

Prediction of fractional processes with long-range dependence

Akihiko INOUE and Vo V. ANH

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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.

Key words: Predictor coefficients, prediction, fractional Brownian motion, long-range dependence.

1. Introduction

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

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

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

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

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

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

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

We will also assume some extra conditions such as

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

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

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

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

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 $c(t)$ (as well as $g(t)$) the *MA(∞) coefficient* of $(X(t))$. We remark that, in the prediction formulas for $(X(t))$ which we consider in this paper, $c(t)$ becomes more relevant than $g(t)$.

A typical example of ν is

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

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

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

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 [K]. For $1/2 < H < 1$, fBm has both *self-similarity* and *long-range dependence* (Samorodnitsky and Taqqu [ST]), and plays an important role in various fields such as network traffic (see, e.g., Mikosch et al. [MRRS]) and finance (see, e.g., Hu et al. [HOS]); see also Taqqu [T] 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 [DU], and Nualart [N]. Grecksch and Anh [GA] introduced Hilbert space-valued fBm and the corresponding stochastic calculus. Duncan et al. [DMP] and Tindel et al. [TTV] studied stochastic evolution equations with fBm in Hilbert spaces. 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

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

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. [NVV], Anh et al. [AIK], Inoue et al. [INA] and Inoue and Nakano [IN]). 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):

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

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

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(∞) coefficient $c(t)$ and AR(∞) coefficient $a(t)$, to be defined in Section 3.1, of $(X(t))$. The integral of $a(t)$ is in fact the coefficient of an AR(∞)-type equation describing $(X(t))$ (see Section 5). 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(∞) coefficient, in addition to the nice MA(∞) coefficient, is a key to the solution to the prediction problems above.

For fBm with $1/2 < H < 1$, the predictor coefficients $b(t, s)$ and $h(t, s)$ are given in Gripenberg and Norros [GN]. See [NVV] and [NP] for different proofs. Fractional Brownian motion has a variety of nice properties, and the methods of proof of [GN], [NVV], [NP] 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 [Y] and Nuzman and Poor [NP], respectively (see also Anh and Inoue [AI1]).

In Inoue and Anh [IA], a class of processes $(\tilde{X}(t))$ of the same form

$$\tilde{X}(t) = \int_{-\infty}^{\infty} \{\tilde{c}(t-s) - \tilde{c}(-s)\} dW(s), \quad t \in \mathbf{R}, \quad (1.13)$$

as (1.1) are introduced. Unlike $g(t)$ in (1.1), however, the kernel $\tilde{c}(t)$ itself is assumed to be of the form

$$\tilde{c}(t) = I_{(0,\infty)}(t) \int_0^{\infty} e^{-ts} \tilde{\nu}(ds), \quad t \in \mathbf{R}, \quad (1.14)$$

with a Borel measure $\tilde{\nu}$ on $(0, \infty)$ satisfying some suitable conditions. This class of $(\tilde{X}(t))$ includes fBm with $H \in (0, 1/2)$ as a typical example. Notice that $\tilde{c}(t)$ in (1.14) (resp., $g(t)$ in (1.1)) is decreasing (resp., increasing) on $(0, \infty)$ as $t^{H-(1/2)}$ with $H \in (0, 1/2)$ (resp., $(1/2, 1)$) is. In [IA], prediction formulas for $(\tilde{X}(t))$ are proved, extending the results for fBm with $H \in (0, 1/2)$ stated above. These prediction formulas for $(\tilde{X}(t))$, including those for fBm with $H \in (0, 1/2)$, have different forms from (1.11) and (1.12), in that 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. Finally in Section 5, we remark on the AR(∞)-type equations describing $(X(t))$ and $(\tilde{X}(t))$.

2. Basic properties and examples

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

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

Then, as in [IA, 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*

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

Then

$$\sigma(t) \sim t^{H_0} \ell_0(1/t) \sqrt{v(H_0)}, \quad t \rightarrow 0+,$$

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

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

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

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

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

Similarly, (2.2) implies

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

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, we omit the details.

Example 2.3 For $H \in (1/2, 1)$, let ν 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 $(B_H(t))$:

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

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 [MV].

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. [BGT, 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

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

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

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 [IA, Theorem 3.2], there exists a unique Borel measure μ 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

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

with

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

We define

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

We call $a(t)$ (as well as $\alpha(t)$) the *AR(∞) coefficient* of $(X(t))$ (see Section 5 for background). 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 [IA, Lemma 3.4], the following equalities hold:

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

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

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 notice that, by Lemma 5.2 below, the function $c(t)$, whence $\mathcal{H}_I(X)$, is uniquely determined by $(X(t))$. We define a subclass \mathcal{H}_I^0 of $\mathcal{H}_I(X)$ by

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

Each member of $f \in \mathcal{H}_I^0$ is 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

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

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. □

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 $f \in H_I(X)$, 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. [I2, Section 2] and [AIK, 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 [IA, 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 $t \in \mathbf{R}$.*

See Section 5 for the proof.

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

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

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

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

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

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

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

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

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 [AIK, 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(∞) coefficient $c(t)$ is given by

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

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

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

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

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

where we have used the equality

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

Thus

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

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

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

This prediction formula was obtained in [GN, 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

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

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 $s > -t_0$,

$$\begin{aligned} & \int_0^\infty du b(t_0 + s, u) \int_0^\infty b_1(t_2 + u, \tau)f(t_1 + \tau)d\tau \\ &= \int_0^\infty b_2(t_0 + s, \tau)f(t_1 + \tau)d\tau. \end{aligned}$$

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

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*

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

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

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

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

Since the proof is similar to that of [AIK, 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

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

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 [AIK, 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 [AIK, Proposition 4.9]).

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

(a) $\|P_1^\perp \int_{t_1}^\infty f(s) dX(s)\|^2 = \int_0^\infty D_0(s, f)^2 ds.$

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

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., [P, Theorem 9.20]), we have $Q = s\text{-}\lim_{n \rightarrow \infty} P_n P_{n-1} \cdots P_1$. Using this, (4.3) and Proposition 4.4, we immediately obtain the next proposition (cf. the proof of [AIK, 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

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

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

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 $C(\cdot) \in L^1_{loc}[0, \infty)$ and $A(0+) < \infty$. We also assume*

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

and put $K(x, y) := \int_0^\infty C(x + u)A(u + y)du$ for $x, y > 0$. Then

$$\sup_{x>0} \int_0^\infty K(x, y)(x/y)^{1/2} dy < \infty, \quad \sup_{y>0} \int_0^\infty K(x, y)(y/x)^{1/2} dx < \infty.$$

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 $L^2((0, \infty), dy)$.

We omit the proof of Theorem 4.8 which is similar to that of [IA, Theorem 5.1].

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. [BGT, Theorem 1.7.6]) applied to this, (2.3) implies $\alpha(t) \sim t^{-(H-\frac{1}{2})} / \{\ell(t)\Gamma((3/2) - H)\}$ as $t \rightarrow \infty$. This and the monotone density theorem give

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

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. □

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) $P_{[-t_0, t_1]} = s\text{-}\lim_{n \rightarrow \infty} P_n P_{n-1} \cdots P_1$.

(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 [AIK, Theorem 4.6], we omit the details.

4.2. Finite past prediction formulas

We define $h(s, u) = h(s, u; t_2)$ by

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

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) $P_{[-t_0, t_1]} \int_{t_1}^\infty f(s)dX(s) = \int_{-t_0}^{t_1} \{ \int_0^\infty h(t_0 + s, u)f(t_1 + u)du \} dX(s)$.
- (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 $h_n(s, u) \uparrow h(s, u)$, we obtain (a) and (b) using the monotone convergence theorem. Finally, (c) follows from Theorem 4.10 (c) and Proposition 4.4. □

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:*

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

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. AR(∞)-type equations

In this section, we consider the AR(∞)-type equations for $(X(t))$ in (1.1) and $(\tilde{X}(t))$ in (1.13). For a Borel measure τ on $(0, \infty)$ satisfying $\int_0^{\infty} (1+s)^{-1} \tau(ds) < \infty$, we write

$$F_\tau(z) := \int_0^{\infty} \frac{1}{\lambda - iz} \tau(d\lambda), \quad \Im z \geq 0.$$

First, we consider the process $X = (X(t))$ in (1.1) with (1.2)–(1.5), (2.1) and (3.1). Let $f_t(s) := g(t-s) - g(-s) = \int_{-s}^{t-s} c(u) du$ for $t, s \in \mathbf{R}$.

Lemma 5.1 *Let $t \in \mathbf{R}$. Then the Fourier transform of $f_t(\cdot)$ in the L^2 -sense is equal to $(i\xi)^{-1}(1 - e^{-it\xi})F_\nu(\xi)$:*

$$\frac{(1 - e^{-it\xi})}{i\xi} F_\nu(\xi) = \text{l.i.m.}_{M \rightarrow \infty} \int_{-M}^M e^{-is\xi} f_t(s) ds. \quad (5.1)$$

Proof. Since $\int_{-\infty}^{\infty} |f_t(s)|^2 ds < \infty$, the limit on the right-hand side of (5.1) exists. Therefore, it is enough to justify the following point-wise convergence:

$$\frac{(1 - e^{-it\xi})}{i\xi} F_\nu(\xi) = \lim_{M \rightarrow \infty} \int_{-M}^M e^{-is\xi} f_t(s) ds, \quad \xi \neq 0. \quad (5.2)$$

Now, if $-M \leq t \leq M$, then

$$\begin{aligned} & \int_{-M}^M e^{-is\xi} f_t(s) ds \\ &= \int_{-M}^M ds e^{-is\xi} \int_0^t c(u-s) du = \int_0^t du \int_{-M}^M e^{-is\xi} c(u-s) ds \\ &= \int_0^t du e^{-iu\xi} \int_{u-M}^{u+M} e^{iv\xi} c(v) dv = \int_0^t du e^{-iu\xi} \int_0^{u+M} e^{iv\xi} c(v) dv \end{aligned}$$

because $u - M \leq 0 \leq u + M$ for u between 0 and t , and $c(s) = 0$ for $s \leq 0$. However,

$$\begin{aligned} & \int_0^t du e^{-iu\xi} \int_0^{u+M} e^{is\xi} c(s) ds \\ &= \int_0^t du e^{-iu\xi} \int_0^\infty \frac{1 - e^{(i\xi-\lambda)(u+M)}}{\lambda - i\xi} \nu(d\lambda) \\ &= \frac{(1 - e^{-it\xi})}{i\xi} F_\nu(\xi) - e^{i\xi M} \int_0^t du \int_0^\infty \frac{e^{-\lambda(u+M)}}{\lambda - i\xi} \nu(d\lambda), \end{aligned}$$

so that, for $\xi \neq 0$,

$$\left| \frac{(1 - e^{-it\xi})}{i\xi} F_\nu(\xi) - \int_{-M}^M e^{-is\xi} f_t(s) ds \right| \leq t \int_0^\infty \frac{e^{-\lambda M}}{|\lambda - i\xi|} \nu(d\lambda) \downarrow 0, \quad M \rightarrow \infty.$$

Thus, (5.2) holds. □

For the Brownian motion $W = (W(t))$ in (1.1), let $DW(\phi) = \int_{-\infty}^\infty \hat{\phi}(\xi) Z_{DW}(d\xi)$ with $\phi \in \mathcal{D}(\mathbf{R})$ be the spectral decomposition of DW as a stationary random distribution, where $\hat{\phi}(\xi) := \int_{-\infty}^\infty e^{-it\xi} \phi(t) dt$ and Z_{DW} is the associated complex-valued random measure such that $E[Z_{DW}(A) \overline{Z_{DW}(B)}] = (2\pi)^{-1} \int_{A \cap B} d\xi$ (see Itô [It]). By Lemma 5.1 and the Parseval-type formula for the homogeneous random measure Z_{DW} , we obtain $X(t) = \int_{-\infty}^\infty [(1 - e^{-it\xi})/(i\xi)] F_\nu(\xi) Z_{DW}(d\xi)$, whence

$$DX(\phi) = \int_{-\infty}^{\infty} \hat{\phi}(\xi) F_{\nu}(\xi) Z_{DW}(d\xi), \quad \phi \in \mathcal{D}(\mathbf{R}). \tag{5.3}$$

Let ρ_{DX} be the spectral measure of DX : $E[X(\phi)\overline{X(\psi)}] = \int_{-\infty}^{\infty} \hat{\phi}(\xi)\overline{\hat{\psi}(\xi)} \rho_{DX}(d\xi)$. Then, from (5.3), we see that $\rho_{DX}(d\xi) = (2\pi)^{-1}|F_{\nu}(\xi)|^2 d\xi$. Thus, DX has the spectral density $\Delta_{DX}(\xi) := (2\pi)^{-1}|F_{\nu}(\xi)|^2$. Since, for $z = x+iy$ with $y > 0$, we have

$$\Re\{F_{\nu}(z)\} = \int_0^{\infty} \frac{s+y}{(s+y)^2+x^2} \nu(ds) > 0,$$

the function $F_{\nu}(z)$ is an outer function on the upper half plane $\Im z > 0$:

$$F_{\nu}(z) = \exp \left\{ \frac{1}{\pi i} \int_{-\infty}^{\infty} \frac{1+\xi z}{\xi-z} \cdot \frac{\log |F_{\nu}(\xi)|}{1+\xi^2} d\xi \right\}, \quad \Im z > 0. \tag{5.4}$$

In particular, Proposition 3.6 follows from this and (5.3).

We also have the next lemma.

Lemma 5.2 *The following equality holds:*

$$\int_0^{\infty} e^{izt} c(t) dt = \sqrt{2\pi} \exp \left\{ \frac{1}{2\pi i} \int_{-\infty}^{\infty} \frac{1+\xi z}{\xi-z} \cdot \frac{\log |\Delta_{DX}(\xi)|}{1+\xi^2} d\xi \right\}, \quad \Im z > 0.$$

Proof. Since $F_{\nu}(z) = \int_0^{\infty} e^{izt} c(t) dt$ and $|F_{\nu}(\xi)| = \{2\pi\Delta_{DX}(\xi)\}^{1/2}$, the lemma follows from (5.4). □

From Lemma 5.2, we see that the kernel $c(\cdot)$ is uniquely determined by DX , whence $(X(t))$, as claimed in Section 3.2.

Let $D^2X := D(DX)$. For the AR(∞) kernel $\alpha(\cdot)$ in Section 3.1, we define the convolution $\alpha * D^2X$, which is also a stationary random distribution, by

$$(\alpha * D^2X)(\phi) := \text{l.i.m.}_{M \rightarrow \infty} \int_0^M \alpha(u) D^2X(\tau_u \phi) du, \quad \phi \in \mathcal{D}(\mathbf{R}), \tag{5.5}$$

where $\tau_u \phi(t) := \phi(t+u)$ and the integral on the right-hand side is an $M(X)$ -valued Bochner integral. Then, by [12, Proposition 2.3] and (5.3), we have

$$(\alpha * D^2 X)(\phi) = - \int_{-\infty}^{\infty} i\xi F_{\mu}(\xi) F_{\nu}(\xi) \hat{\phi}(\xi) Z_{DW}(d\xi).$$

However, since (3.2) implies $-i\xi F_{\mu}(\xi) F_{\nu}(\xi) = 1$ for $\xi \neq 0$, we see that X satisfies

$$\alpha * D^2 X = DW. \tag{5.6}$$

More precisely, we have the next theorem.

Theorem 5.3 *The process $(X(t))$ is the only stationary-increment process with $X(0) = 0$ satisfying the following two conditions:*

- (1) *the stationary random distribution DX is purely nondeterministic;*
- (2) *$(X(t))$ satisfies (5.6).*

The proof of Theorem 5.3 is similar to that of [AI2, Theorem 2.6], whence we omit it. Notice that (5.6) can be written formally as the following AR(∞)-type equation:

$$\int_{-\infty}^t \alpha(t-s) \frac{d^2 X}{ds^2}(s) ds = \frac{dW}{dt}(t). \tag{5.7}$$

Example 5.4 Let $(B_H(t))$ be the fBm in (2.5) with $1/2 < H < 1$. Then, by Example 3.11, we have $\alpha(t) = t^{(1/2)-H} / \Gamma(\frac{3}{2} - H)$ for $t > 0$, whence (5.7) becomes

$$\frac{1}{\Gamma(\frac{3}{2} - H)} \int_{-\infty}^t \frac{1}{(t-s)^{H-(1/2)}} \cdot \frac{d^2 B_H}{ds^2}(s) ds = \frac{dW}{dt}(t).$$

Next, we turn to $\tilde{X} = (\tilde{X}(t))$ in (1.13) with (1.14). We assume that $\tilde{\nu}$ is a Borel measure on $(0, \infty)$ satisfying the following conditions:

$$\int_0^{\infty} \frac{1}{1+s} \tilde{\nu}(ds) < \infty, \quad \tilde{\nu}((0, \infty)) = \int_0^{\infty} \frac{1}{s} \tilde{\nu}(ds) = \infty, \quad \int_0^1 \tilde{c}(t)^2 dt < \infty.$$

By [IA, Theorem 3.2], there exists a unique Borel measure $\tilde{\mu}$ on $(0, \infty)$ satisfying

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

and $-izF_{\tilde{\nu}}(z)F_{\tilde{\mu}}(z) = 1$ for $\Im z > 0$. If we define

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

then the last equality becomes

$$-iz \left\{ \int_0^\infty e^{izt} \tilde{c}(t) dt \right\} \left\{ \int_0^\infty e^{izt} \tilde{\alpha}(t) dt \right\} = 1, \quad \Im z > 0. \quad (5.8)$$

By [IA, (2.3)], we have

$$D\tilde{X}(\phi) = \int_{-\infty}^\infty \hat{\phi}(\xi) (-i\xi) F_{\tilde{\nu}}(\xi) Z_{DW}(d\xi), \quad \phi \in \mathcal{D}(\mathbf{R}), \quad (5.9)$$

whence, in the same way as the proof of [I1, Proposition 5.1], we get

$$(\tilde{\alpha} * D\tilde{X})(\phi) = - \int_{-\infty}^\infty i\xi F_{\tilde{\mu}}(\xi) F_{\tilde{\nu}}(\xi) \hat{\phi}(\xi) Z_{DW}(d\xi), \quad \phi \in \mathcal{D}(\mathbf{R}),$$

where the convolution $\tilde{\alpha} * D\tilde{X}$ is defined in the same way as (5.5). However, since $-i\xi F_{\tilde{\mu}}(\xi) F_{\tilde{\nu}}(\xi) = 1$ for $\xi \neq 0$, we see that $(\tilde{X}(t))$ satisfies

$$\tilde{\alpha} * D\tilde{X} = DW. \quad (5.10)$$

Notice that the equation (5.10) can be written formally as the following AR(∞)-type equation:

$$\int_{-\infty}^t \tilde{\alpha}(t-s) \frac{d\tilde{X}}{ds}(s) ds = \frac{dW}{dt}(t). \quad (5.11)$$

We can also prove an analogue of Theorem 5.3 for $(\tilde{X}(t))$, which we omit in this paper.

Example 5.5 Let $(B_H(t))$ be the fBm in (2.5) with $0 < H < 1/2$. Then, by [IA, Example 3.9], we have $\tilde{\alpha}(t) = t^{-(1/2)-H}/\Gamma(\frac{1}{2}-H)$ for $t > 0$, whence (5.11) becomes

$$\frac{1}{\Gamma(\frac{1}{2}-H)} \int_{-\infty}^t \frac{1}{(t-s)^{H+(1/2)}} \cdot \frac{dB_H}{ds}(s) ds = \frac{dW}{dt}(t).$$

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Akihiko Inoue
Department of Mathematics
Hiroshima University
Higashi-Hiroshima 739-8526, Japan
E-mail: inoue100@hiroshima-u.ac.jp

Vo V. Anh
School of Mathematical Sciences
Queensland University of Technology
GPO Box 2434, Brisbane, Queensland 4001
Australia
E-mail: v.anh@qut.edu.au