Operator Fractional Brownian Motion and Martingale Differences

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Abstract

It is well known that martingale difference sequences are very useful in applications and theory. On the other hand, the operator fractional Brownian motion as an extension of the well-known fractional Brownian motion also plays important role in both applications and theory. In this paper, we study the relationship between them. We will construct an approximation sequence of operator fractional Brownian motion based on a martingale difference sequence.

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1. Introduction

Fractional Brownian motion (FBM) is a continuous Gaussian process with stationary increments. It is one of the well-known self-similar processes. Some studies of financial time series and telecommunication networks have shown that this kind of process with long-range dependency-memory might be a better model in some cases than the traditional standard Brownian motion. Due to its applications in the real world and its interesting theoretical properties, fractional Brownian motion has become an object of intense study. One of those studies concerns obtaining its weak limit theorems; see, for example, Enriquez [7], Nieminen [15], Sottinen [17], Li and Dai [12] and the reference therein.

Based on the study of FBM, many authors have proposed a generalization of it, and have obtained many new processes. An extension of FBM is the operator fractional Brownian motion (OFBM). OFBMs are multivariate analogues of one-dimensional FBM. They arise in the context of multivariate time series and long-range dependence (see, for example, Chung [1], Davidson and de Jong [4], Dolado and Marmol [6], Robinson [16], and Marinucci and Robinson [13]). Another context is that of queuing systems, where reflected OFBMs model the size of multiple queues in particular classes of queuing models. They are also studied in problems related to, for example, large deviations (see Delgado [5], and Konstantopoulos and Lin[9]). Similar to those for FBM, weak limit theorems for OFBMs

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have been studied recently. Some new results on approximations of OFBMs have been
obtained. See Dai [2, 3] and the references therein.

It is well known that a martingale difference sequence is extremely useful because it im-
poses much milder restrictions on the memory of the sequence than under independence,
yet most limit theorems that hold for an independent sequence will also hold for a martin-
gale difference sequence. In recent years, some researchers have used this type of sequences
to construct approximation sequences of some known processes. For example, Nieminen
[15] studied the limit theorems for FBMs based on martingale difference sequences. This
is a natural motivation for this paper. The direct motivation is the recent works by Dai
[2, 3], in which, based on a sequence of I.I.D. random variables, the author presented some
weak limit theorems for some special kinds of OFBMs.

In this short paper, we establish a weak limit theorem for a special case of OFBMs,
which comes from Maejima and Mason [14]. The rest of this paper is organized as follows.
In Section 2, we recall OFBMs and martingale-difference sequences, and present the main
result of this paper. Section 3 is devoted to prove the main result of this paper.

2. Operator fractional Brownian motion and Martingale-differences

In this section, we first introduce a special type of OFBMs. Let \( \text{End}(\mathbb{R}^d) \) be the set of
linear operators on \( \mathbb{R}^d \) (endomorphisms) and \( \text{Aut}(\mathbb{R}^d) \) be the set of invertible linear operators
(automorphisms) in \( \text{End}(\mathbb{R}^d) \). For convenience, we will not distinguish an operator
\( D \in \text{End}(\mathbb{R}^d) \) from its associated matrix relative to the standard basis of \( \mathbb{R}^d \). As usual,
for \( c > 0 \),
\[
c^D = \exp \left( (\log c) D \right) = \sum_{k=0}^{\infty} \frac{1}{k!} (\log c)^k D^k.
\]

Throughout this paper, we will use \( \|x\| \) to denote the usual Euclidean norm of \( x \in \mathbb{R}^d \). Without confusion, for \( A \in \text{End}(\mathbb{R}^d) \), we also let \( \|A\| = \max\|x\|=1 \|Ax\| \) denote the
operator norm of \( A \). It is easy to see that for \( A, B \in \text{End}(\mathbb{R}^d) \),
\[
\|AB\| \leq \|A\| \cdot \|B\|, \quad (2.1)
\]
and for every \( A = (A_{ij})_{d \times d} \in \text{End}(\mathbb{R}^d) \),
\[
\max_{1 \leq i,j \leq d} |A_{ij}| \leq \|A\| \leq d^2 \max_{1 \leq i,j \leq d} |A_{ij}|. \quad (2.2)
\]
Let \( \sigma(A) \) be the collection of all eigenvalues of \( A \). We denote
\[
\lambda_A = \min\{Re \lambda : \lambda \in \sigma(A)\} \quad \text{and} \quad \Lambda_A = \max\{Re \lambda : \lambda \in \sigma(A)\}. \quad (2.3)
\]

Let \( x' \) denote the transpose of a vector \( x \in \mathbb{R}^d \). We now extend the fractional Brownian
motion of Riemann-Liouville type studied by Lévy [11, p. 357] to the multivariate case.

**Definition 2.1** Let \( D \) be a linear operator on \( \mathbb{R}^d \) with \( \frac{1}{2} < \lambda_D, \Lambda_D < 1 \). For \( t \in \mathbb{R}_+ \),
define
\[
X(t) = \int_0^t (t - u)^{D-1/2} dW(u), \quad (2.4)
\]
where $W(u) = \{W^1(u), \cdots, W^d(u)\}'$ is a standard $d$-dimensional Brownian motion. We call the process $X = \{X(t)\}$ an operator fractional Brownian motion of Riemann-Liouville (RL-OFBM).

As is standard for the multivariate context, we assume that RL-OFBM is proper. A random variable in $\mathbb{R}^d$ is proper if the support of its distribution is not contained in a proper hyperplane of $\mathbb{R}^d$.

**Remark 2.1** The operator fractional Brownian motion in the current work is a special case of the operator fractional Brownian motions in the work of Maejima and Mason [14, Theorem 3.1].

**Remark 2.2** The RL-OFBM $X$ defined by (2.4) is an operator self-similar Gaussian process.

In this short note, we want to obtain an approximation of RL-OFBM. Inspired by Nieminen [15], we want to construct an approximation sequence of RL-OFBM $X$ by martingale differences.

Let $\{\xi^{(n)}_i\} = (\xi^{(n)}_i, \mathcal{F}^n_i)_{1 \leq i \leq n}$ be a sequence of square integrable martingale differences such that for every sequence $\{i_n\}$ with $\lim_{n \to \infty} i_n = \infty$, where $1 \leq i_n \leq n$,

$$
\lim_{n \to \infty} \frac{\left(\xi^{(n)}_{i_n}\right)^2}{n} = 1, \text{ a.s.}
$$

and

$$
\max_{1 \leq i \leq n} \left|\xi^{(n)}_i\right| \leq \frac{C}{\sqrt{n}}, \text{ a.s.,}
$$

for some $C \geq 1$.

The following lemma follows from Jacod and Shiryaev [10].

**Lemma 2.1** Under the condition (2.6) and the condition

$$
\sum_{i=1}^{[nt]} \left(\xi^{(n)}_i\right)^2 \to t, \text{ a.s.,}
$$

the processes

$$
B^n(t) = \sum_{i=1}^{[nt]} \xi^{(n)}_i
$$

converge in distribution to a Brownian motion $B$, as $n \to \infty$.

**Remark 2.3** Such a type of sequences is very useful, since it is very easy to obtain it in the real world. See, Nieminen [15], for example.

Below, we extend Lemma 2.1 to the $d$-dimensional case. Define

$$
\eta^{(n)}_i = \left(\xi^{(n)}_{i,1}, \cdots, \xi^{(n)}_{i,d}\right)'
$$
where $\xi_{i,k}^{(n)}$, $k = 1, 2, \ldots, d$, are independent copies of $\xi_i^{(n)}$ in Lemma 2.1. Define

$$\eta_n(t) = \sum_{i=1}^{[nt]} \eta_i^{(n)}.$$  \hfill (2.10)

Then, we can get that $\{\eta_i^{(n)}\}_{n \in \mathbb{N}} = \{\eta_i^{(n)}, \mathcal{F}_i^n\}$ is still a sequence of square integrable martingale differences on the probability space $(\Omega, \mathcal{F}, \mathbb{P})$. Inspired by Lemma 2.1, we have the following lemma.

**Lemma 2.2** Under conditions (2.6) and (2.7), the sequence of processes $\eta_n(t)$ converges in law to a $d$-dimensional Brownian motion $W$, as $n \to \infty$.

Noting that $W^i(u)$, $i = 1, \ldots, d$, are mutually independent, and so are $\xi_{i,k}^{(n)}$, we can directly get Lemma 2.2 from Lemma 2.1 and Theorem 11.4.4 in Whitt [19, Chapter 12].

Inspired by Lemma 2.2 and (2.4), we construct the approximation sequence by

$$X_n(t) = \sum_{i=1}^{[nt]} n \int_{\frac{i}{n}}^{\frac{i+1}{n}} \left(\frac{[nt]}{n} - u\right)^{D-\frac{1}{2}} \eta_i^{(n)} du.$$  \hfill (2.11)

Our main objective in this paper is to explain and prove the following theorem.

**Theorem 2.1** The sequence of processes $\{X_n(t), t \in [0, 1]\}$ given by (2.11), as $n \to \infty$, converges weakly to the operator fractional Brownian motion $X$ given by (2.4).

In the rest of this paper, most of the estimates contain unspecified constants. An unspecified positive and finite constant will be denoted by $\tilde{K}$, which may not be the same in each occurrence.

### 3. Proof of Theorem 2.1

In order to prove the main result of this paper, we need a technical lemma. Before we state this technical lemma, we first introduce the following notation

$$K(t, s) = (t - s)^{D-\frac{1}{2}} = (K_{i,j}(t, s))_{d \times d},$$ \hfill (3.1)

and

$$K^n(t, s) = \left(\frac{[nt]}{n} - s\right)^{D-\frac{1}{2}} = (K_{i,j}^n(t, s))_{d \times d}.$$ \hfill (3.2)

The technical lemma follows.

**Lemma 3.1** For any $k, j \in \{1, 2, \ldots, d\}$,

$$\sum_{i=1}^{n} n^2 \int_{\frac{i}{n}}^{\frac{i+1}{n}} K^n_{k,j}(t_i, s)ds \int_{\frac{i}{n}}^{\frac{i+1}{n}} K^n_{k,j}(t_q, s)ds \left(\xi_{i,j}^{(n)}\right)^2 \to \int_{0}^{1} K_{k,j}(t_i, s)K_{k,j}(t_q, s)ds, \text{ a.s.}$$ \hfill (3.3)

for $t_i, t_q \in [0, 1]$, as $n \to \infty$. 

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Before we prove it, we need the following lemma which is due to Maejima and Mason [14].

**Lemma 3.2** Let $D \in \text{End}(\mathbb{R}^d)$. If $\lambda_D > 0$ and $r > 0$, then for any $\delta > 0$, there exist positive constants $K_1$ and $K_2$ such that

$$
\|r^D\| \leq \begin{cases} 
K_1 r^{\lambda_D - \delta}, & \text{for all } r \leq 1, \\
K_2 r^{\lambda_D + \delta}, & \text{for all } r \geq 1.
\end{cases}
$$

(3.4)

Next, we give the detailed proof of Lemma 3.1.

**Proof of Lemma 3.1:** In order to simplify the discussion, we split the proof into two steps.

**Step 1.** We claim that for any $t \in [0, 1],$

$$
\sum_{i=1}^{n} n^2 \left( \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t, s) \, ds \right)^2 \left( \xi_{i,j}^{(n)} \right)^2 \rightarrow \int_{0}^{1} K_{k,j}^2(t, s) \, ds, \text{ a.s.,}
$$

(3.5)

as $n \to \infty$.

For convenience, define

$$
G_n(t, u) = n \sum_{i=1}^{n} 1_{\left[ \frac{i-1}{n}, \frac{i}{n} \right]}(u) \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t, s) \, ds \frac{\xi_{i,j}^{(n)}}{(\sqrt{n})^{-1}}.
$$

Therefore, we have

$$
\int_{0}^{1} G_n^2(t, u) \, du = \sum_{i=1}^{n} n^2 \left( \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t, s) \, ds \right)^2 \left( \xi_{i,j}^{(n)} \right)^2
\leq \sum_{i=1}^{n} n \int_{\frac{i-1}{n}}^{\frac{i}{n}} \left( K_{k,j}(t, s) \right)^2 \, ds \left( \xi_{i,j}^{(n)} \right)^2,
$$

(3.6)

where we have used the Cauchy-Schwartz inequality and by (2.6).

Therefore,

$$
\int_{0}^{1} G_n^2(t, u) \, du \leq \tilde{K} \int_{0}^{1} \left( K_{k,j}(t, s) \right)^2 \, ds.
$$

(3.7)

On the other hand, by (2.2) and Lemma 3.2,

$$
|K_{k,j}(t, s)| \leq \|K(t, s)\| \leq \tilde{K} (t - s)^{(\lambda_D - \delta) - \frac{1}{2}},
$$

(3.8)

since $t - s \in [0, 1]$.

By (3.7) and (3.8), we have

$$
\int_{0}^{1} G_n^2(t, u) \, du \leq \tilde{K} \int_{0}^{1} \left( K_{k,j}(t, s) \right)^2 \, ds \leq \tilde{K} \int_{0}^{1} (t - s)^{2(\lambda_D - \delta) - 1} < \infty,
$$

(3.9)

since $\lambda_D - \delta > \frac{1}{2}$. Therefore, $\{G_n^2(t, u)\}$ is uniformly integrable.

On the other hand, we have for any $u \in (0, 1],$

$$
G_n^2(t, u) \to K_{k,j}^2(t, u), \text{ a.s.,}
$$

(3.10)
since for $u \in \left( \frac{i-1}{n}, \frac{i}{n} \right]$,
\[
\left( n \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t, s) \, ds \right)^2 \to K_{k,j}^2(t, u), \quad \text{as } n \to \infty,
\]
and the condition (2.5).

By (3.9) and (3.10), we get that as $n \to \infty$
\[
\int_0^1 G_n^2(t, u) \, du \to \int_0^1 K_{k,j}^2(t, s) \, ds, \quad \text{a.s.} \quad (3.11)
\]
Therefore, (3.5) holds.

**Step 2.** We prove the original claim. In order to simplify the discussion, we let
\[
t_{l}^n = \left\lfloor \frac{nt_{l}}{n} \right\rfloor \quad \text{and} \quad t_{q}^n = \left\lfloor \frac{nt_{q}}{n} \right\rfloor.
\]
By (3.5), we can get
\[
\sum_{i=1}^{n} n^2 \int_{i-1/n}^{i/n} K_{k,j}(t_l, s) \, ds \int_{i-1/n}^{i/n} K_{k,j}(t_q, s) \, ds \left( \xi_{i,j}^{(n)} \right)^2 \to \int_0^1 K_{k,j}(t_l, s) K_{k,j}(t_q, s) \, ds, \quad \text{a.s.} \quad (3.12)
\]
for $t_l, t_q \in [0, 1]$, as $n \to \infty$.

In fact, it follows from (3.5) that
\[
\sum_{i=1}^{n} n^2 \left( \int_{i-1/n}^{i/n} K_{k,j}(t_l, s) + K_{k,j}(t_q, s) \, ds \right)^2 \left( \xi_{i,j}^{(n)} \right)^2 \to \int_0^1 \left( K_{k,j}(t_l, s) + K_{k,j}(t_q, s) \right)^2 \, ds. \quad (3.13)
\]
On the other hand, we have
\[
\left( \int_{i-1/n}^{i/n} K_{k,j}(t_l, s) + K_{k,j}(t_q, s) \, ds \right)^2 = \left( \int_{i-1/n}^{i/n} K_{k,j}(t_l, s) \, ds \right)^2 + \left( \int_{i-1/n}^{i/n} K_{k,j}(t_q, s) \, ds \right)^2 + 2 \int_{i-1/n}^{i/n} K_{k,j}(t_l, s) \, ds \int_{i-1/n}^{i/n} K_{k,j}(t_q, s) \, ds. \quad (3.14)
\]
Hence (3.5), (3.13), and (3.14) imply (3.12).

Therefore, in order to prove (3.3), it suffices to prove that
\[
\sum_{i=1}^{n} n^2 \left( \int_{i-1/n}^{i/n} K_{k,j}(t_l, s) \, ds \right)^2 \left( \xi_{i,j}^{(n)} \right)^2 \to 0, \quad \text{a.s.} \quad (3.15)
\]
as $n \to \infty$. 

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For the left-hand side of (3.15), we have

\[
\int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t, s) ds \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t_q, s) ds - \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t^n_q, s) ds \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t^n_q, s) ds
\]

\[= \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t, s) ds \int_{\frac{i-1}{n}}^{\frac{i}{n}} \left( K_{k,j}(t_q, s) - K_{k,j}(t^n_q, s) \right) ds
\]

\[- \int_{\frac{i-1}{n}}^{\frac{i}{n}} \left( K_{k,j}(t^n_q, s) - K_{k,j}(t, s) \right) ds \int_{\frac{i-1}{n}}^{\frac{i}{n}} \left( K_{k,j}(t^n_q, s) - K_{k,j}(t_q, s) \right) ds
\]

\[+ \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_{k,j}(t_q, s) ds \int_{\frac{i-1}{n}}^{\frac{i}{n}} \left( K_{k,j}(t, s) - K_{k,j}(t^n_q, s) \right) ds. \tag{3.16}
\]

By (2.2), we have

\[\left| K_{k,j}(t, s) - K_{k,j}(t^n_q, s) \right| \leq \left| K(t, s) - K(t^n_q, s) \right|. \tag{3.17}\]

On the other hand, using the same method as in the proof of the inequality (3.52) below,

\[\sum_{i=1}^{n} \int_{\frac{i-1}{n}}^{\frac{i}{n}} \left\| K(t, s) - K(t^n_q, s) \right\| ds \leq \int_{0}^{1} \left\| K(t, s) - K(t^n_q, s) \right\| ds
\]

\[\leq \tilde{K}(t^n_q - t_q)^{2H}, \tag{3.18}\]

where \(H = \lambda_D - \delta\).

By the condition (2.6) and (3.16), (3.15) can be bounded by

\[
\tilde{K} n \int_{0}^{1} \left\| K(t, s) \right\| ds \int_{0}^{1} \left\| K(t_q, s) - K(t^n_q, s) \right\| ds
\]

\[+ \tilde{K} n \int_{0}^{1} \left\| K(t^n_q, s) - K(t, s) \right\| ds \int_{0}^{1} \left\| K(t^n_q, s) - K(t_q, s) \right\| ds
\]

\[+ \tilde{K} n \int_{0}^{1} \left\| K(t_q, s) \right\| ds \int_{0}^{1} \left\| K(t, s) - K(t^n_q, s) \right\| ds. \tag{3.19}\]

It follows from (3.9), (3.18), and (3.19) that the left-hand side of (3.15) can be bounded by

\[
\tilde{K} n^{1-2H}, \tag{3.20}\]

since \(|t^n_q - t_q| \leq \frac{1}{n}\) and \(|t^n_q - t_l| \leq \frac{1}{n}\).

From (3.20), we can easily prove the lemma. \(\square\)

From the proof of Lemma 3.1 and (2.2), we can easily get that

**Corollary 3.1** Let \(H(t, s) = \sum_{k=1}^{d} a_k K^n_{k,j}(t_l, s)\) for any \(a_k \in \mathbb{R}\). Then

\[
\sum_{i=1}^{n} n^2 \int_{\frac{i-1}{n}}^{\frac{i}{n}} H(t, s) ds \int_{\frac{i-1}{n}}^{\frac{i}{n}} H(t_q, s) ds (\xi_{i,j})^2 \to \int_{0}^{1} H(t, s) H(t_q, s) ds, \text{ a.s.} \tag{3.21}
\]

for any \(t_l, t_q \in (0, 1]\).
Next, we prove the main result of this paper. Before we give the details, we first introduce a technical tool.

**Lemma 3.3** Let \( t \in (0, 1], \sigma_t^2 > 0 \) and let \( \{\xi^{(n)}\} \) be a sequence of martingale differences as in Section 2 and satisfies the following Lindberg condition: for \( \epsilon > 0 \)

\[
\sum_{i=1}^{\lfloor nt \rfloor} \mathbb{E} \left[ (\xi_i^{(n)})^2 I_{\{\lfloor \xi_i^{(n)} \rfloor > \epsilon\}} \bigg| \mathcal{F}_{i-1} \right] \xrightarrow{P} 0.
\]  

(3.22)

Then

\[
\sum_{i=1}^{\lfloor nt \rfloor} (\xi_i^{(n)})^2 \xrightarrow{P} \sigma_t^2
\]

(3.23)

implies

\[
B^n(t) \xrightarrow{D} N \sim N(0, \sigma_t^2),
\]

(3.24)

where \( \xrightarrow{D} \) denotes convergence in distribution.

Lemma 3.3 can be found in Shiryaev [18, p. 511].

**Proof of Theorem 2.1**: We will prove this theorem by two steps.

**Step 1**: First, we have to show that the finite-dimensional distributions of \( X_n \) converge to those of \( X \). It suffices to prove that for any \( q \in \mathbb{N}, a_1, \ldots, a_q \in \mathbb{R} \) and \( t_1, \ldots, t_q \in [0, 1] \),

\[
\sum_{l=1}^{q} a_lX_n(t_l) \xrightarrow{D} \sum_{l=1}^{q} a_lX(t_l).
\]

(3.25)

By the Cramér-Wold device (see, Whittle [19, Chapter 4]), in order to prove (3.25), we only need to show

\[
\sum_{l=1}^{q} a_l b X_n(t_l) \xrightarrow{D} \sum_{l=1}^{q} a_l b X(t_l),
\]

(3.26)

for any vector \( b = (b^{(1)}, \ldots, b^{(d)}) \in \mathbb{R}^d \).

For convenience, define

\[
X_n(t) = \left( X_1^{(n)}(t), \ldots, X_d^{(n)}(t) \right)',
\]

where

\[
X_j^{(n)}(t) = n \sum_{i=1}^{\lfloor nt \rfloor} \int_{\frac{i-1}{n}}^{\frac{i}{n}} K_j^{(n)}(t, s) \eta_i^{(n)} ds,
\]

with

\[
K_j^{(n)}(t, s) = (K_{j,1}^{(n)}(t, s), \ldots, K_{j,d}^{(n)}(t, s)),
\]

and

\[
X(t) = \left( X^{(1)}(t), \ldots, X^{(d)}(t) \right)',
\]
where

\[ X^{(j)}(t) = \int_0^t K_j(t, s)dW(s), \]

with

\[ K_j(t, s) = \left( K_{j,1}(t, s), \cdots, K_{j,d}(t, s) \right). \]

By some calculations, we can get that (3.26) is equivalent to

\[
\sum_{l=1}^{q} \sum_{k=1}^{d} \sum_{j=1}^{d} n \int_{\frac{nt_l}{n}}^{\frac{nt_l}{n}} a_l b^{(k)} K_{k,j}^n(t_l, s) \xi_{i,j}^{(n)} ds \xrightarrow{D} \sum_{l=1}^{q} \sum_{k=1}^{d} \sum_{j=1}^{d} \int_0^t a_l b^{(k)} K_{k,j}(t_l, s)dW^j(s), \quad (3.27)
\]

In order to simplify the discussion, we define

\[ \bar{X}^n(l, k, j) = \sum_{i=1}^{\left\lfloor \frac{nt_l}{n} \right\rfloor} n \int_{\frac{nt_l}{n}}^{\frac{nt_l}{n}} K_{k,j}^n(t_l, s) \xi_{i,j}^{(n)} ds \]

and

\[ \bar{X}(l, k, j) = \int_0^{t_l} K_{k,j}(t_l, s)dW^j(s). \]

Hence (3.27) can be rewritten as follows.

\[
\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} \bar{X}^n(l, k, j) \xrightarrow{D} \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} \bar{X}(l, k, j). \quad (3.28)
\]

By the independence of \( \xi_{i,j}^{(n)}, j = 1, \cdots, d \), it suffices to show that for every \( j \in \{1, \cdots, d\} \)

\[
\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} \bar{X}^n(l, k, j) \xrightarrow{D} \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} \bar{X}(l, k, j). \quad (3.29)
\]

We will prove (3.29) by Lemma 3.3. We first prove the Lindeberg condition holds in our case. For convenience, define:

\[ Z_{k,i}^n(t) = n \int_{\frac{nt_l}{n}}^{\frac{nt_l}{n}} K_{k,j}^n(t_l, s) \xi_{i,j}^{(n)} ds. \]

We have

\[
\left( Z_{k,i}^n(t) \right)^2 = n^2 \left( \xi_{i,j}^{(n)} \right)^2 \left( \int_{\frac{nt_l}{n}}^{\frac{nt_l}{n}} K_{k,j}^n(t_l, s) ds \right)^2 \leq n \left( \xi_{i,j}^{(n)} \right)^2 \int_{\frac{nt_l}{n}}^{\frac{nt_l}{n}} \left( K_{k,j}^n(t_l, s) \right)^2 ds,
\]

where we have used the Hölder inequality. By (2.2), we have

\[
\left| K_{k,j}^n(t_l, s) \right| \leq \| (t - s)^{D - \frac{1}{2}} \|. \quad (3.31)
\]
By Lemma 3.2, we have
\[ \| (t - s)^{D - \frac{1}{2}} \| \leq \tilde{K}(t - s)^{\lambda_D - \frac{1}{2} - \delta}, \] (3.32)
since \( t, s \in [0, 1] \).

By (3.31) and (3.32),
\[ \int_{\frac{1}{n}}^{\frac{n}{n}} \| K_{k, j}(t, s) \|^{2} ds \leq \tilde{K} \int_{\frac{1}{n}}^{\frac{n}{n}} (t - s)^{2(\lambda_D - \delta) - 1} ds \leq \tilde{K} \int_{0}^{1} (1 - s)^{2(\lambda_D - \delta) - 1} ds, \] (3.33)
since \((1 - s)^{2(\lambda_D - \delta) - 1} \) with \( \lambda_D - \delta > \frac{1}{2} \) is decreasing in \( s \). Therefore, it follows from (3.30) and (3.33) that
\[ \left( Z_{k, i}^{n}(t) \right)^{2} \leq \tilde{K} n \left( \xi_{i, j}^{(n)} \right)^{2} \delta_n \] (3.34)
with \( \delta_n = \int_{0}^{1} (1 - s)^{2(\lambda_D - \delta) - 1} ds \).

On the other hand, from (3.2), we get for any \( s \geq \left\lfloor nt \right\rfloor \),
\[ K_{k, j}^{n}(t, s) = 0. \] (3.35)
Hence, by (3.35),
\[ \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} \bar{X}^{n}(l, k, j) = \sum_{i=1}^{n} \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k, i}^{n}(t_l). \] (3.36)
Finally, we have
\[ \left( \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k, i}^{n}(t_l) \right)^{2} \leq \tilde{K} \left( \sum_{l=1}^{q} \sum_{k=1}^{d} b^{(k)} \right)^{2} a_{l}^{2} \left( Z_{k, i}^{n}(t_l) \right)^{2}. \] (3.37)
Combining (3.33) and (3.37), we have
\[ \left( \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k, i}^{n}(t_l) \right)^{2} \leq \tilde{K} n \left( \xi_{i, j}^{(n)} \right)^{2} \delta_n. \] (3.38)
Noting that
\[ \left\{ \left| \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k, i}^{n}(t_l) \right| > \epsilon \right\} = \left\{ \left( \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k, i}^{n}(t_l) \right)^{2} > \epsilon^{2} \right\}, \] (3.39)
from (3.38), we have
\[ \left\{ \left| \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k, i}^{n}(t_l) \right| > \epsilon \right\} \subset \left\{ \tilde{K} n \left( \xi_{i, j}^{(n)} \right)^{2} \delta_n > \epsilon^{2} \right\}. \] (3.40)
Therefore, by (3.38) and (3.40)
\[
\mathbb{E} \left( (\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k,i}^n(t_t))^{2} I_{\{\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k,i}^n(t_t) > \xi \}} \right)_{\mathcal{F}_{i-1}^n}
\leq \tilde{K} n (\xi_{i,j})^{2} \delta_n \mathbb{E} \left( I_{\left\{ \tilde{K} n (\xi_{i,j})^{2} \delta_n > \epsilon^2 \right\}} \right)_{\mathcal{F}_{i-1}^n}
\leq \tilde{K} \delta_{n} \mathbb{E} \left( I_{\left\{ \tilde{K} \delta_n > \epsilon^2 \right\}} \right)_{\mathcal{F}_{i-1}^n}.
\] (3.41)

Combining (3.36) and (3.41), one can easily prove that, as \(n\) approaches \(\infty\),
\[
\sum_{i=1}^{n} \mathbb{E} \left( (\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k,i}^n(t_t))^{2} I_{\{\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k,i}^n(t_t) > \xi \}} \right)_{\mathcal{F}_{i-1}^n} \to 0.
\]

Hence the Lindeberg condition holds.

Next, we show the condition (3.23) holds. We first study the right-hand side of (3.29). We have
\[
\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} \tilde{X}(l, k, j) = \sum_{l=1}^{q} a_l \tilde{W}(t_l),
\] (3.42)

where
\[
\tilde{W}(t) = \int_{0}^{t} \left[ \sum_{k=1}^{d} b^{(k)} K_{k,j}(t, s) \right] dW^j(s) = \int_{0}^{t} \tilde{K}(t, s) dW^j(s),
\] (3.43)

with
\[
\tilde{K}(t, s) = \sum_{k=1}^{d} b^{(k)} K_{k,j}(t, s).
\]

Combining (3.42) and (3.43), we have
\[
\mathbb{E} \left[ \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} \tilde{X}(l, k, j) \right]^{2} = \mathbb{E} \left[ \sum_{l=1}^{q} a_l \tilde{W}(t_l) \right]^{2} = \sum_{j=1}^{q} a_j a_l \int_{0}^{1} \tilde{K}(t_j, s) \tilde{K}(t_l, s) ds.
\] (3.44)

In order to show the condition (3.23), we only need to show
\[
\sum_{i=1}^{n} \left( \sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k,i}^n \right)^{2} \mathbb{P} \to \sum_{l,j=1}^{q} a_j a_l \int_{0}^{1} \tilde{K}(t_j, s) \tilde{K}(t_l, s) ds.
\] (3.45)

Now, we focus on the left-hand side of (3.45). Similar to (3.42), we have
\[
\sum_{l=1}^{q} \sum_{k=1}^{d} a_l b^{(k)} Z_{k,i}^n = \sum_{l=1}^{q} a_l \tilde{Z}_{i,l}^n,
\] (3.46)
where
\[ \bar{Z}_{n,l,i}^n = n \int_{\frac{l-1}{n}}^{\frac{l}{n}} \bar{K}_j^n(t_l, s) \xi_{i,j}^{(n)} \, ds \] (3.47)
with \( \bar{K}_j^n(t_l, s) = \sum_{k=1}^d b^{(k)} K_{k,j}^n(t_l, s) \). Hence
\[ \sum_{i=1}^n \left( \sum_{l=1}^q \sum_{k=1}^d a_{l_1} b^{(k)} Z_{k,i}^n \right)^2 = \sum_{i=1}^n \sum_{l_1, l_2=1}^q n^2 a_{l_1} a_{l_2} \int_{\frac{l_1-1}{n}}^{\frac{l_1}{n}} \bar{K}_j^n(t_l_1, s) ds \right) \left( \bar{K}_j^n(t_l_2, s) \xi_{i,j}^{(n)} \right)^2. \] (3.48)
It follows from Corollary 3.1 that the right-hand side of the equation (3.48) converges to
\[ \sum_{l_1, l_2=1}^q a_{l_1} a_{l_2} \int_0^1 \bar{K}_j^n(t_l_1, s) \bar{K}_j^n(t_l_2, s) ds, \ \text{a.s.} \] (3.49)
as \( n \to \infty \). On the other hand, one can easily get that
\[ \mathbb{E} \left[ \tilde{W}(t_l) \tilde{W}(t_k) \right] = \int_0^1 \tilde{K}(t_l, s) \tilde{K}(t_k, s) ds. \] (3.50)
By (3.44), (3.49), and (3.50), we get the condition (3.23).

**Step 2:** We need to prove the tightness of the sequence \( \{X_n(t)\} \).
By some calculations,
\[ \mathbb{E} \left( \|X_n(t) - X_n(s)\|^2 \right) \leq \tilde{K} \int_0^1 \left\| \left( \frac{nt}{n} - u \right)^{D-\frac{1}{2}} - \left( \frac{ns}{n} - u \right)^{D-\frac{1}{2}} \right\|^2 \, du. \] (3.51)
In order to simplify the discussion, let
\[ \tilde{t} = \frac{nt}{n}, \ \text{and} \ \tilde{s} = \frac{ns}{n}. \]
Next, we show that
\[ \int_0^1 \left\| \left( \tilde{t} - u \right)^{D-\frac{1}{2}} - \left( \tilde{s} - u \right)^{D-\frac{1}{2}} \right\|^2 \, du \leq K (\tilde{t} - \tilde{s})^{2H}, \] (3.52)
where \( H = \lambda_D - \delta \).
In fact,
\[ \int_0^1 \left\| \left( \tilde{t} - u \right)^{D-\frac{1}{2}} - \left( \tilde{s} - u \right)^{D-\frac{1}{2}} \right\|^2 \, du \]
\[ = \int_0^{\tilde{t}} \left\| \left( \tilde{t} - u \right)^{D-\frac{1}{2}} - \left( \tilde{s} - u \right)^{D-\frac{1}{2}} \right\|^2 \, du \]
\[ + \int_{\tilde{s}}^{\tilde{t}} \left\| \left( \tilde{t} - u \right)^{D-\frac{1}{2}} \right\|^2 \, du. \] (3.53)
It follows from Lemma 3.2 and (2.1) that
\[ \| (\tilde{t} - u)^{D - \frac{\lambda}{2}} \| \leq \tilde{K}(\tilde{t} - u)^{\lambda D - \delta - \frac{\lambda}{2}}, \]
since \( u \leq \tilde{t} \in [0, 1] \).
Therefore,
\[
\int_{\tilde{s}}^{\tilde{t}} \| (\tilde{t} - u)^{D - \frac{\lambda}{2}} \|^2 \, du \leq \tilde{K} \int_{\tilde{s}}^{\tilde{t}} (\tilde{t} - u)^{2(\lambda D - \delta) - 1} \, du
= \frac{\tilde{K}(\tilde{t} - \tilde{s})^{2(\lambda D - \delta)}}{2(\lambda D - \delta)}.
\tag{3.54}
\]

Next, we deal with the first term on the right-hand side of (3.53). Note that
\[
\int_{\tilde{s}}^{\tilde{t}} \| (\tilde{t} - u)^{D - \frac{\lambda}{2}} - (\tilde{s} - u)^{D - \frac{\lambda}{2}} \|^2 \, du = \int_{0}^{\tilde{s}} \| (\tilde{t} - s + u)^{D - \frac{\lambda}{2}} - u^{D - \frac{\lambda}{2}} \|^2 \, du
= \int_{0}^{\tilde{s}/(\tilde{t} - \tilde{s})} \| (\tilde{t} - s)(1 + u) \|^{D - \frac{\lambda}{2}} - (\tilde{t} - s)u \|^{D - \frac{\lambda}{2}} \| \cdot (\tilde{t} - \tilde{s}) \, du(\tilde{t} - \tilde{s})
\leq \| (\tilde{t} - s)^{D - \frac{\lambda}{2}} \|^{2(\tilde{t} - \tilde{s})} \int_{0}^{\tilde{s}/(\tilde{t} - \tilde{s})} \| (1 + u)^{D - \frac{\lambda}{2}} - u^{D - \frac{\lambda}{2}} \|^2 \, du
\leq \| (\tilde{t} - s)^{D - \frac{\lambda}{2}} \|^{2(\tilde{t} - \tilde{s})} \int_{\mathbb{R}_+} \| (1 + u)^{D - \frac{\lambda}{2}} - u^{D - \frac{\lambda}{2}} \|^2 \, du,
\tag{3.55}
\]
where we used the fact that \((\tilde{t}s)^A = \tilde{t}^{A} \cdot s^{A}\).

It follows from Lemma 3.2 and (2.1) that
\[ \| (\tilde{t} - s)^{D - \frac{\lambda}{2}} \|^2 (\tilde{t} - \tilde{s}) \leq \tilde{K}(\tilde{t} - \tilde{s})^{2(\lambda D - \delta)}. \]

In order to prove our result, it suffices to show that
\[ \int_{\mathbb{R}_+} \| (1 + u)^{D - \frac{\lambda}{2}} - u^{D - \frac{\lambda}{2}} \|^2 \, du < \infty. \tag{3.56} \]

Then, in order to prove (3.56), it suffices to show that
\[ \int_{u \leq 1} \| u^{D - \frac{\lambda}{2}} \|^2 \, du < \infty, \tag{3.57} \]
and for large enough \( T > 1 \), that
\[ \int_{u \geq T} \| (1 + u)^{D - \frac{\lambda}{2}} - u^{D - \frac{\lambda}{2}} \|^2 \, du < \infty. \tag{3.58} \]

It follows from Lemma 3.2 and (2.1) that
\[ \| u^{D - \frac{\lambda}{2}} \|^2 \leq \tilde{K}u^{2(\lambda D - \delta) - 1} \text{ for } u \leq 1. \]

Hence, one can easily see that (3.56) holds.
Next, we show that (3.58) holds. We see that
\[(1 + u)^{D - \frac{I}{2}} - u^{D - \frac{I}{2}} = \int_u^{1+u} (D - \frac{I}{2}) s^{D - \frac{I}{2}} s^{-1} ds. \tag{3.59}\]

Then
\[\|(1 + u)^{D - \frac{I}{2}} - u^{D - \frac{I}{2}}\| \leq \|(D - \frac{I}{2})\| \int_u^{1+u} \|s^{D - \frac{I}{2}}\| s^{-1} ds. \tag{3.60}\]

It follows from Lemma 2.1 and (2.1) that
\[\int_u^{1+u} \|s^{D - \frac{I}{2}}\| s^{-1} ds \leq \int_u^{1+u} \tilde{K} s^{\Lambda_D + \delta - \frac{I}{2}} ds, \tag{3.61}\]

since \(u \geq 1\).

By (3.60) and (3.61),
\[\|(1 + u)^{D - \frac{I}{2}} - (u)^{D - \frac{I}{2}}\|^2 \leq \tilde{K} u^{2(\Lambda_D + \delta) - 3}. \tag{3.62}\]

By (3.62), we have that (3.58) holds, since \(\Lambda_D + \delta < 1\).

Therefore, we have
\[\mathbb{E}\left(\|X_n(t) - X_n(s)\|^2\right) \leq \tilde{K}(\hat{t} - \hat{s})^{2H}. \tag{3.63}\]

Hence for any \(s \leq t \leq u \in [0, 1]\), we have
\[
\mathbb{E}\left[\|X_n(t) - X_n(s)\|\|X_n(t) - X_n(u)\|\right] \leq \left[\mathbb{E}\|X_n(t) - X_n(s)\|^2\right]^{\frac{1}{2}} \left[\mathbb{E}\|X_n(t) - X_n(u)\|^2\right]^{\frac{1}{2}} \leq \tilde{K} \left(\frac{|nt|}{n} - \frac{|ns|}{n}\right)^H \left(\frac{|nu|}{n} - \frac{|nt|}{n}\right)^H
\leq \tilde{K} \left(\frac{|nu|}{n} - \frac{|ns|}{n}\right)^{2H}. \tag{3.64}\]

If \(u - s \geq \frac{1}{n}\), then one can easily see that
\[\mathbb{E}\left[\|X_n(t) - X_n(s)\|\|X_n(t) - X_n(u)\|\right] \leq \tilde{K}(u - s)^{2H}. \tag{3.65}\]

On the other hand, if \(u - s < \frac{1}{n}\), then either \(s\) and \(t\) or \(t\) and \(u\) belong to the interval \([\frac{i}{n}, \frac{i+1}{n}]\) for some \(i\). Thus the left-hand side of (3.64) is zero. Therefore (3.65) still holds for this case. Hence it follows from Ethier and Kurtz [8, Chapter 3] that \(\{X_n(t)\}\) is tight.

By Theorem 7.8 in Ethier and Kurtz [8, Chapter 3], we get that Theorem 2.1 holds. This completes the proof. \(\square\)

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