITY OF SCIENCE AND TECHNOLOGY, WUHAN 430074, CHINA

Email address: shenyi@hust.edu.cn