

# FRACTIONAL PROCESSES WITH LONG-RANGE DEPENDENCE

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ABSTRACT. We introduce a class of Gaussian processes with stationary increments which exhibit long-range dependence. The class includes fractional Brownian motion with Hurst parameter  $H > 1/2$  as a typical example. We establish infinite and finite past prediction formulas for the processes in which the predictor coefficients are given explicitly in terms of the MA( $\infty$ ) and AR( $\infty$ ) coefficients. We apply the formulas to prove an analogue of Baxter's inequality, which concerns the  $L^1$ -estimate of the difference between the finite and infinite past predictor coefficients.

## 1. INTRODUCTION

Let  $(X(t) : t \in \mathbf{R})$  be a centered Gaussian process with stationary increments, defined on a probability space  $(\Omega, \mathcal{F}, P)$ , that admits the *moving-average* representation

$$(1.1) \quad X(t) = \int_{-\infty}^{\infty} \{g(t-s) - g(-s)\} dW(s), \quad t \in \mathbf{R},$$

where  $(W(t) : t \in \mathbf{R})$  is a Brownian motion, and  $g(t)$  is a function of the form

$$(1.2) \quad g(t) = \int_0^t c(s) ds, \quad t \in \mathbf{R},$$

$$(1.3) \quad c(t) := I_{(0,\infty)}(t) \int_0^{\infty} e^{-ts} \nu(ds), \quad t \in \mathbf{R},$$

with some Borel measure  $\nu$  on  $(0, \infty)$  satisfying

$$(1.4) \quad \int_0^{\infty} \frac{1}{1+s} \nu(ds) < \infty.$$

We will also assume some extra conditions such as

$$(1.5) \quad \lim_{t \rightarrow 0^+} c(t) = \infty,$$

$$(1.6) \quad g(t) \sim t^{H-(1/2)} \ell(t) \cdot \frac{1}{\Gamma(\frac{1}{2} + H)}, \quad t \rightarrow \infty,$$

where  $\ell(t)$  is a slowly varying function at infinity and  $H$  is a constant such that

$$(1.7) \quad 1/2 < H < 1.$$

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In (1.6), and throughout the paper,  $a(t) \sim b(t)$  as  $t \rightarrow \infty$  means  $\lim_{t \rightarrow \infty} a(t)/b(t) = 1$ . We call  $c(t)$  (rather than  $g(t)$ ) the *MA( $\infty$ ) coefficient* of  $(X(t))$ .

A typical example of  $\nu$  is

$$(1.8) \quad \nu(ds) = \frac{\sin\{\pi(H - \frac{1}{2})\}}{\pi} s^{1/2-H} ds \quad \text{on } (0, \infty)$$

with (1.7). For this  $\nu$ ,  $g(t)$  becomes

$$(1.9) \quad g(t) = I_{(0, \infty)}(t) t^{H-1/2} \frac{1}{\Gamma(\frac{1}{2} + H)}, \quad t \in \mathbf{R},$$

and  $(X(t))$  reduces to *fractional Brownian motion* ( $B_H(t)$ ) with *Hurst parameter*  $H$  (see Example 2.3 below). Fractional Brownian motion, abbreviated fBm, was introduced by Kolmogorov [20]. For  $1/2 < H < 1$ , fBm has both *self-similarity* and *long-range dependence* (Samorodnitsky and Taqqu [27]), and plays an important role in various fields such as network traffic (see, e.g., Mikosch et al. [22]) and finance (see, e.g., Hu et al. [10]); see also Taqqu [28] and other papers in the same volume. Because of its importance, stochastic calculus for fBm has been developed by many authors; see, e.g., Decreusefond and Üstünel [8], and Nualart [24]. Other important examples of  $(X(t))$  are the processes with long-range dependence which, unlike fBm, have two different indices  $H_0$  and  $H$  describing the local properties (path properties) and long-time behavior of  $(X(t))$ , respectively (see Example 2.4 below).

Let  $t_0, t_1$  and  $T$  be real constants such that

$$(1.10) \quad -\infty < -t_0 \leq 0 \leq t_1 < T < \infty, \quad -t_0 < t_1.$$

For  $I = (-\infty, t_1]$  or  $[-t_0, t_1]$ , we write  $P_I X(T)$  for the predictor of the future value  $X(T)$  based on the observable  $(X(s) : s \in I)$  (see Section 3 below). One of the fundamental prediction problems for  $(X(t))$  is to express  $P_I X(T)$  using the segment  $(X(s) : s \in I)$  and some deterministic quantities. Another is to express the variance of the prediction error  $P_I^\perp X(T) := X(T) - P_I X(T)$ . Results of this type become important tools in the analysis of non-Markovian processes and systems modulated by them (see, e.g., Norros et al. [23], Anh et al. [3], Inoue et al. [19] and Inoue and Nakano [18]). One of our main purposes here is to derive such results for  $(X(t))$ .

We establish the following infinite and finite past prediction formulas for  $(X(t))$  (see Theorems 3.8 and 4.12 below):

$$(1.11) \quad P_{(-\infty, t_1]} X(T) = X(t_1) + \int_{-\infty}^{t_1} \left\{ \int_0^{T-t_1} b(t_1 - s, \tau) d\tau \right\} dX(s),$$

$$(1.12) \quad P_{[-t_0, t_1]} X(T) = X(t_1) + \int_{-t_0}^{t_1} \left\{ \int_0^{T-t_1} h(s + t_0, u) du \right\} dX(s).$$

The significance of (1.11) and (1.12) is that the predictor coefficients  $b(t, s)$  and  $h(t, s)$  are given explicitly in terms of the *MA( $\infty$ ) coefficient*  $c(t)$  and *AR( $\infty$ ) coefficient*  $a(t)$  of  $(X(t))$ . We will find that  $a(t)$  has a nice integral representation similar to (1.3) (see (3.3) below). It turns out that the existence of such a nice *AR( $\infty$ ) coefficient*, in addition to the nice *MA( $\infty$ ) coefficient*, is a key to the solution to the prediction problems above.

We apply the results above to the proof of *Baxter's inequality* for  $(X(t))$ , which concerns the  $L^1$ -estimate of the difference between the predictor coefficients  $b(t, s)$

and  $h(t, s)$ . The original inequality of Baxter [4] is an assertion for stationary time series  $(Y_n : n \in \mathbf{N})$  with short memory. It takes the form

$$(1.13) \quad \sum_{j=1}^n |\phi_{n,j} - \phi_j| \leq K \sum_{k=n+1}^{\infty} |\phi_k|, \quad \forall n \geq 1,$$

where  $K$  is a positive constant, and  $\phi_j$  and  $\phi_{n,j}$  are the infinite and finite past predictor coefficients in

$$P_{(-\infty, -1]} Y_0 = \sum_{j=1}^{\infty} \phi_j Y_{-j}, \quad P_{[-n, -1]} Y_0 = \sum_{j=1}^n \phi_{n,j} Y_{-j},$$

respectively, with  $P_{(-\infty, -1]} Y_0$  and  $P_{[-n, -1]} Y_0$  being defined similarly. See Berk [5], Cheng and Pourahmadi [7], and Inoue and Kasahara [17] for related work; for a textbook account, see Pourahmadi [26, Section 7.6.2]. Using the explicit representations of  $b(t, s)$  and  $h(t, s)$ , we can prove an analogue of (1.13) for  $(X(t))$  which are continuous-time stationary-increment processes with long-range dependence.

For fBm with  $1/2 < H < 1$ , the predictor coefficients  $b(t, s)$  and  $h(t, s)$  are given in Gripenberg and Norros [9] (see (3.13) and (5.3) below). See [23] and [25] for different proofs. Fractional Brownian motion has a variety of nice properties, and the methods of proof of [9, 23, 25] naturally rely on such special properties of fBm, hence are not applicable to  $(X(t))$ . The method of this paper is based on the *alternating projections to the past and future* (see Section 4.1 below). As for fBm with  $0 < H < 1/2$ , its infinite and finite past prediction formulas also exist, and are due to Yaglom [29] and Nuzman and Poor [25], respectively (see also Anh and Inoue [2]); see Inoue and Anh [15] for an extension to these results, which have different forms from (1.11) and (1.12) since no stochastic integrals appear there.

We provide the basic properties and examples of  $(X(t))$  in Section 2. We consider the infinite and finite past prediction problems for  $(X(t))$  in Sections 3 and 4, respectively. In Section 5, we prove an analogue of Baxter's inequality for  $(X(t))$ , using the results in Sections 3 and 4.

## 2. BASIC PROPERTIES AND EXAMPLES

In this section, we assume (1.2)–(1.4) and

$$(2.1) \quad \int_1^{\infty} c(t)^2 dt < \infty.$$

Then, as in [15, Lemma 2.1], we have  $\int_{-\infty}^{\infty} |g(t-s) - g(-s)|^2 ds < \infty$  for  $t \in \mathbf{R}$ . Therefore, for a one-dimensional standard Brownian motion  $(W(t) : t \in \mathbf{R})$  with  $W(0) = 0$ , we may define the centered stationary-increment Gaussian process  $(X(t) : t \in \mathbf{R})$  by (1.1).

For  $s > 0$  and  $t \in \mathbf{R}$ , we put  $\Delta_s X(t) := X(t+s) - X(t)$ . Then, by definition,  $(\Delta_s X(t) : t \in \mathbf{R})$  is a stationary process.

**Lemma 2.1.** *Let  $s \in (0, \infty)$ . We assume (1.6) and (1.7). Then*

$$E[\Delta_s X(t) \cdot \Delta_s X(0)] \sim t^{2H-2} \ell(t)^2 \cdot \frac{s^2 \Gamma(2-2H) \sin\{(H-\frac{1}{2})\pi\}}{\pi}, \quad t \rightarrow \infty.$$

Since  $-1 < 2H - 2 < 0$  in Lemma 2.1, we see from this lemma that  $(\Delta_s X(t))$ , whence  $(X(t))$ , has long-range dependence.

We put  $\sigma(t) := E[|X(t+s) - X(s)|^2]^{1/2}$  for  $t \geq 0$  and  $s \in \mathbf{R}$ .

**Lemma 2.2.** *Let  $H_0 \in (1/2, 1)$  and  $\ell_0(\cdot)$  a slowly varying function at infinity. We assume*

$$(2.2) \quad g(t) \sim t^{H_0-(1/2)} \ell_0(1/t) \cdot \frac{1}{\Gamma(\frac{1}{2} + H_0)}, \quad t \rightarrow 0+.$$

Then

$$\sigma(t) \sim t^{H_0} \ell(1/t) \sqrt{v(H_0)}, \quad t \rightarrow 0+,$$

where  $v(H_0) := \Gamma(2 - 2H_0) \cos(\pi H_0) / \{\pi H_0(1 - 2H_0)\}$ . In particular, we have

$$H_0 = \sup\{\beta : \sigma(t) = o(t^\beta), \quad t \rightarrow 0+\} = \inf\{\beta : t^\beta = o(\sigma(t)), \quad t \rightarrow 0+\}.$$

From Lemma 2.2, we see that the index  $H_0$  describes the path properties of  $(X(t))$  (see Adler [1, Section 8.4]).

By the monotone density theorem (cf. Bingham et al. [6, Theorem 1.7.5]), (1.6) with (1.7) implies

$$(2.3) \quad c(t) \sim t^{H-(3/2)} \ell(t) \cdot \frac{1}{\Gamma(H - \frac{1}{2})}, \quad t \rightarrow \infty.$$

Similarly, (2.2) implies

$$(2.4) \quad c(t) \sim t^{H_0-(3/2)} \ell_0(1/t) \cdot \frac{1}{\Gamma(H_0 - \frac{1}{2})}. \quad t \rightarrow 0+.$$

Lemmas 2.1 and 2.2 follow from (2.3) and (2.4), respectively, by standard arguments. However, since we do not use these results in the arguments below, we omit the details.

**Example 2.3.** For  $H \in (1/2, 1)$ , let  $\nu$  be as in (1.8). Then we have (1.9); and so all the conditions above are satisfied. The resulting process  $(X(t))$  is fBm  $(B_H(t))$ :

$$(2.5) \quad B_H(t) = \frac{1}{\Gamma(\frac{1}{2} + H)} \int_{-\infty}^{\infty} \left\{ ((t-s)_+)^{H-(1/2)} - ((-s)_+)^{H-(1/2)} \right\} dW(s),$$

where  $(x)_+ := \max(0, x)$  for  $x \in \mathbf{R}$ . The representation (2.5) of fBm is due to the pioneering work of Mandelbrot and Van Ness [21].

**Example 2.4.** Let  $f(\cdot)$  be a nonnegative, locally integrable function on  $(0, \infty)$ . For  $H_0, H \in (1/2, 1)$  and slowly varying functions  $\ell_0(\cdot)$  and  $\ell(\cdot)$  at infinity, we assume

$$\begin{aligned} f(s) &\sim \frac{\sin\{\pi(H_0 - \frac{1}{2})\}}{\pi} s^{(1/2)-H} \ell(1/s), & s \rightarrow 0+, \\ f(s) &\sim \frac{\sin\{\pi(H_0 - \frac{1}{2})\}}{\pi} s^{(1/2)-H_0} \ell_0(s), & s \rightarrow \infty. \end{aligned}$$

Let  $\nu(ds) = f(s)ds$ . Then, by Abelian theorems for Laplace transforms (cf. [6, Section 1.7]), we have (2.3), whence (1.6). Similarly, we have (2.4), whence (2.2). Thus all the conditions above are satisfied. As we have seen above, the indices  $H_0$  and  $H$  describe the path properties and long-time behavior of  $(X(t))$ , respectively.

### 3. INFINITE PAST PREDICTION PROBLEMS

In this section, we assume (1.1)–(1.5), (2.1) and

$$(3.1) \quad \lim_{t \rightarrow \infty} g(t) = \infty.$$

Notice that, for the processes  $(X(t))$  in Examples 2.3 and 2.4, all these conditions are satisfied. We also assume (1.10).

We write  $M(X)$  for the real Hilbert space spanned by  $(X(t) : t \in \mathbf{R})$  in  $L^2(\Omega, \mathcal{F}, P)$ , and  $\|\cdot\|$  for its norm. Let  $I$  be a closed interval of  $\mathbf{R}$  such as  $[-t_0, t_1]$ ,  $(-\infty, t_1]$ , and  $[-t_0, \infty)$ . Let  $M_I(X)$  be the closed subspace of  $M(X)$  spanned by  $(X(t) : t \in I)$ . We write  $P_I$  for the orthogonal projection operator from  $M(X)$  to  $M_I(X)$ , and  $P_I^\perp$  for its orthogonal complement:  $P_I^\perp Z = Z - P_I Z$  for  $Z \in M(X)$ . Note that, since  $(X(t))$  is a Gaussian process, we have

$$P_I Z = E[Z | \sigma(X(s) : s \in I)], \quad Z \in M(X).$$

**3.1. MA and AR coefficients.** The conditions (1.5) and (3.1) imply  $\nu(0, \infty) = \infty$  and  $\int_0^\infty s^{-1} \nu(ds) = \infty$ , respectively. Therefore, by [15, Theorem 3.2], there exists a unique Borel measure  $\mu$  on  $(0, \infty)$  satisfying

$$\int_0^\infty \frac{1}{1+s} \mu(ds) < \infty, \quad \mu(0, \infty) = \infty, \quad \int_0^\infty \frac{1}{s} \mu(ds) = \infty$$

and

$$(3.2) \quad -iz \left\{ \int_0^\infty e^{izt} c(t) dt \right\} \left\{ \int_0^\infty e^{izt} \alpha(t) dt \right\} = 1, \quad \Im z > 0,$$

with

$$\alpha(t) := \int_0^\infty e^{-st} \mu(ds), \quad t > 0.$$

We define the *AR*( $\infty$ ) coefficient  $a(t)$  of  $(X(t))$  by

$$(3.3) \quad a(t) := -\frac{d\alpha}{dt}(t) = \int_0^\infty e^{-st} s \mu(ds), \quad t > 0.$$

We define the positive kernel  $b(t, s)$  by

$$b(t, s) := \int_0^s c(u) a(t + s - u) du, \quad t, s > 0.$$

Then, by [15, Lemma 3.4], the following equalities hold:

$$(3.4) \quad \int_0^\infty b(t, s) dt = 1, \quad s > 0,$$

$$(3.5) \quad c(t + s) = \int_0^t c(t - u) b(u, s) du, \quad t, s > 0.$$

**3.2. Stochastic integrals.** Let  $I$  be a closed interval of  $\mathbf{R}$ . We define

$$\mathcal{H}_I(X) := \left\{ f : \begin{array}{l} f \text{ is a real-valued measurable function on } I \text{ such} \\ \text{that } \int_{-\infty}^\infty \left\{ \int_I |f(u)| c(u-s) du \right\}^2 ds < \infty. \end{array} \right\}.$$

This is the class of functions  $f$  for which we can define the stochastic integral  $\int_I f(s) dX(s)$ . We define a subclass  $\mathcal{H}_I^0$  of  $\mathcal{H}_I(X)$  by

$$\mathcal{H}_I^0 := \left\{ \sum_{k=1}^m a_k I_{(t_{k-1}, t_k]}(s) : \begin{array}{l} m \in \mathbf{N}, -\infty < t_0 < t_1 < \dots < t_m < \infty \\ \text{with } (t_0, t_m] \subset I, a_k \in \mathbf{R} \ (k = 1, \dots, m) \end{array} \right\}.$$

Each member of  $f \in \mathcal{H}_I^0$  a simple function on  $I$ .

**Definition 3.1.** For  $f = \sum_{k=1}^m a_k I_{(t_{k-1}, t_k]} \in \mathcal{H}_I^0$ , we define

$$\int_I f(s) dX(s) := \sum_{k=1}^m a_k \{X(t_k) - X(t_{k-1})\}.$$

We see that  $\int_I f(s) dX(s) \in M_I(X)$  for  $f \in \mathcal{H}_I^0$ .

**Proposition 3.2.** For  $f \in \mathcal{H}_I^0$ , we have

$$(3.6) \quad \int_I f(s) dX(s) = \int_{-\infty}^{\infty} \left\{ \int_I f(u) c(u-s) du \right\} dW(s).$$

*Proof.* For  $-\infty < a < b < \infty$  with  $(a, b] \subset I$ , we have

$$X(b) - X(a) = \int_{-\infty}^{\infty} \left\{ \int_I I_{(a,b]}(u) c(u-s) du \right\} dW(s),$$

which implies (3.6) for  $f = I_{(a,b]}$ . The general case follows easily from this.  $\square$

**Proposition 3.3.** Let  $f \in \mathcal{H}_I(X)$  such that  $f \geq 0$ , and let  $f_n$  ( $n = 1, 2, \dots$ ) be a sequence of simple functions on  $I$  such that  $0 \leq f_n \uparrow f$  a.e. Then, in  $M(X)$ ,

$$\lim_{n \rightarrow \infty} \int_{-\infty}^{\infty} f_n(s) dX(s) = \int_{-\infty}^{\infty} \left\{ \int_I f(u) c(u-s) du \right\} dW(s).$$

*Proof.* By Proposition 3.2 and the monotone convergence theorem, we have

$$\begin{aligned} & \left\| \int_I f_n(s) dX(s) - \int_{-\infty}^{\infty} \left\{ \int_I f(u) c(u-s) du \right\} dW(s) \right\|^2 \\ & \leq \int_{-\infty}^{\infty} \left\{ \int_I (f(u) - f_n(u)) c(u-s) du \right\}^2 ds \downarrow 0, \quad n \rightarrow \infty. \end{aligned}$$

Thus the proposition follows.  $\square$

For a real-valued function  $f$  on  $I$ , we write  $f(x) = f^+(x) - f^-(x)$ , where

$$f^+(x) := \max(f(x), 0), \quad f^-(x) := \max(-f(x), 0), \quad x \in I.$$

**Definition 3.4.** For  $f \in \mathcal{H}_I(X)$ , we define

$$\int_I f(s) dX(s) := \lim_{n \rightarrow \infty} \int_I f_n^+(s) dX(s) - \lim_{n \rightarrow \infty} \int_I f_n^-(s) dX(s) \quad \text{in } M(X),$$

where  $\{f_n^+\}$  and  $\{f_n^-\}$  are arbitrary sequences of non-negative simple functions on  $I$  such that  $f_n^+ \uparrow f^+$ ,  $f_n^- \uparrow f^-$ , as  $n \rightarrow \infty$ , a.e.

From the definition above, we see that  $\int_I f(s) dX(s) \in M_I(X)$  for  $f \in \mathcal{H}_I(X)$ . The next proposition follows immediately from Proposition 3.3.

**Proposition 3.5.** The equality (3.6) also holds for  $f \in \mathcal{H}_I(X)$ .

**3.3. Infinite past prediction formulas.** We denote by  $\mathcal{D}(\mathbf{R})$  the space of all  $\phi \in C^\infty(\mathbf{R})$  with compact support, endowed with the usual topology. For a random distribution  $Y$  (cf. [11, Section 2] and [3, Section 2]), we write  $DY$  for its derivative. For  $t \in \mathbf{R}$ , we write  $M_{(-\infty, t]}(Y)$  for the closed linear hull of  $\{Y(\phi) : \phi \in \mathcal{D}(\mathbf{R}), \text{supp } \phi \subset (-\infty, t]\}$  in  $L^2(\Omega, \mathcal{F}, P)$ . Notice that  $M_I(X)$  here coincides with that defined above.

As in [15, Proposition 2.4], we have the next proposition.

**Proposition 3.6.** *The derivative  $DX$  of  $(X(t))$  is a purely nondeterministic stationary random distribution, and  $(W(t) : t \in \mathbf{R})$  is a canonical Brownian motion of  $DX$  in the sense that  $M_{(-\infty, t]}(DX) = M_{(-\infty, t]}(DW)$  for every  $t \in \mathbf{R}$ .*

Here is the infinite past prediction formula for  $\int_t^\infty f(s)dX(s)$ .

**Theorem 3.7.** *For  $t \in [0, \infty)$  and  $f \in \mathcal{H}_{[t, \infty)}(X)$ , the following assertions hold:*

- (a)  $\int_0^\infty b(t - \cdot, \tau)f(t + \tau)d\tau \in \mathcal{H}_{(-\infty, t]}(X)$ .
- (b)  $P_{(-\infty, t]} \int_t^\infty f(s)dX(s) = \int_{-\infty}^t \left\{ \int_0^\infty b(t - s, \tau)f(t + \tau)d\tau \right\} dX(s)$ .

*Proof.* Since  $f \in \mathcal{H}_{[t, \infty)}(X)$  iff  $|f| \in \mathcal{H}_{[t, \infty)}(X)$ , we may assume  $f \geq 0$ . Since

$$(3.7) \quad c(u) = 0, \quad t \leq 0,$$

it follows from (3.5) and the Fubini–Tonelli theorem that, for  $s < t$ ,

$$(3.8) \quad \begin{aligned} \int_t^\infty f(u)c(u - s)du &= \int_0^\infty d\tau f(t + \tau) \int_0^{t-s} c(t - s - u)b(u, \tau)du \\ &= \int_{-\infty}^t duc(u - s) \int_0^\infty b(t - u, \tau)f(t + \tau)d\tau. \end{aligned}$$

Thus we obtain (a). By Proposition 3.6 and [3, Proposition 2.3 (2)], we have

$$(3.9) \quad M_{(-\infty, t]}(X) = M_{(-\infty, t]}(DW).$$

This and Proposition 3.5 yield

$$P_{(-\infty, t]} \int_t^\infty f(s)dX(s) = \int_{-\infty}^t \left\{ \int_t^\infty f(u)c(u - s)du \right\} dW(s).$$

By (3.7), (3.8) and Proposition 3.5, the integral on the right-hand side is

$$\begin{aligned} &\int_{-\infty}^t \left\{ \int_{-\infty}^t duc(u - s) \int_0^\infty b(t - u, \tau)f(t + \tau)d\tau \right\} dW(s) \\ &= \int_{-\infty}^t \left\{ \int_0^\infty b(t - s, \tau)f(t + \tau)d\tau \right\} dX(s). \end{aligned}$$

Thus (b) follows. □

By putting  $f(s) = I_{(t_1, T]}(s)$  in Theorem 3.7 (b), we immediately obtain the next infinite past prediction formula for  $(X(t))$ .

**Theorem 3.8.** *Let  $0 \leq t_1 < T < \infty$ . Then  $\int_0^{T-t_1} b(t_1 - \cdot, \tau)d\tau \in \mathcal{H}_{(-\infty, t_1]}(X)$  and the infinite past prediction formula (1.11) holds.*

Using the Hilbert space isomorphism  $\theta : M(X) \rightarrow M(X)$  characterized by  $\theta(X(t)) = X(-t)$  for  $t \in \mathbf{R}$ , we obtain the next theorem from Theorem 3.7 (see the proof of [3, Theorem 3.6]).

**Theorem 3.9.** For  $t \in [0, \infty)$  and  $f \in \mathcal{H}_{[t, \infty)}(X)$ , the following assertions hold:

- (a)  $\int_0^\infty b(t + \cdot, \tau) f(t + \tau) d\tau \in \mathcal{H}_{[-t, \infty)}(X)$ .
- (b)  $P_{[-t, \infty)} \int_{-\infty}^{-t} f(-s) dX(s) = \int_{-t}^\infty \left\{ \int_0^\infty b(t + s, \tau) f(t + \tau) d\tau \right\} dX(s)$ .

As in [3, Definition 2.2], we define another Brownian motion  $(W^*(t) : t \in \mathbf{R})$  by

$$(3.10) \quad W^*(t) := \theta(W(-t)), \quad t \in \mathbf{R}.$$

**Proposition 3.10.** Let  $I$  be a closed interval of  $\mathbf{R}$  and let  $f \in \mathcal{H}_I(X)$ . Then

$$\int_I f(s) dX(s) = \int_{-\infty}^\infty \left\{ \int_I f(u) c(s - u) du \right\} dW^*(s).$$

The proof of Proposition 3.10 is the same as that of [3, Proposition 3.5], whence we omit it. We need Theorem 3.9 and Proposition 3.10 in the next section.

**Example 3.11.** As in Example 2.3, we consider fBm  $(B_H(t))$  with  $1/2 < H < 1$ . Then the MA( $\infty$ ) coefficient  $c(t)$  is given by

$$(3.11) \quad c(t) = t^{H-3/2} \frac{1}{\Gamma(H - \frac{1}{2})}, \quad t > 0,$$

so that  $\int_0^\infty e^{izt} c(t) dt = (-iz)^{1/2-H}$  for  $\Im z > 0$ . From (3.2), we have

$$\int_0^\infty e^{izt} \alpha(t) dt = (-iz)^{H-3/2}.$$

Hence,  $\alpha(t) = t^{\frac{1}{2}-H} / \Gamma(\frac{3}{2} - H)$ , so that the AR( $\infty$ ) coefficient  $a(t)$  is given by

$$(3.12) \quad a(t) = t^{-(H+\frac{1}{2})} \frac{H - \frac{1}{2}}{\Gamma(\frac{3}{2} - H)}, \quad t > 0.$$

By the change of variable  $u = sv$ ,  $\int_0^s (s - u)^{H-(3/2)} (t + u)^{-H-(1/2)} du$  becomes

$$s^{H-\frac{1}{2}} t^{-H-\frac{1}{2}} \int_0^1 (1 - v)^{H-\frac{3}{2}} \{1 + (s/t)v\}^{-H-\frac{1}{2}} dv = \frac{1}{(H - \frac{1}{2})} \left(\frac{s}{t}\right)^{H-\frac{1}{2}} \frac{1}{t + s},$$

where we have used the equality

$$\int_0^1 (1 - v)^{p-1} (1 + xv)^{-p-1} dv = \frac{1}{p(x+1)}, \quad p > 0, \quad x > -1.$$

Thus

$$(3.13) \quad b(t, s) = \frac{\sin\{\pi(H - \frac{1}{2})\}}{\pi} \left(\frac{s}{t}\right)^{H-\frac{1}{2}} \frac{1}{t + s}, \quad t > 0, \quad s > 0;$$

and so, from Theorem 3.8, we see that, for  $0 \leq t < T$ ,

$$\begin{aligned} & E[B_H(T) | \sigma(B_H(s) : -\infty < s \leq t)] \\ &= B_H(t) + \frac{\sin\{\pi(H - \frac{1}{2})\}}{\pi} \int_{-\infty}^t \left\{ \int_0^{T-t} \left(\frac{\tau}{t-s}\right)^{H-\frac{1}{2}} \frac{1}{t-s+\tau} d\tau \right\} dB_H(s). \end{aligned}$$

This prediction formula was obtained in [9, Theorem 3.1] by a different method.

#### 4. FINITE PAST PREDICTION PROBLEMS

In this section, we assume (1.1)–(1.7) and (1.10). Notice that (1.6) with (1.7) implies (3.1) as well as (2.3), whence (2.1). For  $t_0, t_1$ , and  $T$  in (1.10), we put

$$t_2 := t_0 + t_1, \quad t_3 := T - t_1.$$



**4.1. Alternating projections to the past and future.** For  $n \in \mathbf{N}$ , we define the orthogonal projection operator  $P_n$  by

$$P_n := \begin{cases} P_{(-\infty, t_1]}, & n = 1, 3, 5, \dots, \\ P_{[-t_0, \infty)}, & n = 2, 4, 6, \dots \end{cases}$$

It should be noted that  $\{P_n\}_{n=1}^\infty$  is merely an alternating sequence of projection operators, first to  $M_{(-\infty, t_1]}(X)$ , then to  $M_{[-t_0, \infty)}(X)$ , and so on. This sequence plays a key role in the proof of the finite past prediction formula for  $(X(t))$ .

For  $t, s \in (0, \infty)$  and  $n \in \mathbf{N}$ , we define  $b_n(t, s) = b_n(t, s; t_2)$  iteratively by

$$(4.1) \quad \begin{cases} b_1(t, s) := b(t, s), \\ b_n(t, s) := \int_0^\infty b(t, u) b_{n-1}(t_2 + u, s) du, \quad n = 2, 3, \dots \end{cases}$$

**Proposition 4.1.** *For  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ , the following assertions hold:*

- (a)  $\int_0^\infty b_n(t_1 - \cdot, \tau) f(t_1 + \tau) d\tau \in \mathcal{H}_{(-\infty, t_1]}(X)$  for  $n = 1, 3, 5, \dots$
- (b)  $\int_0^\infty b_n(t_0 + \cdot, \tau) f(t_1 + \tau) d\tau \in \mathcal{H}_{[-t_0, \infty)}(X)$  for  $n = 2, 4, 6, \dots$

*Proof.* We may assume that  $f \geq 0$ . By Theorem 3.7, (a) holds for  $n = 1$ . By the Fubini–Tonelli theorem, we have, for  $s > -t_0$ ,

$$\int_0^\infty du b(t_0 + s, u) \int_0^\infty b_1(t_2 + u, \tau) f(t_1 + \tau) d\tau = \int_0^\infty b_2(t_0 + s, \tau) f(t_1 + \tau) d\tau.$$

Hence, by Theorem 3.9, we have (b) for  $n = 2$ . Repeating this procedure, we obtain the proposition.  $\square$

Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . By Proposition 4.1, we may define the random variables  $G_n(f)$  by

$$G_n(f) := \begin{cases} \int_{-t_0}^{t_1} \left\{ \int_0^\infty b_n(t_1 - s, \tau) f(t_1 + \tau) d\tau \right\} dX(s), & n = 1, 3, \dots, \\ \int_{-t_0}^{t_1} \left\{ \int_0^\infty b_n(t_0 + s, \tau) f(t_1 + \tau) d\tau \right\} dX(s), & n = 2, 4, \dots \end{cases}$$

We may also define the random variables  $\epsilon_n(f)$  by  $\epsilon_0(f) := \int_{t_1}^\infty f(s) dX(s)$  and

$$\epsilon_n(f) := \begin{cases} \int_{-\infty}^{-t_0} \left\{ \int_0^\infty b_n(t_1 - s, \tau) f(t_1 + \tau) d\tau \right\} dX(s), & n = 1, 3, \dots, \\ \int_{t_1}^\infty \left\{ \int_0^\infty b_n(t_0 + s, \tau) f(t_1 + \tau) d\tau \right\} dX(s), & n = 2, 4, \dots \end{cases}$$

**Proposition 4.2.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$  and  $n \in \mathbf{N}$ . Then*

$$(4.2) \quad P_n P_{n-1} \cdots P_1 \int_{t_1}^\infty f(s) dX(s) = \epsilon_n(f) + \sum_{k=1}^n G_k(f).$$

We can prove (4.2) using Proposition 4.1 and the facts

$$(4.3) \quad M_{[-t_0, t_1]}(X) \subset M_{(-\infty, t_1]}(X) \cap M_{[-t_0, \infty)}(X),$$

$$(4.4) \quad G_k \in M_{[-t_0, t_1]}(X), \quad k = 1, 2, \dots$$

Since the proof is similar to that of [3, Proposition 4.4], we omit the details.

We are about to investigate the limit of (4.2) as  $n \rightarrow \infty$  (see Lemma 4.9 below).

For  $f \in \mathcal{H}_{[t_1, \infty)}(X)$  and  $s > 0$ , we define  $D_n(s, f) = D_n(s, f; t_1, t_2)$  by

$$D_n(s, f) := \begin{cases} \int_0^\infty c(u) f(t_1 + s + u) du, & n = 0, \\ \int_0^\infty du c(u) \int_0^\infty b_n(t_2 + u + s, \tau) f(t_1 + \tau) d\tau, & n = 1, 2, \dots \end{cases}$$

From the proof of the next proposition, we see that these integrals converge absolutely. Recall  $(W^*(t))$  from (3.10).

**Proposition 4.3.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . Then*

$$P_{n+1}^\perp \epsilon_n(f) = \begin{cases} \int_{t_1}^\infty D_n(s - t_1, f) dW(s), & n = 0, 2, 4, \dots, \\ \int_{-\infty}^{-t_0} D_n(-t_0 - s, f) dW^*(s), & n = 1, 3, 5, \dots \end{cases}$$

*Proof.* By (3.9) and Proposition 3.5,

$$P_1^\perp \epsilon_0(f) = \int_{t_1}^\infty \left\{ \int_s^\infty f(u) c(u - s) du \right\} dW(s) = \int_{t_1}^\infty D_0(s - t_1, f) dW(s).$$

Thus the assertion holds for  $n = 0$ . Let  $n = 1, 3, \dots$ . Then, by Proposition 3.10,

$$\epsilon_n(f) = \int_{-\infty}^\infty \left\{ \int_{-\infty}^{-t_0} duc(s - u) \int_0^\infty b_n(t_1 - u, \tau) f(t_1 + \tau) d\tau \right\} dW^*(s).$$

Hence, using [3, Proposition 2.3 (7)] and (3.7),

$$\begin{aligned} P_{n+1}^\perp \epsilon_n(f) &= \int_{-\infty}^{-t_0} \left\{ \int_{-\infty}^s duc(s - u) \int_0^\infty b_n(t_1 - u, \tau) f(t_1 + \tau) d\tau \right\} dW^*(s) \\ &= \int_{-\infty}^{-t_0} \left\{ \int_0^\infty duc(u) \int_0^\infty b_n(t_2 + u - t_0 - s, \tau) f(t_1 + \tau) d\tau \right\} dW^*(s) \\ &= \int_{-\infty}^{-t_0} D_n(-t_0 - s, f) dW^*(s). \end{aligned}$$

Thus we obtain the assertion for  $n = 1, 3, \dots$ . The proof for  $n = 2, 4, \dots$  is similar; and so we omit it.  $\square$

From Propositions 4.2 and 4.3, we immediately obtain the next proposition (cf. the proof of [3, Proposition 4.9]).

**Proposition 4.4.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . Then the following assertions hold:*

- (a)  $\|P_1^\perp \int_{t_1}^\infty f(s) dX(s)\|^2 = \int_0^\infty D_0(s, f)^2 ds.$
- (b)  $\|P_{n+1}^\perp P_n P_{n-1} \cdots P_1 \int_{t_1}^\infty f(s) dY(s)\|^2 = \int_0^\infty D_n(s, f)^2 ds$  for  $n = 1, 2, \dots$

We write  $Q$  for the orthogonal projection operator from  $M(X)$  onto the intersection  $M_{(-\infty, t_1]}(X) \cap M_{[-t_0, \infty)}(X)$ . Then, by von Neumann's alternating projection theorem (see, e.g., [26, Theorem 9.20]), we have  $Q = \text{s-lim}_{n \rightarrow \infty} P_n P_{n-1} \cdots P_1$ . Using this, (4.3) and Proposition 4.4, we immediately obtain the next proposition (cf. the proof of [3, Proposition 4.9 (3)]).

**Proposition 4.5.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . Then  $\lim_{n \rightarrow \infty} \int_0^\infty D_n(s, f)^2 ds = 0$ .*

We need the next proposition.

**Proposition 4.6.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . Then, for  $t > 0$  and  $n = 0, 1, \dots$ , we have*

$$\int_0^\infty b_{n+1}(t, \tau) f(t_1 + \tau) d\tau = \int_0^\infty a(t + u) D_n(u, f) du.$$

*Proof.* We may assume  $f \geq 0$ . By the Fubini–Tonelli theorem, we have, for  $t > 0$ ,

$$\begin{aligned} \int_0^\infty b_1(t, \tau) f(t_1 + \tau) d\tau &= \int_0^\infty \left\{ \int_0^\tau c(\tau - u) a(t + u) du \right\} f(t_1 + \tau) d\tau \\ &= \int_0^\infty a(t + u) \left\{ \int_0^\infty c(\tau) f(t_1 + u + \tau) d\tau \right\} du = \int_0^\infty a(t + u) D_0(u, f) du. \end{aligned}$$

Thus the assertion holds for  $n = 0$ . Now we assume that  $n \geq 1$ . Since we have

$$b_{n+1}(t, \tau) = \int_0^\infty a(t + v) \left\{ \int_0^\infty c(u) b_n(t_2 + u + v, \tau) du \right\} dv, \quad t, \tau > 0,$$

we obtain the assertion, again using the Fubini–Tonelli theorem.  $\square$

For  $t, s > 0$ , we define  $k(t, s) = k(t, s; t_2)$  by

$$k(t, s) := \int_0^\infty c(t + u) a(t_2 + u + s) du.$$

Notice that  $k(t, s) < \infty$  for  $t, s > 0$  since  $k(t, s) \leq c(t) \int_{t_2+s}^\infty a(u) du$ .

**Proposition 4.7.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . Then*

$$P_{n+1}\epsilon_n(f) = \begin{cases} \int_{-\infty}^{t_1} \left\{ \int_0^\infty k(t_1 - s, u) D_{n-1}(u, f) du \right\} dW(s), & n = 2, 4, \dots, \\ \int_{-t_0}^\infty \left\{ \int_0^\infty k(t_0 + s, u) D_{n-1}(u, f) du \right\} dW^*(s), & n = 1, 3, \dots \end{cases}$$

*Proof.* We assume  $n = 2, 4, \dots$ . Then, by Propositions 3.5 and 4.6, we have

$$\begin{aligned} P_{n+1}\epsilon_n(f) &= \int_{-\infty}^{t_1} \left\{ \int_{t_1}^\infty duc(u - s) \int_0^\infty b_n(t_0 + u, \tau) f(t_1 + \tau) d\tau \right\} dW(s) \\ &= \int_{-\infty}^{t_1} \left\{ \int_0^\infty dvc(t_1 - s + v) \int_0^\infty a(t_2 + v + u) D_{n-1}(u, f) du \right\} dW(s) \\ &= \int_{-\infty}^{t_1} \left\{ \int_0^\infty k(t_1 - s, u) D_{n-1}(u, f) du \right\} dW(s). \end{aligned}$$

The proof of the case  $n = 1, 3, \dots$  is similar.  $\square$

We need the next  $L^2$ -boundedness theorem.

**Theorem 4.8.** *Let  $p \in (0, 1/2)$  and let  $\ell(\cdot)$  be a slowly varying function at infinity. Let  $C(\cdot)$  and  $A(\cdot)$  be nonnegative and decreasing functions on  $(0, \infty)$ . We assume  $C(\cdot) \in L^1_{\text{loc}}[0, \infty)$  and  $A(0+) < \infty$ . We also assume*

$$\begin{aligned} A(t) &\sim t^{-(1+p)} \ell(t)^p, \quad t \rightarrow \infty, \\ C(t) &\sim \frac{t^{-(1-p)}}{\ell(t)} \cdot \frac{\sin(p\pi)}{\pi}, \quad t \rightarrow \infty. \end{aligned}$$

Then

$$\begin{aligned} \sup_{0 < x < \infty} \int_0^\infty K(x, y) (x/y)^{1/2} dy &< \infty, \\ \sup_{0 < y < \infty} \int_0^\infty K(x, y) (y/x)^{1/2} dx &< \infty, \end{aligned}$$

where  $K(x, y) := \int_0^\infty C(x + u) A(u + y) du$  for  $x, y > 0$ . In particular, the integral operator  $K$  defined by  $(Kf)(x) := \int_0^\infty K(x, y) f(y) dy$  for  $x > 0$  is a bounded operator on  $L^2((0, \infty), dy)$ .

The proof of Theorem 4.8 is similar to that of [15, Theorem 5.1], whence we omit it.

By putting  $z = iy$  in (3.2), we get

$$y \left\{ \int_0^\infty e^{-yt} c(t) dt \right\} \left\{ \int_0^\infty e^{-yt} \alpha(t) dt \right\} = 1, \quad y > 0.$$

By Karamata's Tauberian theorem (cf. [6, Theorem 1.7.6]) applied to this, (2.3) implies

$$\alpha(t) \sim \frac{t^{-(H-\frac{1}{2})}}{\ell(t)} \cdot \frac{1}{\Gamma(\frac{3}{2}-H)}, \quad t \rightarrow \infty.$$

This and the monotone density theorem give

$$(4.5) \quad a(t) \sim \frac{t^{-(H+\frac{1}{2})}}{\ell(t)} \cdot \frac{(H-\frac{1}{2})}{\Gamma(\frac{3}{2}-H)}, \quad t \rightarrow \infty.$$

The next lemma is a key to our arguments.

**Lemma 4.9.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . Then  $\|\epsilon_n(f)\| \rightarrow 0$  as  $n \rightarrow \infty$ .*

*Proof.* It follows from (2.3), (4.5) and Theorem 4.8 below that the integral operator  $K$  defined by  $Kf(t) := \int_0^\infty k(t, s)f(s)ds$  is a bounded operator on  $L^2((0, \infty), ds)$ . Hence, by Propositions 4.3, 4.5 and 4.7, we have

$$\begin{aligned} \|\epsilon_n(f)\|^2 &= \int_0^\infty D_n(s, f)^2 ds + \int_0^\infty \left\{ \int_0^\infty k(s, u) D_{n-1}(u, f) du \right\}^2 ds \\ &\leq \int_0^\infty D_n(s, f)^2 ds + \|K\|^2 \int_0^\infty D_{n-1}(s, f)^2 ds \rightarrow 0, \quad n \rightarrow \infty. \end{aligned}$$

Thus the lemma follows.  $\square$

We can now state the conclusions of the arguments above.

**Theorem 4.10.** *The following assertions hold:*

- (a)  $M_{[-t_0, t_1]}(X) = M_{(-\infty, t_1]}(X) \cap M_{[-t_0, \infty)}(X)$ .
- (b)  $P_{[-t_0, t_1]} = \text{s-lim}_{n \rightarrow \infty} P_n P_{n-1} \cdots P_1$ .
- (c)  $\|P_{[-t_0, t_1]}^\perp Z\|^2 = \|P_1^\perp Z\|^2 + \sum_{n=1}^\infty \|(P_{n+1})^\perp P_n \cdots P_1 Z\|^2$  for  $Z \in M(X)$ .

We can prove Theorem 4.10 using Proposition 4.2 and Lemma 4.9. Since the proof is similar to that of [3, Theorem 4.6], we omit the details.

**4.2. Finite past prediction formulas.** We define  $h(s, u) = h(s, u; t_2)$  by

$$(4.6) \quad h(s, u) := \sum_{k=1}^\infty \{b_{2k-1}(t_2 - s, u) + b_{2k}(s, u)\}, \quad 0 < s < t_2, \quad u > 0.$$

Here is the finite past prediction formula for  $\int_{t_1}^\infty f(s)dX(s)$ .

**Theorem 4.11.** *Let  $f \in \mathcal{H}_{[t_1, \infty)}(X)$ . Then the following assertions hold:*

- (a)  $\int_0^\infty h(t_0 + \cdot, u)f(t_1 + u)du \in \mathcal{H}_{[-t_0, t_1]}(X)$ .
- (b)  $P_{[-t_0, t_1]} \int_{t_1}^\infty f(s)dX(s) = \int_{-t_0}^{t_1} \left\{ \int_0^\infty h(t_0 + s, u)f(t_1 + u)du \right\} dX(s)$ .
- (c)  $\|P_{[-t_0, t_1]}^\perp \int_{t_1}^\infty f(s)dX(s)\|^2 = \sum_{n=0}^\infty \int_0^\infty D_n(s, f)^2 ds$ .

*Proof.* We may assume that  $f \geq 0$ . By Theorem 4.10 (b), Proposition 4.2 and Lemma 4.9, we have, in  $M(X)$ ,

$$\begin{aligned} P_{[-t_0, t_1]} \int_{t_1}^{\infty} f(s) dX(s) &= \lim_{n \rightarrow \infty} P_n P_{n-1} \cdots P_1 \int_{t_1}^{\infty} f(s) dX(s) \\ &= \lim_{n \rightarrow \infty} \int_{-t_0}^{t_1} \left\{ \int_0^{\infty} h_n(t_0 + u, v) f(t_1 + v) dv \right\} dX(s), \end{aligned}$$

where, for  $0 < s < t_2$  and  $u > 0$ , we define  $h_n(s, u) = h_n(s, u; t_2)$  by

$$h_n(s, u) = \begin{cases} b_1(t_2 - s, u) + b_2(s, u) + \cdots + b_n(t_2 - s, u), & n = 1, 3, 5, \dots, \\ b_1(t_2 - s, u) + b_2(s, u) + \cdots + b_n(s, u), & n = 2, 4, 6, \dots \end{cases}$$

Since  $h_n(s, u) \uparrow h(s, u)$  as  $n \rightarrow \infty$ , we obtain (a) and (b) using the monotone convergence theorem. Finally, (c) follows immediately from Theorem 4.11 (c) and Proposition 4.4.  $\square$

For  $s, u > 0$ , we define  $D_n(s) = D_n(s; t_2, t_3)$  by

$$D_n(s) := \int_0^{\infty} duc(u) \int_0^{t_3} b_n(t_2 + u + s, \tau) d\tau, \quad n = 1, 2, \dots$$

Here are the solutions to the finite past prediction problems for  $(X(t))$ .

**Theorem 4.12.** *The finite past prediction formula (1.12) and the following equality for the mean-square prediction error hold:*

$$\left\| P_{[-t_0, t_1]}^{\perp} X(T) \right\|^2 = \int_0^{T-t_1} g(s)^2 ds + \sum_{n=1}^{\infty} \int_0^{\infty} D_n(s)^2 ds.$$

*Proof.* We put  $f(s) = I_{(t_1, T]}(s)$ . Then  $\int_{t_1}^{\infty} f(s) dX(s) = X(T) - X(t_1)$  and

$$\int_0^{\infty} h(t_0 + s, u) f(t_1 + u) du = \int_0^{t_3} h(t_0 + s, u) du, \quad -t_0 < s < t_1.$$

We also have  $D_n(s, f) = D_n(s)$  for  $n = 1, 2, \dots$  and  $D_0(s, f) = g(t_3 - s)$ . Thus the theorem follows from Theorem 4.11.  $\square$

## 5. BAXTER'S INEQUALITY

In this section, we assume (1.1)–(1.7) and (1.10). Let  $t_2 := t_0 + t_1$  as before. By (4.6), the infinite and finite past predictor coefficients  $b(t, s)$  and  $h(s, u) = h(s, u; t_2)$  satisfy, for  $s \in (-t_0, t_1)$  and  $u > 0$ ,

$$(5.1) \quad h(s + t_0, u) - b(t_1 - s, u) = \sum_{k=1}^{\infty} \{b_{2k}(s + t_0, u) + b_{2k+1}(t_1 - s, u)\} > 0,$$

where we recall that  $b_n(t, s) = b_n(t, s; t_2)$  from (4.1).

The aim here is to prove Baxter's inequality for  $(X(t))$ .

**Theorem 5.1.** *There exists a positive constant  $K$  such that, for all  $t_0 \geq 1$ ,*

$$\int_{-t_0}^{t_1} ds \int_0^{T-t_1} \{h(s + t_0, u) - b(t_1 - s, u)\} du \leq K \int_{-\infty}^{-t_0} ds \int_0^{T-t_1} b(t_1 - s, u) du.$$

5.1. **Representation in terms of  $\beta$ .** We define a positive function  $\beta(t)$  by

$$\beta(t) := \int_0^\infty c(v)a(t+v)dv, \quad t > 0.$$

We next derive the representation of the finite past prediction coefficient  $h(s, u) = h(s, u; t_2)$  in terms of  $\beta(t)$  (and  $c(t)$  and  $a(t)$ ). We need this result in Section 5.2. See [16, 17, 14] for the usefulness of such expressions in terms of  $\beta(t)$  in the discrete-time setting.

For  $t, u, v > 0$ , we define  $\delta_1(u, v; t) := \beta(t + v + u)$ ,

$$\delta_2(u, v; t) := \int_0^\infty dw_1 \beta(t + v + w_1) \beta(t + w_1 + u),$$

and, for  $k = 3, 4, \dots$ ,

$$\begin{aligned} \delta_k(u, v; t) := & \int_0^\infty dw_{k-1} \cdots \int_0^\infty dw_1 \beta(t + v + w_{k-1}) \\ & \times \left\{ \prod_{l=1}^{k-2} \beta(t + w_{l+1} + w_l) \right\} \beta(t + w_1 + u). \end{aligned}$$

For  $t, s > 0$ , we define  $B_1(t, s; t_2) := b(t, s)$ , and, for  $k \geq 2$ ,

$$B_k(t, s; t_2) := \int_0^s dvc(s-v) \int_0^\infty a(t+u) \delta_{k-1}(u, v; t_2) du$$

The next proposition gives the desired representation of  $h(s, u)$ .

**Proposition 5.2.** *For  $t, s > 0$  and  $k \geq 1$ ,  $b_k(t, s; t_2) = B_k(t, s; t_2)$ , that is,*

$$b_k(t, s; t_2) = \int_0^s dvc(s-v) \int_0^\infty a(t+u) \delta_{k-1}(u, v; t_2) du, \quad k = 2, 3, \dots$$

*Proof.* It is enough to show that, for  $t, s > 0$  and  $k = 1, 2, \dots$ ,

$$B_{k+1}(t, s; t_2) = \int_0^\infty b(t, \tau) B_k(t_2 + \tau, s; t_2) d\tau.$$

However, from the Fubini–Tonelli theorem, we see that

$$\begin{aligned} & \int_0^\infty b(t, \tau) B_k(t_2 + \tau, s; t_2) d\tau \\ &= \int_0^\infty \left\{ \int_0^\tau c(\tau-z)a(t+z)dz \right\} \\ & \quad \times \left\{ \int_0^s dvc(s-v) \int_0^\infty a(t_2 + \tau + u) \delta_{k-1}(u, v; t_2) du \right\} d\tau \\ &= \int_0^s dvc(s-v) \int_0^\infty dza(t+z) \\ & \quad \int_0^\infty du \left\{ \int_z^\infty dvc(\tau-z)a(t_2 + \tau + u) \right\} \delta_{k-1}(u, v; t_2) \\ &= \int_0^s dvc(s-v) \int_0^\infty dza(t+z) \int_0^\infty \beta(t_2 + z_u) \delta_{k-1}(u, v; t_2) du \\ &= \int_0^s dvc(s-v) \int_0^\infty dza(t+z) \delta_k(z, v; t_2) \\ &= B_{k+1}(t, s; t_2). \end{aligned}$$

Thus the proposition follows.  $\square$

**5.2. Proof of Baxter's inequality.** For simplicity, we write  $d := H - \frac{1}{2}$ . Then  $0 < d < 1/2$ . From (2.3), (4.5) and [11, Proposition 4.3], we have

$$(5.2) \quad \beta(t) \sim t^{-1} \cdot \frac{\sin(\pi d)}{\pi}, \quad t \rightarrow \infty.$$

As in [12, Section 6], [13, Section 3] and [17, Section 3]], we put, for  $u \geq 0$ ,

$$f_1(u) := \frac{1}{\pi(1+u)}, \quad f_2(u) := \frac{1}{\pi^2} \int_0^\infty \frac{ds_1}{(s_1+1)(s_1+1+u)},$$

and, for  $k = 3, 4, \dots$ ,

$$f_k(u) := \frac{1}{\pi^k} \int_0^\infty ds_{k-1} \cdots \int_0^\infty ds_1 \frac{1}{(1+s_{k-1})} \times \left\{ \prod_{l=1}^{k-2} \frac{1}{(1+s_{l+1}+s_l)} \right\} \\ \times \frac{1}{(1+s_1+u)}.$$

**Proposition 5.3.** *The following assertions hold:*

(a) *For  $r \in (1, \infty)$ , there exists  $N > 0$  such that*

$$0 < \delta_k(u, v; t) \leq \frac{f_k(0) \{r \sin(\pi d)\}^k}{t}, \quad u, v > 0, \quad k \in \mathbf{N}, \quad t \geq N.$$

(b) *For  $k \in \mathbf{N}$  and  $u > 0$ ,  $\delta_k(tu, v; t) \sim t^{-1} f_k(u) \sin^k(\pi d)$  as  $t \rightarrow \infty$ .*

For example, we see from (5.2) that, formally,

$$t\delta_k(tu, v; t) = \int_0^\infty ds_{k-1} \cdots \int_0^\infty ds_1 t\beta(t(1+(v/t)+s_{k-1})) \\ \times \left\{ \prod_{l=1}^{k-2} t\beta(t(1+s_{l+1}+s_l)) \right\} t\beta(t(1+s_1+u)) \\ \rightarrow f_k(u) \sin^k(\pi d), \quad t \rightarrow \infty,$$

which is (b) of the proposition above. Since we can prove the two assertions rigorously as in the proof of [17, Proposition 3.2], we omit the details.

Theorem 5.1 follows immediately from the next more precise result.

**Lemma 5.4.** *For  $0 \leq t_1 < T$ , we have, as  $t_0 \rightarrow \infty$ ,*

$$\int_{-t_0}^{t_1} ds \int_0^{T-t_1} \{h(s+t_0, u; t_2) - b(t_1-s, u)\} du \\ \sim t_2 a(t_2) \cdot \left\{ \int_0^{T-t_1} ds \int_0^s c(v) dv \right\} \cdot \left\{ \int_0^1 s^{-d-1} [(1-s)^{-d} - 1] ds \right\} \\ \sim \int_{-\infty}^{-t_0} ds \int_0^{T-t_1} b(t_1-s, u) du \cdot d \int_0^1 s^{-d-1} [(1-s)^{-d} - 1] ds.$$

*Proof.* By (5.1) and Proposition 5.2, we have

$$\begin{aligned}
& \int_{-t_0}^{t_1} ds \int_0^{T-t_1} \{h(s+t_0, u; t_2) - b(t_1-s, u)\} du \\
&= \sum_{k=2}^{\infty} \int_0^{T-t_1} ds \int_0^{t_2} d\tau b_k(\tau, s; t_2) \\
&= \sum_{k=2}^{\infty} \int_{-t_0}^{t_1} ds \int_0^s dvc(s-v) \int_0^{t_2} d\tau \int_0^{\infty} a(\tau+u) \delta_{k-1}(u, v; t_2) du \\
&= (t_2)^2 \sum_{k=2}^{\infty} \int_{-t_0}^{t_1} ds \int_0^s dvc(s-v) \int_0^{t_2} d\tau \int_0^{\infty} a(t_2(\tau+u)) \delta_{k-1}(t_2u, v; t_2) du.
\end{aligned}$$

Therefore, by (4.5), Proposition 5.3 and standard arguments involving Potter's theorem (cf. [6, Theorem 1.5.6]) and the dominated convergence theorem,

$$\begin{aligned}
& \frac{1}{t_2 a(t_2)} \int_{-t_0}^{t_1} ds \int_0^{T-t_1} \{h(s+t_0, u; t_2) - b(t_1-s, u)\} du \\
&= \sum_{k=2}^{\infty} \int_{-t_0}^{t_1} ds \int_0^s dvc(s-v) \int_0^{t_2} d\tau \int_0^{\infty} \frac{a(t_2(\tau+u))}{a(t_2)} t_2 \delta_{k-1}(t_2u, v; t_2) du \\
&\rightarrow \left\{ \int_0^{T-t_1} ds \int_0^s c(v) dv \right\} \sum_{m=1}^{\infty} \left\{ \int_0^{\infty} du f_m(u) \int_0^1 \frac{1}{(\tau+u)^{d+1}} d\tau \right\} \sin^m(\pi d)
\end{aligned}$$

as  $t_0 \rightarrow \infty$ . On the other hand,

$$\begin{aligned}
& \frac{1}{t_2 a(t_2)} \int_{-\infty}^{-t_0} ds \int_0^{T-t_1} b(t_1-s, u) du \\
&= \int_0^{T-t_1} du \int_0^u dvc(u-v) \int_0^{\infty} \frac{a(t_2(1+s+(v/t_2)))}{a(t_2)} ds \\
&\rightarrow \int_0^{T-t_1} du \int_0^u duc(u-v) \int_0^{\infty} \frac{1}{(1+s)^{d+1}} ds = \frac{1}{d} \int_0^{T-t_1} du \int_0^u dvc(u-v)
\end{aligned}$$

as  $t_0 \rightarrow \infty$ . Therefore, by Lemma 5.5 below, we obtain the lemma.  $\square$

**Lemma 5.5.** *For  $0 < d < 1/2$ , it holds that*

$$\sum_{m=1}^{\infty} \left\{ \int_0^{\infty} du f_m(u) \int_0^1 \frac{1}{(\tau+u)^{d+1}} d\tau \right\} \sin^m(\pi d) = \int_0^1 s^{-d-1} [(1-s)^{-d} - 1] ds.$$

*Proof.* Though the lemma is a general result, we give a proof based on the results for fBm. Thus we take fBm  $(B_H(t))$  as  $(X(t))$ . Then we have (3.11) and (3.12). Also, by [9], we have (3.13) and

$$(5.3) \quad h(s, u; t_2) = \frac{\sin \left\{ \pi \left( H - \frac{1}{2} \right) \right\}}{\pi} (t_2 - s)^{-(H-\frac{1}{2})} s^{-(H-\frac{1}{2})} \frac{\{u(u+t_2)\}^{H-\frac{1}{2}}}{u+t_2-s},$$



whence, as  $t_0 \rightarrow \infty$ ,

$$\begin{aligned} & \frac{1}{t_2 a(t_2)} \int_{-t_0}^{t_1} ds \int_0^{T-t_1} \{h(s+t_0, u; t_2) - b(t_1-s, u)\} du \\ &= \frac{1}{t_2 a(t_2)} \int_0^{t_2} ds \int_0^{T-t_1} \{h(t_2-s, u; t_2) - b(s, u)\} du \\ &= \frac{1}{\Gamma(d+1)} \int_0^{T-t_1} du u^d \int_0^1 ds \frac{s^{-d} t_2}{u+t_2 s} \left\{ \left( \frac{u}{t_2} + 1 \right)^d (1-s)^{-d} - 1 \right\} \\ &\quad \rightarrow \frac{(T-t_1)^{d+1}}{\Gamma(d+2)} \int_0^1 s^{-d-1} [(1-s)^{-d} - 1] ds. \end{aligned}$$

However, by the proof of Theorem 5.1, this limit must be equal to

$$\left\{ \int_0^{T-t_1} ds \int_0^s c(v) dv \right\} \sum_{m=1}^{\infty} \left\{ \int_0^{\infty} du f_m(u) \int_0^1 \frac{1}{(\tau+u)^{d+1}} d\tau \right\} \sin^m(\pi d).$$

Thus the lemma follows.  $\square$

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