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**Article:**
Issoglio, E and Riedle, M (2014) Cylindrical fractional Brownian motion in Banach spaces. Stochastic Processes and their Applications, 124 (11). pp. 3507-3534. ISSN 0304-4149

http://dx.doi.org/10.1016/j.spa.2014.05.010

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In this article we introduce cylindrical fractional Brownian motions in Banach spaces and develop the related stochastic integration theory. Here a cylindrical fractional Brownian motion is understood in the classical framework of cylindrical random variables and cylindrical measures. The developed stochastic integral for deterministic operator valued integrands is based on a series representation of the cylindrical fractional Brownian motion, which is analogous to the Karhunen-Loève expansion for genuine stochastic processes. In the last part we apply our results to study the abstract stochastic Cauchy problem in a Banach space driven by cylindrical fractional Brownian motion.

1. Introduction

In the past decades, a wide variety of infinite dimensional stochastic equations have been studied, due to their broad range of applications in physics, biology, neuroscience and in numerous other areas. A comprehensive study of stochastic evolution equations in Hilbert spaces driven by cylindrical Wiener processes, based on a semigroup approach, can be found in the monograph of Da Prato and Zabczyk [8]. Various extensions and modifications have been studied, such as different types of noises as well as generalisations to Banach spaces. For the latter see for example Brzeźniak [6] and van Neerven et al. [30, 31].

Fractional Brownian motion (fBm) has become very popular in recent years as driving noise in stochastic equations, in particular as an alternative to the classical Wiener noise. This is mainly due to properties of fBms, such as long-term dependence, which leads to a memory effect, and self-similarity, features which show great potential for applications, for example in hydrology, telecommunication traffic, queueing theory and mathematical finance. Since fBms are not semi-martingales, Itô-type calculus cannot be applied. Several different stochastic integrals with respect to real valued fBm have been introduced in the literature, e.g. Wiener integrals for deterministic integrands, Skorohod integrals using Malliavin calculus techniques, pathwise integrals using generalised Stieltjes integrals or integrals based on rough path theory. For more details see e.g. [5, 19, 20] and references therein.

2010 Mathematics Subject Classification. Primary 60G22; Secondary 60H05, 60H15, 28C20.

Key words and phrases. Cylindrical fractional Brownian motion, stochastic integration in Banach spaces, stochastic partial differential equations, fractional Ornstein-Uhlenbeck process, $\gamma$-radonifying, cylindrical measures.

The second named author acknowledges the EPSRC grant EP/I036990/1.
The purpose of this paper is to begin a systematic study of cylindrical fractional Brownian motion in Banach and Hilbert spaces and, starting from this, to build up a related stochastic calculus with respect to cylindrical fBm. Our approach is strongly based on the theory of cylindrical measures and cylindrical random variables, as it was exploited by Kallianpur and Xiong in [16] for the Wiener case and Metivier and Pellaumail in [18] for the cylindrical martingale case. In this paper we consider an extension beyond the martingale case, since fractional Brownian motion is not a semi-martingale. The cylindrical approach enables us to develop a theory that does not require a Hilbert space structure of the underlying space because the cylindrical fBm is defined through finite dimensional projections. We can characterise the cylindrical fBm by a series representation, which can be considered as the analogue of the Karhunen-Loève expansion in the classical situation of genuine stochastic processes. This representation is exploited to define the stochastic integral of deterministic, operator valued integrands with respect to a cylindrical fBm. The stochastic integral is defined as a stochastic version of a Pettis integral, as accomplished in van Neerven and Weis [31] for Wiener processes and in Riedle and van Gaans [25] for Lévy processes.

We apply our theory to a class of parabolic stochastic equations in Banach spaces of the form

\[ dY(t) = AY(t) \, dt + C \, dB(t), \] (1.1)

where \( B \) is a cylindrical fBm in a separable Banach space \( U \), \( A \) is a generator of a strongly continuous semigroup in a separable Banach space \( V \) and \( C \) is a linear and continuous operator from \( U \) to \( V \). We give necessary and sufficient conditions for the existence and uniqueness of a weak solution, which is a genuine stochastic process in the Banach space \( U \). For comparison, we apply our methods to an example often considered in the literature and typically formulated in a Hilbert space setting.

There are several works in the literature devoted to a similar or related problem, such as Brzeźniak et al. [7] for the case of a Banach space, and Grecksch and Anh [13], Duncan and coauthors in a series of papers [9, 10, 11, 12], Tindel et al. [28], Maslowski and Nualart [17], Gubinelli et al. [14] in the Hilbert space. Among these the papers [7] by Brzeźniak et al. and [28] by Tindel et al. are the most related to our work, and thus, it might be worth to comment on them in more detail in the sequel: in [7] the authors consider an abstract Cauchy problems in Banach spaces driven by a cylindrical Liouville fBm. The stochastic integral is defined by applying the principle of extension by continuity. When applied to the parabolic stochastic equations (1.1), it results in the requirement that the diffusion operator \( C \) must be \( \gamma \)-radonifying, i.e. the random perturbation \( C \, dB(t) \) becomes a genuine classical noise in the underlying Banach space. In our case, by following strictly the Gaussian path, we can still have a cylindrical noise. Furthermore, our cylindrical approach provides a natural candidate for the solution of (1.1). Since this candidate has a cylindrical Gaussian distribution, we can characterise the existence of a solution equivalently in terms of the corresponding covariance operator being associated to a Radon
measure. minimal requirement, corresponds to Brzeźniak + Jan, ich fuer cyli, extends to non-semimartingales.

It is shown that for $H < 1/2$ the theory for Liouville fBm is equivalent to the one for fBm, while for $H > 1/2$ the space of integrable functions is slightly different.

Since our paper is close to

Our methodology, based on cylindrical measures and cylindrical random variables, has the advantage that it is intrinsic in the sense that it does not require the construction of a larger space in which the cylindrical noise exists as a genuine stochastic process. Due to the connection between cylindrical measures and the theory of geometry of Banach spaces, our methodology relates the study of fBm and stochastic differential equations driven by fBm to other areas of mathematics, such as operator theory, functional analysis and harmonic analysis, therefore providing a wider range of tools and techniques.

Our long-term aim is to study general stochastic equations in Banach spaces driven by cylindrical fBms, which involves stochastic integration for random integrands. We are inspired by the paper of van Neerven et al. [30] in which they deal with the Wiener case. Here, the approach is based on a two-sided decoupling inequality which enables the authors to define the stochastic integral for random integrands by means of the integral for deterministic integrands. The latter is introduced in van Neerven and Weis [31], and we hope that our present work will play an analogous role for equations driven by fractional Brownian motions.

Only a few works deal with fBm in Banach spaces and related stochastic integration theory. Brzeźniak et al. [7] consider abstract Cauchy problems in Banach spaces driven by cylindrical Liouville fBm. It is shown that for $H < 1/2$ the theory for Liouville fBm is equivalent to the one for fBm, while for $H > 1/2$ the space of integrable functions is slightly different. In our paper we extend their results related to the Cauchy problem, as we consider mild and weak solutions and we obtain necessary and sufficient conditions for the existence of a solution. In contrast to [7], we do not assume any further regularity for the diffusion operator $C$, and therefore we keep the irregular character of the cylindrical noise in the space where the equation is considered. Note however, that in some special cases the authors in [7] get around this restriction by means of interpolation techniques. Furthermore, our approach enables us to guarantee the existence of a solution for $H > 1/2$ without any further constraints, whereas the results in this case in [7] are restricted either to analytic semigroups or to Banach spaces of type larger than 1. Another approach for the study of an evolution equation driven by a fractional Brownian motion is considered by Balan in [4]. The author considers a stochastic heat equation with infinite dimensional fractional noise by using Malliavin calculus, but her approach is strictly limited to a Sobolev-space context.

In the special case of Hilbert spaces, quite some literature on stochastic evolution equations with fBm noise can be found – see amongst others Grecksch and Anh [13], Duncan and coauthors in a series of papers [9, 10, 11, 12], Tindel et al. [28], Maslowski and Nualart [17], Gubinelli et
al. [14]. When restricted to the Hilbert space case, our approach is to some extent similar to the one in [9, 11, 12]. But our method has the advantage of providing not only sufficient, but also necessary conditions for the existence of a solution, which turns out to be both mild and weak. Tindel et al. in [28], who also provide necessary and sufficient conditions for the existence of a solution, derive their results under a spectral gap assumption on the semigroup, which is assumed to be self-adjoint. Our approach enables us to avoid such kind of restrictive assumptions.

The paper is structured as follows. Section 2 contains a brief overview of cylindrical measures and cylindrical processes in Banach spaces, with emphasis on cylindrical Gaussian processes. In Section 3 we recall the construction of the Wiener integral for real valued fBm and its relation to fractional integral and derivative operators. In Section 4 we introduce cylindrical fBms in separable Banach spaces and provide a characterisation in terms of a series representation. We illustrate our notion of fBm by several examples, such as anisotropic fBm, i.e. spatially non-symmetric noise, and we give conditions under which such cylindrical noises are genuine fBms in the underlying space. Section 5 is dedicated to the construction and the study of the stochastic integral in a Banach space. In Section 6 we use this integral to construct the fractional Ornstein-Uhlenbeck process as the mild and weak solution of a abstract stochastic Cauchy problem in a Banach space. Finally, in Section 7 we consider the special case of the stochastic heat equation with fractional noise in a Hilbert space and compare our results with the existing literature.

2. Preliminaries

Throughout this paper, $U$ indicates a separable Banach space over $\mathbb{R}$ with norm $\| \cdot \|_U$. The topological dual of $U$ is denoted by $U^*$ and the algebraic one by $U''$. For $u^* \in U^*$ we indicate the dual pairing by $\langle u, u^* \rangle$. If $U$ is a Hilbert space we identify the dual space $U^*$ with $U$. The Borel $\sigma$-algebra on a Banach space $U$ is denoted by $B(U)$. If $V$ is another Banach space then $L(U, V)$ denotes the space of bounded, linear operators from $U$ to $V$ equipped with the operator norm topology.

For a measure space $(S, S, \mu)$ we denote by $L^p_{\mu}(S; U)$, $p \geq 0$, the space of equivalence classes of measurable functions $f : S \to U$ with $\int \| f(s) \|^p_{U} \mu(ds) < \infty$. If $S \in B(\mathbb{R})$ and $\mu$ is the Lebesgue measure we use the notation $L^p(S; U)$.

Next we recall some notions about cylindrical measures and cylindrical random variables as it can be found in Badrikian [3] or Schwartz [27]. Let $\Gamma$ be a subset of $U^*$, $n \in \mathbb{N}$, $u_1^* , \ldots , u_n^* \in \Gamma$ and $B \in B(\mathbb{R}^n)$. A set of the form

$$\mathcal{Z}(u_1^* , \ldots , u_n^* , B) := \{ u \in U : (\langle u, u_1^* \rangle , \ldots , \langle u, u_n^* \rangle ) \in B \},$$

is called a cylindrical set. We denote by $\mathcal{Z}(U, \Gamma)$ the set of all cylindrical sets in $U$ for a given $\Gamma$. It turns out this is an algebra. Let $\mathcal{C}(U, \Gamma)$ be the generated $\sigma$-algebra. When $\Gamma = U^*$ the notation is $\mathcal{Z}(U)$ and $\mathcal{C}(U)$, respectively. If $U$ is separable then both the Borel $\sigma$-algebra $\mathcal{B}(U)$ and the cylindrical $\sigma$-algebra $\mathcal{C}(U)$ coincide.

A function $\mu : Z(U) \to [0, \infty]$ is called a cylindrical measure on $Z(U)$ if for each finite subset $\Gamma \subseteq U^*$ the restriction of $\mu$ to the $\sigma$-algebra $\mathcal{C}(U, \Gamma)$
is a measure. It is called finite if \( \mu(U) \) is finite and cylindrical probability measure if \( \mu(U) = 1 \).

For every function \( f : U \to \mathbb{C} \) which is measurable with respect to \( Z(U, \Gamma) \) for a finite subset \( \Gamma \subseteq U^* \), the integral \( \int f(u) \mu(du) \) is well defined as a complex valued Lebesgue integral if it exists. In particular, the characteristic function \( \varphi_\mu : U^* \to \mathbb{C} \) of a finite cylindrical measure \( \mu \) is defined by

\[
\varphi_\mu(u^*) := \int_U e^{i\langle u, u^* \rangle} \mu(du) \quad \text{for all } u^* \in U^*.
\]

Let \((\Omega, \mathcal{A}, P)\) be a probability space. A cylindrical random variable \( Z \) in \( U \) is a linear and continuous map

\[ Z : U^* \to L^0_\mu(\Omega; \mathbb{R}), \]

where \( L^0_\mu(\Omega; \mathbb{R}) \) is equipped with the topology of convergence in probability. The characteristic function of a cylindrical random variable \( Z \) is defined by

\[ \varphi_Z : U^* \to \mathbb{C}, \quad \varphi_Z(u^*) = E[\exp(i \langle Zu^* \rangle)]. \]

A cylindrical process in \( U \) is a family \((Z(t) : t \geq 0)\) of cylindrical random variables in \( U \).

Let \( Z : U^* \to L^0_\mu(\Omega; \mathbb{R}) \) be a cylindrical random variable in \( U \). The cylindrical distribution of \( Z \) is the cylindrical probability measure \( \mu \) defined by the prescription

\[ \mu(\mathcal{B}) := P((Zu_1^*, \ldots, Zu_n^*) \in \mathcal{B}), \]

for a cylindrical set \( \mathcal{B} = \mathcal{B}(u_1^*, \ldots, u_n^*; B) \) for \( u_1^*, \ldots, u_n^* \in U^* \) and \( B \in \mathcal{B}(\mathbb{R}^n) \).

The characteristic functions \( \varphi_\mu \) and \( \varphi_Z \) of \( \mu \) and \( Z \) coincide. Conversely, for every cylindrical measure \( \mu \) on \( Z(U) \) there exist a probability space \((\Omega, \mathcal{A}, P)\) and a cylindrical random variable \( Z : U^* \to L^0_\mu(\Omega; \mathbb{R}) \) such that \( \mu \) is the cylindrical distribution of \( Z \).

A cylindrical probability measure \( \mu \) on \( Z(U) \) is called Gaussian if the image measure \( \mu \circ (u^*)^{-1} \) is a Gaussian measure on \( \mathcal{B}(\mathbb{R}) \) for all \( u^* \in U^* \).

The characteristic function \( \varphi_\mu : U^* \to \mathbb{C} \) of a Gaussian cylindrical measure \( \mu \) is of the form

\[ \varphi_\mu(u^*) = \exp(i m(u^*) - \frac{1}{2} s(u^*)) \quad \text{for all } u^* \in U^*, \tag{2.1} \]

where the mappings \( m : U^* \to \mathbb{R} \) and \( s : U^* \to \mathbb{R}_+ \) are given by

\[ m(u^*) = \int_U \langle u, u^* \rangle \mu(du), \quad s(u^*) = \int_U \langle u, u^* \rangle^2 \mu(du) - m(u^*)^2. \]

Conversely, if \( \mu \) is a cylindrical measure with characteristic function of the form (2.1) for a linear functional \( m : U^* \to \mathbb{R} \) and a quadratic form \( s : U^* \to \mathbb{R}_+ \), then \( \mu \) is a Gaussian cylindrical measure.

For a Gaussian cylindrical measure \( \mu \) with characteristic function of the form (2.1) one defines the covariance operator \( Q : U^* \to (U^*)' \) by

\[ (Qu^*) v^* = \int_U \langle u, u^* \rangle \langle u, v^* \rangle \mu(du) - m(u^*)m(v^*) \quad \text{for all } u^*, v^* \in U^*. \]

Contrary to Gaussian Radon measures, the covariance operator might take values only in the algebraic dual of \( U^* \), that is the linear map \( Qu^* : U^* \to \mathbb{R} \) might be not continuous for some \( u^* \in U^* \). However often we exclude this rather general situation by requiring that at least \( Qu^* \) is norm continuous,
that is \( Q : U^* \to U^{**} \). Note that in this situation the characteristic function \( \varphi_\mu \) of \( \mu \) in (2.1) can be written as
\[
\varphi_\mu(u^*) = \exp \left( \frac{1}{2} \langle u^*, Qu^* \rangle \right) \quad \text{for all } u^* \in U^*.
\]
A cylindrical random variable \( Z : U^* \to L^0_P(\Omega; \mathbb{R}) \) is called Gaussian if its cylindrical distribution is Gaussian. Since we require from the cylindrical random variable \( Z \) to be continuous it follows that its characteristic function \( \varphi_Z : U^* \to \mathbb{C} \) is continuous. The latter occurs if and only if the covariance operator \( Q \) maps to \( U^{**} \).

3. Wiener integrals for Hilbert space valued integrands

In the following we recall the construction of the Wiener integral with respect to a real valued fractional Brownian motion for integrands which are Hilbert space valued deterministic functions. For real valued integrands the construction is accomplished for example in [5] and for Hilbert space valued integrands in [9, 11, 22].

We begin with recalling the definition of a fractional Brownian motion (fBm) and for later purpose, we introduce it in \( \mathbb{R}^n \). A Gaussian process \((b(t) : t \geq 0)\) in \( \mathbb{R}^n \) is an \( \mathbb{R}^n \)-valued fractional Brownian motion with Hurst parameter \( H \in (0, 1) \) if there exists a matrix \( M \in \mathbb{R}^{n \times n} \) such that
\[
E[\langle \alpha, b(s) \rangle] = 0, \quad E[\langle \alpha, b(s) \rangle \langle \beta, b(t) \rangle] = \langle M\alpha, \beta \rangle R(s, t)
\]
for all \( s, t \geq 0 \) and \( \alpha, \beta \in \mathbb{R}^n \), where
\[
R(s, t) := \frac{1}{2} \left( s^{2H} + t^{2H} - |s - t|^{2H} \right) \quad \text{for all } s, t \geq 0.
\]
The matrix \( M = (m_{i,j})_{i,j=1}^n \) is called the covariance matrix of the fBm \((b_1(t), \ldots, b_n(t)) : t \geq 0)\) in \( \mathbb{R}^n \) since it follows that
\[
m_{i,j} = E[b_i(1)b_j(1)] \quad \text{for all } i, j = 1, \ldots, n.
\]
Thus, \( M \) is a positive and symmetric matrix. If \( M = \text{Id} \) then \( b \) is called standard fractional Brownian motion. It follows from Kolmogorov’s continuity theorem by the Garsia-Rodemich-Rumsey inequality, that there exists a version of a fBm with Hölder continuous paths of any order smaller than \( H \).

We fix for the complete work the Hurst parameter and assume \( H \in (0, 1) \setminus \{ \frac{1}{2} \} \). The covariance function has an integral representation given by
\[
R(s, t) = \int_0^{s \wedge t} \kappa(s, u)\kappa(t, u) \, du \quad \text{for all } s, t \geq 0,
\]
where the kernel \( \kappa \) has different expressions depending on the Hurst parameter, see e.g. [5] Chapter 2. If \( H > \frac{1}{2} \) then
\[
\kappa(t, u) = b_H u^{1/2-H} \int_u^t (r - u)^{-3/2} r^{H-1/2} \, dr \quad \text{for all } 0 \leq u < t,
\]
Thus, \( E \) is a pre-Hilbert space. We denote the completion of \( E \) with respect to \( \langle \cdot, \cdot \rangle \) by \( M \).

Let \( (b(t) : t \geq 0) \) be a real valued fractional Brownian motion with Hurst parameter \( H \). For a simple, \( X \)-valued function \( f : [0, T] \to X \) of the form

\[
f(t) = \sum_{i=0}^{n-1} x_i \mathbb{1}_{[t_i, t_{i+1})}(t) \quad \text{for all } t \in [0, T],
\]

where \( x_i \in X \), \( 0 = t_0 < t_1 < \cdots < t_n = T \) and \( n \in \mathbb{N} \). The space of all simple, \( X \)-valued functions is denoted by \( \mathcal{E} \) and it is equipped with an inner product defined by

\[
\left\langle \sum_{i=0}^{m-1} x_i \mathbb{1}_{[0, s_i)} \cdot \sum_{j=0}^{n-1} y_j \mathbb{1}_{(0, t_j]}, \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} [x_i, y_j] R(s_i, t_j) \right\rangle_M := \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} [x_i, y_j] R(s_i, t_j).
\]

Thus, \( \mathcal{E} \) is a pre-Hilbert space. We denote the completion of \( \mathcal{E} \) with respect to \( \langle \cdot, \cdot \rangle_M \) by \( \mathcal{M} \).

The integral \( \int_0^T f \, db \) is a random variable which characterises an equivalent class in \( L^2_p(\Omega; X) \) (it will be denoted again by \( \int_0^T f \, db \) for simplicity). The map \( f \mapsto \int_0^T f \, db \) defines an isometry between \( \mathcal{E} \) and \( L^2_p(\Omega; X) \), since

\[
\left\| \int_0^T f \, db \right\|_{L^2_p}^2 = \| f \|_{\mathcal{M}}^2.
\]

Consequently, we can extend the mapping \( f \mapsto \int_0^T f \, db \) to the space \( \mathcal{M} \) and the extension still satisfies the isometry (3.4).

There is an alternative description of the space \( \mathcal{M} \) of possible integrands. For that purpose, we introduce the linear operator \( K^* : \mathcal{E} \to L^2(0, T; X) \), which is defined for all \( t \in [0, T] \) in case \( H < \frac{1}{2} \) by

\[
(K^* f)(t) := f(t)\kappa(T, t) + \int_t^T (f(s) - f(t)) \frac{\partial \kappa}{\partial s}(s, t) \, ds,
\]

and in case \( H > \frac{1}{2} \) by

\[
(K^* f)(t) := \int_t^T f(s) \frac{\partial \kappa}{\partial s}(s, t) \, ds.
\]
The integrals appearing on the right-hand side are both Bochner integrals. Since the operator $K^*$ satisfies
\[(K^* f, K^* g)_{L^2} = (f, g)_\mathcal{M} \quad \text{for all } f, g \in \mathcal{E}, \tag{3.5}\]
it can be extended to an isometry $K^*$ between $\mathcal{M}$ and $L^2([0,T]; X)$. Together with (3.4) we obtain
\[\left\| \int_0^T f \, db \right\|^2_{L^2_p} = \|K^* f\|^2_{L^2} = \|f\|^2_{\mathcal{M}} \quad \text{for all } f \in \mathcal{M}. \tag{3.6}\]

The operator $K^*$ can be rewritten using the notion of fractional integrals and derivatives. For this purpose, define for $\alpha > 0$ the fractional integral operator $I^\alpha_T : L^2([0,T]; X) \to L^2([0,T]; X)$ by
\[(I^\alpha_T f) (t) := \frac{1}{\Gamma(\alpha)} \int_t^T (s-t)^{\alpha-1} f(s) \, ds \quad \text{for all } t \in [0,T].\]
Young’s inequality guarantees that $I^\alpha_T f \in L^2([0,T]; X)$ and that the operator $I^\alpha_T$ is bounded on $L^2([0,T]; X)$. We define the space
\[H^\alpha_T([0,T]; X) := I^\alpha_T(L^2([0,T]; X))\]
and equip it with the norm
\[\|I^\alpha_T f\|_{H^\alpha_T} := \|f\|_{L^2} \quad \text{for all } f \in L^2([0,T]; X).\]
It follows that the space $H^\alpha_T([0,T]; X)$ is a Hilbert space and it is continuously embedded in $L^2([0,T]; X)$.

For $\alpha \in (0,1)$ the fractional differential operator $D^\alpha_T : H^\alpha_T([0,T]; X) \to L^2([0,T]; X)$ is defined by
\[(D^\alpha_T f)(t) := \frac{1}{\Gamma(1-\alpha)} \left( \frac{f(t)}{(T-t)^\alpha} + \alpha \int_t^T \frac{f(t) - f(s)}{(s-t)^{\alpha+1}} \, ds \right) \quad \text{for all } t \in [0,T].\]
The fractional integral and differential operators obey the inversion formulas (see e.g. [26] or [15, Section 2.1]):
\[I^\alpha_T(D^\alpha_T f) = f \quad \text{for all } f \in H^\alpha_T([0,T]; X),\]
and
\[D^\alpha_T(I^\alpha_T f) = f \quad \text{for all } f \in L^1([0,T]; X).\]

Let $p^{H-1/2}$ denote the function $p^{H-1/2}(t) = t^{H-1/2}$ for all $t \in [0,T]$. The operator $K^*$ can be rewritten in the case $H > 1/2$ as
\[(K^* f)(t) = b_H \Gamma \left( H - \frac{1}{2} \right) t^{1/2-H} I^H_{T-} \left( p^{H-1/2} f \right) (t) \tag{3.7}\]
for all $t \in [0,T]$ and in the case of $H < 1/2$ in the form
\[(K^* f)(t) = b_H \Gamma \left( H + \frac{1}{2} \right) t^{1/2-H} D^H_{T-} \left( p^{H-1/2} f \right) (t). \tag{3.8}\]
It can be seen from (3.7) that $\mathcal{M}$ contains distribution for $H > \frac{1}{2}$. Thus it became standard to consider a smaller space of integrands in place of the
space $\mathcal{M}$, see for example [5, 11, 23]. It turns out that an appropriate choice is the function space
\[ |\mathcal{M}| := \left\{ f : [0, T] \rightarrow X : \int_0^T \int_0^T \| f(s) \| \| f(t) \| | s - t |^{2H-2} \, ds \, dt < \infty \right\}, \]
equipped with the norm
\[ \| f \|_{|\mathcal{M}|}^2 = H(2H - 1) \int_0^T \int_0^T | f(s) \| \| f(t) \| | s - t |^{2H-2} \, ds \, dt. \]
The space $|\mathcal{M}|$ is complete and it is continuously embedded in $\mathcal{M}$. The proof of this fact is analogous to the real valued case, see e. g. [5, Pro.2.1.13]. If $H > \frac{1}{2}$ then the covariance function $R$ is differentiable with
\[ \frac{\partial^2 R}{\partial s \partial t}(s, t) = H(2H - 1) | s - t |^{2H-2} \quad \text{for all } s, t \geq 0, \]
and we can rewrite (3.3) as
\[ \langle f, g \rangle_{\mathcal{M}} = H(2H - 1) \int_0^T \int_0^T [ f(s), g(t) ] | s - t |^{2H-2} \, ds \, dt \quad (3.9) \]
for all simple functions $f, g \in \mathcal{E}$. Since $\mathcal{E}$ is dense in $|\mathcal{M}|$, equation (3.9) is true for all $f, g \in |\mathcal{M}|$, see [11, Eq.(2.14)].

We summarise the two cases by defining
\[ \widehat{\mathcal{M}} := \begin{cases} \mathcal{M} & \text{if } H \in (0, 1/2), \\ |\mathcal{M}| & \text{if } H \in (1/2, 1). \end{cases} \quad (3.10) \]
Recall that $\widehat{\mathcal{M}}$ is a Banach space and the operator $K^*$ satisfies
\[ \| K^* f \|_{L^2} \leq c \| f \|_{\widehat{\mathcal{M}}} \quad \text{for all } f \in \widehat{\mathcal{M}} \quad (3.11) \]
for a constant $c > 0$. Inequality (3.11) follows from (3.6) and, if $H > \frac{1}{2}$, from the continuous embedding $|\mathcal{M}| \hookrightarrow \mathcal{M}$. If $H < \frac{1}{2}$ we can choose $c = 1$.

In the sequel, we collect some properties of the spaces $\mathcal{M}$ and $|\mathcal{M}|$. Recall that the time interval $[0, T]$ is fixed. In our first result the coincidence of the spaces are well known, whereas we are only aware that the equivalence of the norms is stated in [7] but without a proof.

**Proposition 3.1.** For $H < \frac{1}{2}$ the spaces $\mathcal{M}$ and $H^{1/2-H}_{T^-}([-T, T]; X)$ coincide and the norms are equivalent.

**Proof.** The fact that the spaces coincide is shown in [1, Pro.6]. The proof of the equivalence of the norms is based on the following relation, which can be found in the proof of [1, Pro.6]:
\[ K^* f = a \left(D^{1/2-H}_{T^-} f\right) + Rf \quad \text{for all } f \in \mathcal{M}, \quad (3.12) \]
where $a := b_H \Gamma(H + \frac{1}{2})$ and $R : L^2([-T, T]; X) \rightarrow L^2([-T, T]; X)$ is a linear and continuous operator. Since $H^{1/2-H}_{T^-}([-T, T]; X)$ is continuously embedded in $L^2([-T, T]; X)$ there exists a constant $c > 0$ such that for each $f \in \mathcal{M}$ we have
\[ \| f \|_{\mathcal{M}} = \| K^* f \|_{L^2} \leq a \left\| D^{1/2-H}_{T^-} f \right\|_{L^2} + \left\| Rf \right\|_{L^2} \leq (a + c \| R \|) \| f \|_{H^{1/2-H}_{T^-}}. \]
On the other hand, the Hardy-Littlewood inequality in weighted spaces guarantees that $\mathcal{M}$ is continuously embedded in $L^{1/H}([0,T];X)$. More specifically, by choosing $p = 2, \alpha = \frac{1}{2} - H, m = 0, q = \frac{1}{H}, \mu = 2\alpha, \nu = q\alpha$ in [26, Th.5.4], we obtain for $f \in \mathcal{M}$

$$\|f\|_{L^q} = \left(\int_0^T \|f(t)\|^q \, dt \right)^{1/q}$$

$$= \left(\int_0^T t^\alpha \|t^{-\alpha}f(t)\|^q \, dt \right)^{1/q}$$

$$= \left(\int_0^T t^\alpha \|(I_{T-}^\alpha D_{T-}^{\alpha/p} - \alpha f)(t)\|^q \, dt \right)^{1/q}$$

$$\leq c \left(\int_0^T t^{\alpha} \|(D_{T-}^{\alpha/p} - \alpha f)(t)\|^p \, dt \right)^{1/p}$$

$$= c(b_H \Gamma(H + \frac{1}{2}))^{-1/p} \|K^*f\|_{L^2}$$

$$= c(b_H \Gamma(H + \frac{1}{2}))^{-1/p} \|f\|_{\mathcal{M}},$$

(3.13)

for a constant $c > 0$. Consequently, together with the continuous embedding of $L^{1/H}([0,T];X)$ in $L^2([0,T];X)$, it follows from (3.12) that each $f \in \mathcal{M}$ satisfies

$$a \|f\|_{H^{1/2-H}_{T-}} \leq \|K^*f\|_{L^2} + \|Rf\|_{L^2} = \left(1 + c(b_H \Gamma(H + \frac{1}{2}))^{-1/2} \|R\|\right) \|f\|_{\mathcal{M}},$$

which completes the proof. \hfill \square

**Proposition 3.2.** For every $t \in [0,T]$ there exists a constant $c_t > 0$ such that each $f \in \mathcal{M}$ obeys:

(a) $\mathbb{1}_{[0,t]}f \in \mathcal{M}$ and $\|\mathbb{1}_{[0,t]}f\|_{\mathcal{M}} \leq c_t \|f\|_{\mathcal{M}}$.

(b) $\mathbb{1}_{[0,t]}f(t - \cdot) \in \mathcal{M}$ and $\|\mathbb{1}_{[0,t]}f(t - \cdot)\|_{\mathcal{M}} = \|\mathbb{1}_{[0,T]}f\|_{\mathcal{M}}$.

**Proof.** If $H > \frac{1}{2}$, both properties (a) and (b) follow from (3.9) with $c_t = 1$ for all $t \in [0,T]$. If $H < \frac{1}{2}$, note that it is known for $f \in H^{1/2-H}_{T-}([0,T];X)$ that $\mathbb{1}_{[0,t]}f$ and $\mathbb{1}_{[0,t]}f(t - \cdot)$ are in $H^{1/2-H}_{T-}([0,T];X)$, see [26, Th.13.9, Th.13.10, Re.13.3] or [7, Le.2.1, Le.2.2]. Furthermore, there exists a constant $a_t > 0$ such that

$$\|\mathbb{1}_{[0,t]}f\|_{H^{1/2-H}_{T-}} \leq a_t \|f\|_{H^{1/2-H}_{T-}}.$$ 

Thus Proposition 3.1 implies part (a) and $\mathbb{1}_{[0,t]}f(t - \cdot) \in \mathcal{M}$. To show the norm equality in part (b), note the identity

$$\langle g, h \rangle = e_H \left\langle D^{1/2-H}_-, g, D^{1/2-H}_+ h \right\rangle_{L^2} \quad \text{for all } g, h \in \mathcal{M},$$

(3.14)

where $e_H$ denotes a constant depending only on $H$, see [21, page 286]. Here $D^{\alpha}_\pm$ denote the right-sided/left-sided Weyl-Marchaud fractional derivatives defined by

$$D^{\alpha}_\pm g(r) := \frac{\alpha}{\Gamma(1-\alpha)} \int_0^\infty \frac{g(r) - g(r \mp s)}{s^{1+\alpha}} \, ds \quad \text{for all } r \in \mathbb{R}.$$
It follows from (3.14) that

\[
\|1_{[0,t]}f(t-\cdot)\|^2_M = e_H\langle (\mathbb{D}_+^{1/2-H}1_{[0,t]}f(t-\cdot))\cdot, (\mathbb{D}_+^{1/2-H}1_{[0,t]}f(t-\cdot))\cdot \rangle
\]

\[
e_H\langle (\mathbb{D}_-^{1/2-H}1_{[0,t]}f(t-\cdot))\cdot, (\mathbb{D}_-^{1/2-H}1_{[0,t]}f(t-\cdot))\cdot \rangle
\]

\[
= \|1_{[0,t]}f\|^2_M,
\]

which completes the proof. \(\square\)

In the following we prove a technical result that links the real case, that is \(X = \mathbb{R}\) in the above, with the Hilbert case. For this reason we will stress the dependence on the underlying space by writing either \(K^*_R\) or \(K^*_X\). Analogous notation will be adopted for the space \(\mathcal{M}\).

**Proposition 3.3.**

(a) Let \(f\) be in \(\widetilde{\mathcal{M}}_R\) and \(x \in X\). Then

\[
F: [0,T] \to X, \quad F(t) = x f(t),
\]

belongs to \(\widetilde{\mathcal{M}}_X\) satisfying \((K^*_X F)(\cdot) = x(K^*_R f)(\cdot)\).

(b) Let \(F\) be in \(\mathcal{M}_X\) and \(x \in X\). Then

\[
f: [0,T] \to \mathbb{R}, \quad f(t) = [F(t),x],
\]

belongs to \(\widetilde{\mathcal{M}}_R\) satisfying \((K^*_X F)(\cdot), x) = (K^*_R f)(\cdot)\).

**Proof.** We prove only part (a) as part (b) can be done analogously. If \(H < \frac{1}{2}\), then by Proposition 3.1 there exists \(\varphi_f \in L^2([0,T];\mathbb{R})\) such that \(f = I^{1/2-H}_T \varphi_f\). Since \(x\varphi_f \in L^2([0,T];X)\) and \(F = xI^{1/2-H}_T \varphi_f = I^{1/2-H}_T x\varphi_f\), it follows that \(F \in \mathcal{M}_X\). If \(H > \frac{1}{2}\), the assumption \(f \in M_R\) implies \(F \in |M|_X\). In both cases, the very definition of \(K^*_X\) and \(K^*_R\) shows \(K^*_X F = xK^*_R f\). \(\square\)

4. Cylindrical fractional Brownian motion

We define cylindrical fractional Brownian motions in a separable Banach space \(U\) by following the classical approach of cylindrical processes. In the same way, one can introduce cylindrical Wiener processes, see for instance [16, 18, 24], and recently, this approach has been accomplished in [2] to give the first systematic treatment of cylindrical Lévy processes.

**Definition 4.1.** A cylindrical process \((B(t) : t \geq 0)\) in \(U\) is a cylindrical fractional Brownian motion with Hurst parameter \(H \in (0,1)\) if

(a) for any \(u_1^*, \ldots, u_n^* \in U^*\) and \(n \in \mathbb{N}\), the stochastic process

\[
((B(t)u_1^*, \ldots, B(t)u_n^*) : t \geq 0)
\]

is a fractional Brownian motion with Hurst parameter \(H\) in \(\mathbb{R}^n\);

(b) the covariance operator \(Q: U^* \to U^{**}\) of \(B(1)\) defined by

\[
(Qv^*, v^*) = \mathbb{E}[(B(1)u^*)(B(1)v^*)]
\]

for all \(u^*, v^* \in U^*\),

is \(U\)-valued.
By applying part (a) for \( n = 2 \) it follows that a cylindrical fBm \((B(t) : t \geq 0)\) with covariance operator \( Q \) obeys
\[
E[(B(s)u^*)(B(t)v^*)] = (Qu^*, v^*)R(s,t)
\]
fors, \( t \geq 0 \) and \( u^*, v^* \in U^* \). Note that if \( H = \frac{1}{2} \) then Definition 4.1 covers the cylindrical Wiener process as defined in [16, 18, 24].

Definition 4.1 involves all possible \( n \)-dimensional projections of the process, but since we are dealing with Gaussian processes the condition can be simplified using only two-dimensional projections.

**Lemma 4.2.** For a cylindrical process \( B := (B(t) : t \geq 0) \) in \( U \) the following are equivalent:

(a) \( B \) is a cylindrical fractional Brownian motion with Hurst parameter \( H \in (0, 1) \);

(b) \( B \) satisfies:

(i) for each \( u^*, v^* \in U^* \) the stochastic process \((B(t)u^*, B(t)v^*) : t \geq 0\) is a two-dimensional fBm;

(ii) the covariance operator of \( B(1) \) is \( U \)-valued.

**Proof.** We have to prove only the implication \((b) \Rightarrow (a)\). For \( u_1^*, \ldots, u_n^* \in U^* \) define the stochastic process \( Y = ((B(t)u_1^*, \ldots, B(t)u_n^*) : t \geq 0) \). It follows that \( Y \) is Gaussian and satisfies \( E[\langle \alpha, Y(t) \rangle] = 0 \) for all \( t \geq 0 \) and \( \alpha = (\alpha_1, \ldots, \alpha_n) \in \mathbb{R}^n \) since
\[
\langle \alpha, Y(t) \rangle = \sum_{i=1}^{n} \alpha_i B(t)u_i^* = B(t) \left( \sum_{i=1}^{n} \alpha_i u_i^* \right).
\]
Let \( M = (m_{i,j})_{i,j=1}^{n} \) be the \( n \)-dimensional matrix defined by
\[
m_{i,j} = E \left[ (B(1)u_i^*) (B(1)u_j^*) \right], \quad i, j = 1, \ldots, n.
\]
Since it follows from (b) that \( E[(B(s)u_i^*)(B(t)u_j^*)] = m_{i,j}R(s,t) \) for all \( s, t \geq 0 \) and \( i, j = 1, \ldots, n \) we obtain
\[
E[\langle \alpha, Y(s) \rangle \langle \beta, Y(t) \rangle] = E \left[ \sum_{i=1}^{n} \sum_{j=1}^{n} \alpha_i \beta_j (B(s)u_i^*) (B(t)u_j^*) \right]
\]
\[
= \sum_{i=1}^{n} \sum_{j=1}^{n} \alpha_i \beta_j m_{i,j} R(s,t)
\]
\[
= (M\alpha, \beta)R(s,t)
\]
for each \( \alpha = (\alpha_1, \ldots, \alpha_n) \) and \( \beta = (\beta_1, \ldots, \beta_n) \) in \( \mathbb{R}^n \).

The following result provides an analogue of the Karhunen-Loève expansion for cylindrical Wiener processes.

**Theorem 4.3.** For a cylindrical process \( B := (B(t) : t \geq 0) \) the following are equivalent:

(a) \( B \) is a cylindrical fractional Brownian motion with Hurst parameter \( H \in (0, 1) \);
(b) there exist a Hilbert space $X$ with an orthonormal basis $(e_k)_{k \in \mathbb{N}}$, $i \in \mathcal{L}(X, U)$ and a sequence $(b_k)_{k \in \mathbb{N}}$ of independent, real valued standard fBms with Hurst parameter $H \in (0, 1)$ such that
\[
B(t)u^* = \sum_{k=1}^{\infty} (ie_k, u^*)b_k(t) \tag{4.1}
\]
in $L_p^2(\Omega; \mathbb{R})$ for all $u^* \in U^*$ and $t \geq 0$.

In this situation the covariance operator of $B(1)$ is given by $Q = ii^* : U^* \rightarrow U$.

Proof. The implication (a) $\Rightarrow$ (b) can be proved as Theorem 4.8 in [2]. For establishing the implication (b) $\Rightarrow$ (a), it is immediate that the right hand side of (4.1) converges. Fix $u^*_1, \ldots, u^*_n \in U^*$ and define the $n$-dimensional stochastic process $Y := (Y(t) : t \geq 0)$ by
\[
Y(t) := (B(t)u^*_1, \ldots, B(t)u^*_n) \quad \text{for all } t \geq 0.
\]
It follows that $Y$ is Gaussian and satisfies $E[\langle \alpha, Y(t) \rangle] = 0$ for all $t \geq 0$ and $\alpha = (\alpha_1, \ldots, \alpha_n) \in \mathbb{R}^n$ since
\[
\langle \alpha, Y(t) \rangle = \sum_{i=1}^{n} \alpha_i B(t)u^*_i = B(t) \left( \sum_{i=1}^{n} \alpha_i u^*_i \right).
\]
Let $M = (m_{i,j})_{i,j=1}^{n}$ be the $n \times n$-dimensional covariance matrix of the random vector $Y(1)$, that is $m_{i,j} := E[\langle (B(1)u^*_i)(B(1)u^*_j) \rangle]$. The definition of $Y$ yields
\[
m_{i,j} = \sum_{k=1}^{\infty} \sum_{\ell=1}^{\infty} \langle ie_k, u^*_i \rangle \langle ie_\ell, u^*_j \rangle E[b_k(1)b_\ell(1)] = \sum_{k=1}^{\infty} \langle ie_k, u^*_i \rangle \langle ie_k, u^*_j \rangle.
\]
Let $\alpha = (\alpha_1, \ldots, \alpha_n) \in \mathbb{R}^n$ and $\beta = (\beta_1, \ldots, \beta_n) \in \mathbb{R}^n$. By using the independence of $b_k$ and $b_\ell$ for each $k \neq \ell$ we obtain for every $s, t \geq 0$
\[
E[\langle \alpha, Y(s) \rangle \langle \beta, Y(t) \rangle] = E \left[ \left( \sum_{i=1}^{n} \alpha_i \sum_{k=1}^{\infty} \langle ie_k, u^*_i \rangle b_k(s) \right) \left( \sum_{\ell=1}^{n} \beta_\ell \sum_{\ell=1}^{\infty} \langle ie_\ell, u^*_\ell \rangle b_\ell(t) \right) \right]
\]
\[
= \sum_{i=1}^{n} \sum_{j=1}^{n} \alpha_i \beta_j \sum_{k=1}^{\infty} \sum_{\ell=1}^{\infty} \langle ie_k, u^*_i \rangle \langle ie_\ell, u^*_j \rangle E[b_k(s)b_\ell(t)]
\]
\[
= \sum_{i=1}^{n} \sum_{j=1}^{n} \alpha_i \beta_j \sum_{k=1}^{\infty} \langle ie_k, u^*_i \rangle \langle ie_k, u^*_j \rangle E[b_k(s)b_k(t)]
\]
\[
= \sum_{i=1}^{n} \sum_{j=1}^{n} \alpha_i \beta_j m_{i,j}R(s, t)
\]
\[
= \langle Ma, \beta \rangle R(s, t).
\]
It is left to prove that $B(1) : U^* \rightarrow L^0_p(\Omega; \mathbb{R})$ is continuous and its covariance operator $Q : U^* \rightarrow U^{\ast\ast}$ is $U$-valued. By independence of $b_k$ and $b_\ell$ for $k \neq \ell$
it follows for \( u^* \in U^* \) that
\[
\varphi_{B(1)}(u^*) = \prod_{k=1}^{\infty} E[\exp(i\langle e_k, u^* \rangle b_k(1))]
\]
\[
= \prod_{k=1}^{\infty} \exp(-\frac{1}{2} \langle i e_k, u^* \rangle^2) = \exp\left(-\frac{1}{2} \|i^* u^*\|_X^2 \right).
\]
Thus, the characteristic function \( \varphi_{B(1)} : U^* \to \mathbb{C} \) is continuous, which entails the continuity of \( B(1) \) by [29, Pro. IV.3.4]. Moreover, it follows that \( Q = ii^* \), that is the covariance operator \( Q \) is \( U \)-valued and of the claimed form. \( \square \)

**Example 4.4.** Let \( U \) be a Hilbert space with orthonormal basis \((e_k)_{k \in \mathbb{N}}\), identify the dual space \( U^* \) with \( U \), and let \((q_k)_{k \in \mathbb{N}} \subseteq \mathbb{R} \) be a sequence satisfying \( \sup_{k \in \mathbb{N}} |q_k| < \infty \). It follows by Theorem 4.3 that for an arbitrary sequence \((b_k)_{k \in \mathbb{N}}\) of independent, real valued standard fBms, the series
\[
B(t)u := \sum_{k=1}^{\infty} q_k \langle e_k, u \rangle b_k(t), \quad u \in U,
\]
defines a cylindrical fBm \((B(t) : t \geq 0)\) in \( U \). The covariance operator \( Q \) is given by \( Q = ii^* \), where \( i : U \to U \) is defined as \( iu = \sum_{k=1}^{\infty} q_k \langle e_k, u \rangle e_k \).

**Example 4.5.** For a set \( D \in \mathcal{B}(\mathbb{R}^n) \) let \((e_k)_{k \in \mathbb{N}} \subseteq L^2(D; \mathbb{R})\) be an orthonormal basis and let \((\tau_k)_{k \in \mathbb{N}} \) be a sequence of functions \( \tau_k \in L^2(D; \mathbb{R}) \) satisfying \( \sum_{k=1}^{\infty} \|\tau_k\|_{L^2}^2 < \infty \). Applying Cauchy-Schwarz inequality twice shows that
\[
i : L^2(D; \mathbb{R}) \to L^1(D; \mathbb{R}), \quad if = \sum_{k=1}^{\infty} \langle e_k, f \rangle \tau_k(\cdot)e_k(\cdot) \quad (4.2)
\]
defines a linear and continuous mapping. It follows from Theorem 4.3 that for an arbitrary sequence \((b_k)_{k \in \mathbb{N}}\) of independent, real valued standard fBm, the series
\[
B(t)f := \sum_{k=1}^{\infty} \langle i e_k, f \rangle b_k(t), \quad f \in L^\infty(D; \mathbb{R}),
\]
defines a cylindrical fBm \((B(t) : t \geq 0)\) in \( L^1(D; \mathbb{R}) \) with covariance operator \( Q = ii^* : L^\infty(D; \mathbb{R}) \to L^1(D; \mathbb{R}) \).

**Example 4.6.** A special case of Example 4.5 is obtained by choosing the functions \( \tau_k \in L^2(D; \mathbb{R}) \) as \( \tau_k = q_k \mathbb{1}_{A_k} \) for \( q_k \in \mathbb{R} \) and \( A_k \in \mathcal{B}(D) \) satisfying \( \sum_{k=1}^{\infty} q_k^2 \text{Leb}(A_k) < \infty \). Then the cylindrical fBm of Example 4.5 has the form
\[
B(t)f = \sum_{k=1}^{\infty} q_k \langle \mathbb{1}_{A_k} e_k, f \rangle b_k(t).
\]
This process can be considered as an anisotropic cylindrical fractional Brownian sheet in \( L^1(D; \mathbb{R}) \) since its covariance structure might vary in different directions.

In the final part of this section we consider the relation between cylindrical and genuine fractional Brownian motion in a separable Banach space \( U \). For this purpose, we generalise the definition of a fractional Brownian motion.
in \( \mathbb{R}^n \) to Banach spaces. This definition is consistent with others in the literature, in particular the one in [9] for Hilbert spaces.

**Definition 4.7.** A \( U \)-valued Gaussian stochastic process \((Y(t) : t \geq 0)\) is called a fractional Brownian motion in \( U \) with Hurst parameter \( H \in (0, 1) \) if there exists a mapping \( Q : U^* \to U \) such that

\[
(Y(t), u^*) = 0, \quad E[(Y(s), u^*)(Y(t), v^*)] = \langle Qu^*, v^* \rangle R(s, t)
\]

for all \( s, t \geq 0 \) and \( u^*, v^* \in U^* \).

By taking \( s = t = 1 \) it follows that

\[
\langle Qu^*, v^* \rangle = E[(Y(1), u^*)(Y(1), v^*)]
\]

for all \( u^*, v^* \in U^* \).

Thus, \( Q \) is the covariance operator of the Gaussian measure \( P_{Y(1)} \) and it must be a symmetric and positive operator in \( L(U^*, U) \).

Clearly every fBm in a Banach space \( U \) is a cylindrical fBm in \( U \) and thus, it obeys the representation (4.1). However, the operator \( i \), or in other words the embedding of the reproducing kernel Hilbert space, must yield a Radon measure in \( U \), which basically leads to the following result:

**Theorem 4.8.** For a \( U \)-valued stochastic process \( Y := (Y(t) : t \geq 0) \) the following are equivalent:

(a) \( Y \) is a fBm in \( U \) with Hurst parameter \( H \in (0, 1) \);
(b) there exist a Hilbert space \( X \) with an orthonormal basis \((e_k)_{k \in \mathbb{N}}\), a \( \gamma \)-radonifying operator \( i \in L(X, U) \) and independent, real valued standard fBms \((b_k)_{k \in \mathbb{N}}\) such that

\[
Y(t) = \sum_{k=1}^{\infty} i e_k b_k(t)
\]

in \( L^2_P(\Omega; U) \) for all \( t \geq 0 \).

In this situation the covariance operator of \( Y(1) \) is given by \( Q = ii^* : U^* \to U \).

**Proof.** The result can be proved as Theorem 23 in [24]. \( \square \)

In the literature a fractional Brownian motion in a Hilbert space is often defined by a series representation as in Theorem 4.8, in which case the space of \( \gamma \)-radonifying operators coincides with Hilbert-Schmidt operators.

If \((B(t) : t \geq 0)\) is a cylindrical fBm which is induced by a \( U \)-valued process \((Y(t), t \geq 0)\), i.e.

\[
B(t) u^* = \langle Y(t), u^* \rangle \quad \text{for all } t \geq 0, u^* \in U^*, \tag{4.3}
\]

then \( Y \) is a \( U \)-valued fBm. Vice versa, if \( Y \) is a \( U \)-valued fBm then \( B \) defined by (4.3) is a cylindrical fBm, and in both cases the covariance operators coincide. This can be seen by the fact that (4.3) determines uniquely the characteristic functions of \((B(s), B(t))\) and \((Y(s), Y(t))\) for all \( s, t \geq 0 \). Moreover, a cylindrical fBm with the representation (4.1) is a \( U \)-valued fBm if and only if the embedding \( i \) is \( \gamma \)-radonifying. This result can be established as in [24, Th.25].
Example 4.9. If we assume in Example 4.5 that the functions $\tau_k$ are in $L^\infty(D;\mathbb{R})$ and satisfy $\sum \|\tau_k\|_\infty < \infty$ then the mapping $i$, defined in (4.2), maps to $L^2(D;\mathbb{R})$. Moreover, $i$ is a Hilbert-Schmidt operator, as
\[
\sum_{k=1}^{\infty} \|ie_k\|_2^2 = \sum_{k=1}^{\infty} \|\tau_k e_k\|_2^2 \leq \sum_{k=1}^{\infty} \|\tau_k\|_\infty^2 .
\]
Since $\gamma$-radonifying and Hilbert-Schmidt operators coincide in Hilbert spaces, Theorem 4.8 implies that the cylindrical fBm in Example 4.9 is induced by a genuine fractional Brownian motion in $L^2(D;\mathbb{R})$.

5. Integration

In this section we introduce the stochastic integral $\int \Psi(s) dB(s)$ as a $V$-valued random variable for deterministic, operator valued functions $\Psi : [0,T] \to \mathcal{L}(U,V)$, where $V$ is another separable Banach space. Our approach is based on the idea to introduce firstly a cylindrical random variable $Z_\Psi : V^* \to L^0_P(\Omega;\mathbb{R})$ as a cylindrical integral. Then we call a $V$-valued random variable $I_\Psi : \Omega \to V$ the stochastic integral of $\Psi$ if it satisfies
\[
Z_\Psi v^* = \langle I_\Psi, v^* \rangle \quad \text{for all } v^* \in V^* .
\]
In this way, the stochastic integral $I_\Psi$ can be considered as a stochastic Petits integral. This approach enables us to have a candidate of the stochastic integral, i.e. the cylindrical random variable $Z_\Psi$, under very mild conditions at hand because cylindrical random variables are more general objects than genuine random variables. The final requirement, that the cylindrical random variable $Z_\Psi$ is in fact a classical Radon random variable, can be equivalently described in terms of the corresponding covariance operator and thus, it solely depends on geometric properties of the underlying Banach space $V$.

For defining the cylindrical integral, recall the representation of a cylindrical fBm $(B(t) : t \geq 0)$ with Hurst parameter $H \in (0,1)$ in the Banach space $U$, according to Theorem 4.3:
\[
B(t)u^* = \sum_{k=1}^{\infty} \langle ie_k, u^* \rangle b_k(t) \quad \text{for all } u^* \in U^*, t \geq 0 .
\]
Here, $X$ is a Hilbert space with an orthonormal basis $(e_k)_{k \in \mathbb{N}}$, $i : X \to U$ is a linear, continuous mapping and $(b_k)_{k \in \mathbb{N}}$ is a sequence of independent, real valued standard fBms. If we assume momentarily that we have already introduced a stochastic integral $\int_0^T \Psi(t) dB(t)$ as a $V$-valued random variable, then the representation (5.1) of $B$ naturally results in
\[
\sum_{k=1}^{\infty} \int_0^T \langle \Psi(t)ie_k, v^* \rangle db_k(t) \quad \text{for all } v^* \in V^* .
\]
By swapping the terms in the dual pairing, the integrals can be considered as the Fourier coefficients of the $X$-valued integral
\[
\int_0^T i^* \Psi^*(t)v^* db_k(t) ,
\]
which we introduce in Section 3. This results in the minimal requirement that the function \( t \mapsto i^*\Psi^*(t)v^* \) must be integrable with respect to the real valued standard \( \text{BM} b_k \) for every \( v^* \in V^* \) and \( k \in \mathbb{N} \), that is the function \( \Psi \) must be in the linear space

\[ \mathcal{I} := \{ \Phi : [0, T] \to \mathcal{L}(U, V) : i^*\Psi^*(\cdot)v^* \in \hat{\mathcal{M}} \text{ for all } v^* \in V^* \}. \]

Here, \( \hat{\mathcal{M}} = \hat{\mathcal{M}}_X \) denotes the Banach space of functions \( f : [0, T] \to X \) introduced in Section 3. For this class of integrands we have the following property.

**Proposition 5.1.** For each \( \Psi \in \mathcal{I} \) the mapping

\[ L_{\Psi} : V^* \to \hat{\mathcal{M}}, \quad L_{\Psi} v^* = i^*\Psi^*(\cdot)v^* \]

is linear and continuous.

**Proof.** The operator \( L = L_{\Psi} \) is linear and takes values in \( \hat{\mathcal{M}} \) by definition of \( \mathcal{I} \). We prove that \( L \) is continuous by the closed mapping theorem. For this purpose, let \( v^*_n \to v^*_0 \) in \( V^* \) and \( L_{\Psi} v^*_n \to g \in \hat{\mathcal{M}} \). We consider the cases \( H < 1/2 \) and \( H > 1/2 \) separately.

- **Case** \( H < 1/2 \). From the Hardy-Littlewood inequality in weighted spaces, see (3.13), it follows that

\[ \|f\|_{L^q} \leq c(b_\Gamma(H + \frac{1}{2}))^{-1/2} \|f\|_{\mathcal{M}} \quad \text{for all } f \in \mathcal{M} \]

for a constant \( c > 0 \) and \( q = \frac{1}{n} \). Consequently, the convergence \( L_{\Psi} v^*_n \to g \) in \( \mathcal{M} \) implies that there exists a subsequence \( (n_k)_{k \in \mathbb{N}} \subseteq \mathbb{N} \) such that \( L_{\Psi} v^*_n(t) \to g(t) \) as \( k \to \infty \) for Lebesgue almost all \( t \in [0, T] \). On the other hand, we have \( i^*\Psi^*(t)v^*_n \to i^*\Psi^*(t)v^*_0 \) in \( X \) as \( k \to \infty \) for all \( t \in [0, T] \), because \( i^* \) and \( \Psi^*(t) \) are continuous. Consequently, we arrive at \( g(t) = i^*\Psi^*(t)v^*_0 \) for Lebesgue almost all \( t \in [0, T] \), and thus, \( g = L_{\Psi} v^*_0 \) as functions in \( L^2([0, T]; X) \).

- **Case** \( H > 1/2 \). In this case \( \hat{\mathcal{M}} = |\mathcal{M}| \). Let us remark, that if \( f \in |\mathcal{M}| \) then \( f \in L^1([0, T]; X) \) and

\[ (2T)^{2H-2}\|f\|^2_{L^1} = (2T)^{2H-2} \int_0^T \int_0^T \|f(s)||f(t)|| s \, ds \, dt \]

\[ \leq \int_0^T \int_0^T \|f(s)||f(t)|| |s-t|^{2H-2} \, ds \, dt \]

\[ = \frac{1}{H(2H-1)} \|f\|^2_{|\mathcal{M}|}. \]

Using this fact, the convergence \( L_{\Psi} v^*_n \to g \) in \( |\mathcal{M}| \) implies that \( L_{\Psi} v^*_n(t) \to g(t) \) as \( k \to \infty \) for Lebesgue almost all \( t \in [0, T] \) for a subsequence \( (n_k)_{k \in \mathbb{N}} \subseteq \mathbb{N} \).

The continuity of the mapping \( v^* \mapsto i^*\Psi^*(t)v^* \) for all \( t \in [0, T] \) shows that \( g(t) = i^*\Psi^*(t)v^*_0 \) for Lebesgue almost all \( t \in [0, T] \) and thus, \( g = L_{\Psi} v^*_0 \) in \( |\mathcal{M}| \).

Before we establish the existence of the cylindrical integral as motivated in (5.2), we introduce an operator which will turn out to be the factorisation of the covariance operator of the cylindrical integral.
Lemma 5.2. For every $\Psi \in \mathcal{I}$ we define

$$
\langle \Gamma \psi, \nu^* \rangle = \int_0^T [K^s(i^s \Psi^* \nu^*)(t), f(t)] \, dt \quad \text{for all } f \in L^2([0, T]; X), \nu^* \in V^*.
$$

In this way, one obtains a linear, bounded operator $\Gamma : L^2([0, T]; X) \to V^{**}$.

Proof. Proposition 5.1, together with equation (3.11), implies

$$
\langle \Gamma \psi, \nu^* \rangle = |\langle K^s(i^s \Psi^*(-)\nu^*), f \rangle|_2 
\leq c_1 \|i^s \Psi^*(-)\nu^*\|_{\hat{M}} \|f\|_{L^2} \leq c_2 \|\nu^*\|_{V^*} \|f\|_{L^2},
$$

for some constants $c_1, c_2 > 0$, which shows boundedness of $\Gamma \psi$.

\[ \Box \]

Proposition 5.3. Let the fBm $B$ be represented in the form (5.1). Then for each $\Psi \in \mathcal{I}$ the mapping

$$
Z_\psi : V^* \to L^2_P(\Omega; \mathbb{R}), \quad Z_\psi \nu^* := \sum_{k=1}^{\infty} \int_0^T (\Psi(t)i\epsilon_k, \nu^*) \, db_k(t)
$$

(5.3)
defines a Gaussian cylindrical random variable in $V$ with covariance operator $Q_\psi : V^* \to V^{**}$, factorised by $Q_\psi = \Gamma \psi \Gamma^*_\psi$. Furthermore, the cylindrical random variable $Z_\psi$ is independent of the representation (5.1).

Proof. Since $\langle \Psi(-)i\epsilon_k, \nu^* \rangle = [\epsilon_k, i^s \Psi^*(-)\nu^*]$ and $i^s \Psi^*(-)\nu^* \in \hat{M}$ for every $\nu^* \in V^*$, Proposition 3.3 guarantees that the one-dimensional integrals in (5.3) are well defined, and it implies that

$$
\|Z_\psi \nu^*\|_{L^2_P}^2 = \sum_{k=1}^{\infty} E \left| \int_0^T (\Psi(t)i\epsilon_k, \nu^*) \, db_k(t) \right|^2 
= \sum_{k=1}^{\infty} \int_0^T |K^*_\psi((\Psi(-)i\epsilon_k, \nu^*))(t)|^2 \, dt 
= \sum_{k=1}^{\infty} \int_0^T |K^*_\psi([\epsilon_k, i^s \Psi^*(-)\nu^*])(t)|^2 \, dt 
= \sum_{k=1}^{\infty} \int_0^T [\epsilon_k, K^*_\chi(i^s \Psi^*(-)\nu^*))(t)]^2 \, dt 
= \sum_{k=1}^{\infty} \int_0^T [\epsilon_k, (\Gamma^*_\psi \nu^*)(t)]^2 \, dt 
= \|\Gamma^*_\psi \nu^*\|_{L^2_P}^2.
$$

Consequently, the sum in (5.3) converges in $L^2_P(\Omega; \mathbb{R})$ and the limit is a zero mean Gaussian random variable. The continuity of the operator $\Gamma^*_\psi : V^* \to L^2([0, T]; X)$ implies the continuity of $Z_\psi : V^* \to L^2_P(\Omega; \mathbb{R})$. It follows for the characteristic function of $Z_\psi$ that

$$
\varphi_{Z_\psi}(\nu^*) = \exp \left( -\frac{1}{2} \|\Gamma^*_\psi \nu^*\|_{L^2_P}^2 \right) \quad \text{for all } \nu^* \in V^*.
$$

Since Lemma 5.2 implies that

$$
\|\Gamma^*_\psi \nu^*\|_{L^2_P}^2 = (\Gamma \psi, \Gamma^*_\psi \nu^*, \nu^*) \quad \text{for all } \nu^* \in V^*,
$$

it follows that the covariance operator $Q_\psi$ of $Z_\psi$ obeys $Q_\psi = \Gamma \psi \Gamma^*_\psi$. 


The independence of $Z_\Psi$ of the representation (5.1) can be established as in [24, Le.2]. □

For $\Psi \in \mathcal{I}$ we call the cylindrical random variable $Z_\Psi$, defined in (5.3), the \textit{cylindrical integral} of $\Psi$. Apart from the restriction of the space $\mathcal{M}$ of all integrable distributions to $\mathcal{M}$, the condition for a mapping $\Psi$ to be in $\mathcal{I}$ is the minimal requirement to guarantee that the real valued integrals in (5.2) exist. Thus without any further condition the cylindrical integral $Z_\Psi$ exists in the Banach space $U$. However, in order to obtain that the cylindrical integral $Z_\Psi$ extends to a genuine random variable in $U$, the integrand must exhibit further properties.

\textbf{Definition 5.4.} A function $\Psi \in \mathcal{I}$ is called \textit{stochastically integrable} if there exists a random variable $I_\Psi : \Omega \to V$ such that

$$Z_\Psi v^* = \langle I_\Psi, v^* \rangle \quad \text{for all } v^* \in V^*,\nonumber$$

where $Z_\Psi$ denotes the cylindrical integral of $\Psi$. We use the notation

$$I_\Psi := \int_0^T \Psi(t) \, dB(t).\nonumber$$

In other words, a function $\Psi \in \mathcal{I}$ is stochastically integrable if and only if the cylindrical random variable $Z_\Psi$ is induced by a Radon random variable. This occurs if and only if the cylindrical distribution of $Z_\Psi$ extends to a Radon measure. In Sazonov spaces this is equivalent to the condition that the characteristic function of $Z_\Psi$ is Sazonov continuous. However, since the cylindrical distribution of $Z_\Psi$ is Gaussian, one can equivalently express the stochastic integrability in terms of the covariance operator.

\textbf{Theorem 5.5.} For $\Psi \in \mathcal{I}$ the following are equivalent:

(a) $\Psi$ is stochastically integrable;

(b) the operator $\Gamma_\Psi$ is $V$-valued and $\gamma$-radonifying.

\textit{Proof.} (b) $\Rightarrow$ (a). Let $\gamma$ be the canonical Gaussian cylindrical measure on $L^2([0,T];X)$. It follows from Proposition 5.3 that the cylindrical distribution of $Z_\Psi$ is the image cylindrical measure $\gamma \circ \Gamma_\Psi^{-1}$. According to [29, Thm.IV.2.5, p.216], the cylindrical random variable $Z_\Psi$ is induced by a $V$-valued random variable if and only if its cylindrical distribution $\gamma \circ \Gamma_\Psi^{-1}$ extends to a Radon measure on $\mathcal{B}(V)$, which is guaranteed by (b).

(a) $\Rightarrow$ (b). The proof follows closely some arguments in the proof of Theorem 2.3 in [31]. Let $Q : V^* \to V$ be the covariance operator of the Gaussian random variable $\int \Psi(t) \, dB(t)$. Proposition 5.3 implies that $Q = \Gamma_\Psi \Gamma_\Psi^* : V^* \to V$. Define the set $S := \{K^*(i^* \Psi^*(\cdot) v^*) : v^* \in V^*\}$, which is a subset of $L^2([0,T];X)$. By the very definition of $\Gamma_\Psi$, a function $f \in L^2([0,T];X)$ is in ker $\Gamma_\Psi$ if and only if $f \perp S$, which yields

$$L^2([0,T];X) = \tilde{S} \oplus \ker \Gamma_\Psi. \quad (5.4)\nonumber$$

Since for all $v^*, w^* \in V^*$ we have

$$\langle \Gamma_\Psi K^*(i^* \Psi^*(\cdot) v^*), w^* \rangle = \langle \Gamma_\Psi \Gamma_\Psi^* v^*, w^* \rangle = \langle Q v^*, w^* \rangle, \nonumber$$
it follows that $\Gamma_\Psi K^*(i^* \Psi^*(\cdot) v^*) = Q v^*$ for all $v^* \in V^*$. Consequently, $\Gamma_\Psi f \in V$ for all $f \in S$ and the decomposition (5.4) implies that $\Gamma_\Psi f \in V$.
for all \( f \in L^2([0,T];X) \). Clearly, since \( Q \) is a Gaussian covariance operator, the operator \( \Gamma_\Psi \) is \( \gamma \)-radonifying, which completes the proof. \( \square \)

**Corollary 5.6.** If \( \Phi \) and \( \Psi \) are mappings in \( \mathcal{I} \) satisfying

\[
\| i^* \Phi^* (\cdot) v^* \|_{\mathcal{M}} \leq c \| i^* \Psi^* (\cdot) v^* \|_{\mathcal{M}} \quad \text{for all } v^* \in V^*,
\]

for a constant \( c > 0 \) and if \( \Psi \) is stochastically integrable then \( \Phi \) is also stochastically integrable.

**Proof.** The proof follows some arguments in the proof of Theorem 2.3 in [31]. Define the operator \( Q := \Gamma_\Psi \Gamma_\Phi^* : V^* \to V^{**} \). The isometry (3.6) implies for every \( v^* \in V^* \) that

\[
\langle v^*, Q v^* \rangle^2 = \langle \Gamma_\Phi^* v^*, \Gamma_\Phi^* v^* \rangle_{L^2}^2 = \left\| K^* (i^* \Phi^* (\cdot) v^*) \right\|_{L^2}^2 \\
\quad \leq c \left\| i^* \Phi^* (\cdot) v^* \right\|_{\mathcal{M}}^2 = c \langle \Gamma_\Psi \Gamma_\Phi^* v^*, v^* \rangle.
\]

Since \( \Gamma_\Psi \Gamma_\Phi^* \) and \( Q \) are positive, symmetric operators in \( L(V^*, V^{**}) \) and the first one is \( V \)-valued according to Theorem 5.5, it follows by an argument based on a result of the domination of Gaussian measures, see [31, Sec.1.1], that \( Q \) is also \( V \)-valued and a Gaussian covariance operator. As in the proof of the implication (a) \( \Rightarrow \) (b) in Theorem 5.5 we can conclude that \( \Gamma_\Phi \) is \( V \)-valued. \( \square \)

If the mapping \( \Psi \in \mathcal{I} \) is stochastically integrable, Proposition 3.2 implies for each \( t \in [0,T] \) that \( \mathbb{1}_{[0,t]} \Psi \in \mathcal{I} \) and it satisfies

\[
\left\| \mathbb{1}_{[0,t]} i^* \Psi^* (\cdot) v^* \right\|_{\mathcal{M}} \leq c_t \left\| i^* \Psi^* (\cdot) v^* \right\|_{\mathcal{M}} \quad \text{for all } v^* \in V^*,
\]

for a constant \( c_t > 0 \). Corollary 5.6 enables us to conclude that \( \mathbb{1}_{[0,t]} \Psi \) is stochastically integrable, and thus we can define

\[
\int_0^t \Psi(s) \, dB(s) := \int_0^T \mathbb{1}_{[0,t]}(s) \Psi(s) \, dB(s) \quad \text{for all } t \in [0,T].
\]

The integral process \( \left( \int_0^t \Psi(s) \, dB(s) : t \in [0,T] \right) \) is continuous in \( p \)-th mean for each \( p \geq 1 \). In order to see that let \( t_n \to t \) as \( n \to \infty \) for \( t_n \geq t \) and let \( Q^{(n)}_\Psi \) denote the covariance operator of the Gaussian random variable \( \int_{t_n}^t \Psi(s) \, dB(s) \). It follows for each \( v^* \in V^* \) that

\[
\langle Q^{(n)}_\Psi v^*, v^* \rangle = \left\| K^* (\mathbb{1}_{[t_n,t]} (\cdot)) i^* \Psi^* (\cdot) v^* \right\|_{L^2}^2 = \left\| \mathbb{1}_{[t_n,t]} (\cdot) i^* \Psi^* (\cdot) v^* \right\|_{\mathcal{M}}^2.
\]

Each \( f \in \widehat{\mathcal{M}} \) satisfies \( \left\| \mathbb{1}_{[t_n,t]}(\cdot) f \right\|_{\mathcal{M}} \to 0 \) as \( t_n \to t \) which follows from (3.9) in case \( H > \frac{1}{2} \) and from results in [26, Ch.13.3] in case \( H < \frac{1}{2} \), see also Proposition 3.2. Consequently, we obtain that \( \langle Q^{(n)}_\Psi v^*, v^* \rangle \to 0 \) as \( t_n \to t \) and we can conclude as in the proof of Corollary 2.8 in [31] that the integral process is continuous in \( p \)-th mean.
6. The Cauchy problem

In this section, we apply our previous results to consider stochastic evolution equations driven by cylindrical fractional Brownian motions of the form

\[
dY(t) = AY(t) \, dt + C \, dB(t), \quad t \in (0, T],
\]

\[
Y(0) = y_0.
\]

(6.1)

Here \(B\) is a cylindrical fBm in a separable Banach space \(U\), \(A\) is a generator of a strongly continuous semigroup \((S(t), t \geq 0)\) in a separable Banach space \(V\) and \(C\) is an operator in \(\mathcal{L}(U, V)\). The initial condition \(y_0\) is an element in \(V\).

The paths of a solution exhibit some kind of regularity, which is weaker than \(P\)-a.s. Bochner integrable paths:

**Definition 6.1.** A \(V\)-valued stochastic process \((X(t) : t \in [0, T])\) is called weakly Bochner regular if for every sequence \((H_n)_{n \in \mathbb{N}}\) of continuous functions \(H_n : [0, T] \to V^*\) it satisfies:

\[
\sup_{t \in [0, T]} \|H_n(t)\| \to 0 \Rightarrow \int_0^T |\langle X(t), H_n(t) \rangle|^2 \, dt \to 0 \quad P\text{-a.s. for } k \to \infty,
\]

for a subsequence \((H_{nk})_{k \in \mathbb{N}}\) of \((H_n)_{n \in \mathbb{N}}\).

**Definition 6.2.** A stochastic process \((Y(t) : t \in [0, T])\) in \(V\) is called a weak solution of (6.1) if it is weakly Bochner regular and for every \(v^* \in D(A^*)\) and \(t \in [0, T]\) we have \(P\)-a.s.,

\[
\langle Y(t), v^* \rangle = \langle y_0, v^* \rangle + \int_0^t \langle Y(s), A^*v^* \rangle \, ds + B(t)(C^*v^*).
\]

(6.2)

From a proper integration theory we can expect that if the convoluted semigroup \(S(t - \cdot)C\) is integrable for all \(t \in [0, T]\) then a weak solution of (6.1) exists and can be represented by the usual variation of constants formula. Recall that the space of integrands is

\[
\mathcal{I} := \{ \Phi : [0, T] \to \mathcal{L}(U, V) : i^*\Phi^*(\cdot)v^* \in \hat{\mathcal{M}} \text{ for all } v^* \in V^* \}.
\]

**Theorem 6.3.** Assume that \(S(\cdot)C\) is in \(\mathcal{I}\). Then the following are equivalent:

(a) the Cauchy problem (6.1) has a weak solution \(Y\);

(b) the mapping \(S(\cdot)C\) is stochastically integrable.

In this situation the solution \((Y(t) : t \in [0, T])\) can be represented by

\[
Y(t) = S(t)y_0 + \int_0^t S(t - s)C \, dB(s) \quad \text{for all } t \in [0, T].
\]

(6.3)

**Proof.** (b) \(\Rightarrow\) (a): Proposition 3.2 guarantees for each \(t \in [0, T]\) that the mapping \(\mathbb{1}_{[0, t]}S(t - \cdot)C\) is in \(\mathcal{I}\) and that there exists a constant \(c_t > 0\) such that

\[
\|\mathbb{1}_{[0, t]}i^*C^*S^*(t - \cdot)v^*\|_{\mathcal{M}} \leq c_t \|i^*C^*S^*(\cdot)v^*\|_{\mathcal{M}} \quad \text{for all } v^* \in V^*.
\]
Thus Corollary 5.6 guarantees that \(\mathbb{1}_{[0,t]} S(t-)C\) is stochastically integrable, which enables to define the stochastic integral
\[
X(t) := \int_0^t S(t-s)C \, dB(s) \quad \text{for all } t \in [0,T].
\]

It follows from representation (5.3) that the real valued stochastic process \((X(t),v^*) : t \in [0,T]\) is adapted for each \(v^* \in V^*\). Pettis’ measurability theorem implies that \(X := (X(t) : t \in [0,T])\) is adapted.

By linearity we can assume that \(y_0 = 0\). The stochastic Fubini theorem for real valued fBm implies for each \(v^* \in D(A^*)\) and \(t \in [0,T]\), that
\[
\int_0^t \langle X(s), A^*v^* \rangle \, ds = \sum_{k=1}^{\infty} \int_0^t \langle S(s-r) C ie_k, A^*v^* \rangle \, db_k(r) \, ds
\]
\[
= \sum_{k=1}^{\infty} \int_0^t \langle S(s-r) C ie_k, A^*v^* \rangle \, ds \, db_k(r)
\]
\[
= \sum_{k=1}^{\infty} \int_0^t \langle S(t-r) C ie_k - C ie_k, v^* \rangle \, db_k(r)
\]
\[
= \langle \int_0^t S(t-r)C \, dB(r), v^* \rangle - \sum_{k=1}^{\infty} (ie_k, C^*v^*)b_k(t)
\]
\[
= \langle X(t), v^* \rangle - B(t)(C^*v^*),
\]
which shows that the process \(X\) satisfies (6.2). In order to show that \(X\) is weakly Bochner regular define \(\Psi_t := \mathbb{1}_{[0,t]}(\cdot) S(t-)C\) for each \(t \in [0,T]\). Note that Proposition 3.2 guarantees that there exists a constant \(c_t > 0\) such that
\[
\|\Gamma_{\Psi_t} v^*\|_{L^2} = \|\mathbb{1}_{[0,t]}(\cdot) i^* C^* S^* (t-) v^*\|_{\mathcal{M}} \leq c_t \|i^* C^* S^* (\cdot) v^*\|_{\mathcal{M}}
\]
for every \(v^* \in V^*\). Since the derivation of the constant \(c_t\) in [26, Ch.13.3] shows that \(\sup_{t \in [0,T]} c_t < \infty\), the uniform boundedness principle implies that \(\sup_{t \in [0,T]} \|\Gamma_{\Psi_t} v^*\|_{V^* \to L^2} < \infty\). Thus for a sequence \((H_n)_{n \in \mathbb{N}}\) of continuous mappings \(H_n : [0,T] \to V^*\) we obtain
\[
E \left[ \left| \int_0^T \langle X(t), H_n(t) \rangle \, dt \right|^2 \right] \leq T \int_0^T E \left[ |Z_{\Psi_t} H_n(t)|^2 \right] \, dt
\]
\[
= T \int_0^T \|\Gamma_{\Psi_t} H_n(t)\|^2 \, dt
\]
\[
\leq T^2 \sup_{t \in [0,T]} \|\Gamma_{\Psi_t} v^*\|_{V^* \to L^2}^2 \sup_{t \in [0,T]} \|H_n(t)\|^2,
\]
which shows the weak Bochner regularity.

(a) \(\Rightarrow\) (b): by applying Itô’s formula for real valued fBm, see e.g. [5, Thm 6.3.1], one deduces for every continuously differentiable function \(f : [0,T] \to \mathbb{R}\) and real valued fBm \(b\)
\[
\int_0^T f'(s)b(s) \, ds = f(T)b(T) - \int_0^T f(s) \, db(s) \quad \text{P-a.s.},
\]
(6.4)
where the integral on the right-hand side can be understood as a Wiener integral, since \( f \) is deterministic. Let \( Y \) be a weak solution of (6.1) and denote by \( A^\circ \) the part of \( A^* \) in \( \mathcal{D}(A^*) \). Then \( \mathcal{D}(A^\circ) \) is a weak*-sequentially dense subspace of \( V^* \). From the integration by parts formula (6.4) it follows as in the proof of Theorem 7.1 in [31] that

\[
\langle Y(T), v^* \rangle = Z_\Psi v^* \quad \text{for all } v^* \in \mathcal{D}(A^\circ),
\]

where \( Z_\Psi \) denotes the cylindrical integral of \( \Psi := S(T - \cdot)C \). It remains to show that (6.5) holds for all \( v^* \in V^* \), for which we mainly follow the arguments of the proof of Theorem 2.3 in [31]. Observe that the random variable \( Y(T) \) is Gaussian since the right hand side in (6.5) is Gaussian for each \( v^* \in \mathcal{D}(A^\circ) \) and Gaussian distributions are closed under weak limits.

Let \( R : V^* \to V \) and \( Q : V^* \to V^{**} \) denote the covariance operators of \( Y(T) \) and \( Z_\Psi \), respectively. Since \( R \) is the covariance operator of a Gaussian measure there exists a Hilbert space \( H \) which is continuously embedded by a \( \gamma \)-radonifying mapping \( j : H \to V \) such that \( R = jj^* \). Equality (6.5) implies

\[
\langle Rw^*, v^* \rangle = \langle Qv^*, v^* \rangle \quad \text{for all } v^* \in \mathcal{D}(A^\circ).
\]

Let \( (v^*_n)_{n \in \mathbb{N}} \) be a sequence in \( \mathcal{D}(A^\circ) \) converging weakly* to \( v^* \) in \( V^* \). Thus

\[
\lim_{n \to \infty} j^*v^*_n = j^*v^* \quad \text{weakly in } H \quad \text{since } H \text{ is a Hilbert space and } j^* \text{ is weak* continuous.}
\]

As a consequence of the Hahn-Banach theorem one can construct a convex combination \( w^*_n \) of the \( v^*_n \) such that \( \lim_{n \to \infty} j^*w^*_n = j^*v^* \) strongly in \( H \) and \( \lim_{n \to \infty} w^*_n = v^* \) weakly* in \( V^* \). Since \( w^*_m - w^*_n \) is in \( \mathcal{D}(A^\circ) \) for all \( m, n \in \mathbb{N} \), inequality (6.6) implies

\[
\|i^*C^*S^*(T - \cdot)(w^*_m - w^*_n)\|_M^2 = \|K^*(i^*C^*S^*(T - \cdot)(w^*_m - w^*_n))\|_{L^2}^2 = \langle Q(w^*_m - w^*_n), w^*_m - w^*_n \rangle = \langle R(w^*_m - w^*_n), w^*_m - w^*_n \rangle = \|j^*(w^*_m - w^*_n)\|_H \to 0 \quad \text{as } m, n \to \infty.
\]

Thus, \( (i^*C^*S^*(T - \cdot)w^*_n)_{n \in \mathbb{N}} \) is a Cauchy sequence in \( M \) and therefore it converges to some \( g \in M \). By the same arguments as in the proof of Proposition 5.1 it follows that there is a subsequence such that \( \lim_{k \to \infty} i^*C^*S^*(T - s)w^*_{n_k} = g(s) \) for Lebesgue almost all \( s \in [0, T] \). On the other hand, the weak* convergence of \( (w^*_n)_{k \in \mathbb{N}} \) implies that \( \lim_{k \to \infty} \langle i^*C^*S^*(T - s)w^*_{n_k}, x \rangle = \langle i^*C^*S^*(T - s)v^*, x \rangle \) for all \( x \in X \) and \( s \in [0, T] \), which yields \( g = i^*C^*S^*(T - \cdot)v^* \). It follows that

\[
\langle Rw^*_{n_k}, w^*_{n_k} \rangle = \|i^*C^*S^*(t - \cdot)w^*_{n_k}\|_M^2 \to \|i^*C^*S^*(t - \cdot)v^*\|_M^2 = \langle Qv^*, v^* \rangle,
\]

as \( k \to \infty \). Therefore the covariance operators \( R \) and \( Q \) coincide on \( V^* \), which yields that the cylindrical distribution of \( Z_\Psi \) extends to a Radon measure.

If \( H > \frac{1}{2} \) then the first condition in Theorem 6.3, i.e. \( S(\cdot)C \in \mathcal{I} \), is satisfied for every strongly continuous semigroup as \( L^2([0, T]; X) \subseteq \mathcal{M} \). If \( H < \frac{1}{2} \) this condition is not obvious but an important case is covered by the following result.
exists a constant the same computations as in the proof of [22, Le.11.7] one derives that there

\[ (6.8) \]

can be estimated by

for all \( s \) embedded in \( M \) negative type, then the mapping \( S(\cdot)C \) is in \( \mathcal{I} \).

**Proposition 6.4.** Let \( H < \frac{1}{2} \). If \( (S(t), t \geq 0) \) is an analytic semigroup of negative type, then the mapping \( S(\cdot)C \) is in \( \mathcal{I} \).

**Proof.** Define for arbitrary \( f \in L^2([0,T];X) \) the function

\[ G_f(s) := b_H \left( \frac{f(s)}{(T-s)^{\frac{3}{2}-H}} + \left( \frac{1}{2} - H \right) s^{\frac{1}{2}-H} \int_s^T \frac{s^H-\frac{1}{2}f(s) - t^H-\frac{1}{2}f(t)}{(t-s)^{\frac{3}{2}-H}} dt \right) \]  

(6.7)

for all \( s \in (0,T] \). It follows from (3.8) that a function \( f \in L^2([0,T];X) \) is in \( \mathcal{M} \) if and only \( G_f \in L^2([0,T];X) \), in which case \( \|f\|_{\mathcal{M}} = \|G_f\|_{L^2} \). By the same computations as in the proof of [22, Le.11.7] one derives that there exists a constant \( c_1 > 0 \) such that

\[ \int_0^T \|G_f(s)\|^2 ds \leq c_1 \left( \int_0^T \frac{\|f(s)\|^2}{(T-s)^{1-2H}} ds + \int_0^T \frac{\|f(s)\|^2}{s^{1-2H}} ds \right) + \int_0^T \left( \int_s^T \frac{\|f(t) - f(s)\|}{(t-s)^{\frac{3}{2}-H}} dt \right)^2 ds \]  

(6.8)

We check that each term on the right hand side of (6.8) is finite for \( f = i^*C^*S^*(\cdot)v^*, v^* \in V^* \). The growth bound of the semigroup guarantees that there exist some constants \( \beta, c_2 > 0 \) such that

\[ \|S(s)\| \leq c_2 e^{-\beta s} \quad \text{for all } s \in [0,T]. \]

It is immediate that the first two integrals on the right hand side in (6.8) are finite since \( 1 - 2H < 1 \). In order to estimate the last term, recall that, as \( S \) is analytic, there exists for each \( \alpha \geq 0 \) a constant \( c_3 > 0 \) such that for every \( 0 < s \leq t \) we have

\[ \|S(t) - S(s)\| = \|(S(t-s) - \text{Id})S(s)\| \leq c_3 (t-s)\alpha s^{-\alpha} e^{-\beta s}. \]

Fix some \( \alpha \in (\frac{1}{2} - H, \frac{1}{2}) \). The third summand on the right hand side of (6.8) can be estimated by

\[ \int_0^T \left( \int_s^T \frac{\|i^*C^*S^*(t)v^* - i^*C^*S^*(s)v^*\|}{(t-s)^{\frac{3}{2}-H}} dt \right)^2 ds \]

\[ \leq (c_3 \|i\| \|C\| \|v^*\|)^2 \int_0^T \left( \int_s^T \frac{e^{-\beta s}}{s^\alpha} \frac{(t-s)\alpha}{(t-s)^{2-H}} dt \right)^2 ds \]

\[ \leq (c_3 \|i\| \|C\| \|v^*\|)^2 \left( \int_0^T \frac{e^{-2\beta s}}{s^{2\alpha}} ds \right) \left( \int_0^T \frac{1}{t^{2-H-\alpha}} dt \right)^2 \]

\[ \leq (c_3 \|i\| \|C\| \|v^*\|)^2 (2\beta)^{2\alpha-1} \Gamma(1-2\alpha) T^{2(H+\alpha-\frac{1}{2})}, \]

which completes the proof. \( \square \)

Another example of a semigroup satisfying \( S(\cdot)C \in \mathcal{I} \) is considered in Section 7. Further examples can be derived using the known fact that the space of Hölder continuous functions of index larger than \( \frac{1}{2} - H \) is continuously embedded in \( \mathcal{M} \).
Example 6.5. If \( V \) is a Hilbert space then a function \( \Psi \in \mathcal{I} \) is stochastically integrable if and only if \( \Gamma \Psi \) is Hilbert-Schmidt, according to Theorem 5.5. Thus, if \((f_k)_{k \in \mathbb{N}}\) denotes an orthonormal basis of \( V \), the function \( \Psi \) is stochastically integrable if and only if the adjoint operator \( \Gamma^* \Psi \) is Hilbert-Schmidt, that is

\[
\sum_{k=1}^{\infty} \| \Gamma^* f_k \|_{L_2}^2 = \sum_{k=1}^{\infty} \| K^*(i^* \Psi^*(\cdot)f_k) \|_{L_2}^2 = \sum_{k=1}^{\infty} \| i^* \Psi^*(\cdot)f_k \|_{\mathcal{M}}^2 < \infty.
\]

In the case \( H > \frac{1}{2} \) we obtain that there exists a weak solution of (6.1) if

\[
\sum_{k=1}^{\infty} \int_0^T \int_0^T \| i^* C^s S^t(s) f_k \| \| i^* C^s S^t(t) f_k \| |s-t|^{2H-2} \, ds \, dt < \infty.
\]

For the case \( H < \frac{1}{2} \) assume that the semigroup \((S(t), t \geq 0)\) is analytic and of negative type. A similar calculation as in the proof of Proposition 6.4 shows that if there exists a constant \( \alpha \in \left( \frac{1}{2} - H, \frac{1}{2} \right) \) such

\[
\int_0^T \frac{\| S(s)C \|_{HS}^2}{s^{2\alpha}} \, ds < \infty,
\]

then \( S(\cdot)C \) is stochastically integrable. Here \( \| \cdot \|_{HS} \) denotes the Hilbert-Schmidt norm of an operator \( U : L^2([0,T];X) \to V \).

7. Example: The stochastic heat equation

As an example we consider a self-adjoint generator \( A \) of a semigroup \((S(t), t \geq 0)\) in a separable Hilbert space \( V \) such that there exists an orthonormal basis \((e_k)_{k \in \mathbb{N}}\) of \( V \) satisfying \( Ae_k = -\lambda_k e_k \) for some \( \lambda_k > 0 \) for all \( k \in \mathbb{N} \) and \( \lambda_k \to \infty \) as \( k \to \infty \). Thus the semigroup satisfies

\[
S(t)e_k = e^{-\lambda_k t} e_k \quad \text{for all } t \geq 0 \text{ and } k \in \mathbb{N}.
\]

A specific instance is the Laplace operator with Dirichlet boundary conditions on \( V = L^2(D;\mathbb{R}) \) for a set \( D \in \mathcal{B}(\mathbb{R}^n) \). We assume that \( C = \text{Id} \) and we identify the dual space \( V^* \) with \( V \), i.e. we consider the Cauchy problem

\[
dY(t) = AY(t) \, dt + dB(t) \quad \text{for all } t \in [0,T].
\]

The system (7.1) is perturbed by a cylindrical fBm \( B \) in \( V \) which is independent along the orthonormal basis \((e_k)_{k \in \mathbb{N}}\) of eigenvectors \( e_k \) of \( A \), that is we consider the cylindrical fBm \((B(t) : t \geq 0)\) in \( V \) from Example 4.4:

\[
B(t)v = \sum_{k=1}^{\infty} (ie_k, v) b_k(t) \quad \text{for all } v \in V, t \geq 0,
\]

where \((b_k)_{k \in \mathbb{N}}\) is a sequence of independent, real valued standard fBms of Hurst parameter \( H \in (0,1) \) and the embedding \( i : V \to V \) is defined by

\[
iv = \sum_{k=1}^{\infty} q_k (e_k, v) e_k
\]

for a sequence \((q_k)_{k \in \mathbb{N}} \subseteq \mathbb{R} \) satisfying \( \sup_k |q_k| < \infty \). Note that in this case \( X = V \).
Theorem 7.1. Let $A$ be a self-adjoint generator satisfying the conditions described above. If

$$\sum_{k=1}^{\infty} \frac{q_k^2}{\lambda_k^{2H}} < \infty,$$

then equation (7.1) has a weak solution $(Y(t) : t \in [0,T])$ in $V$. The solution can be represented by the variation of constants formula (6.3).

Proof. Note that in this situation we have

$$i^*S^*(t)e_k = q_k e^{-\lambda_k t} e_k \quad \text{for each } k \in \mathbb{N} \text{ and } t \in [0,T]. \quad (7.2)$$

According to Theorem 5.5 and Theorem 6.3 we have to establish that $S(\cdot)$ is in $\mathcal{I}$ and the operator $\Gamma : L^2([0,T];V) \to V$ defined by

$$\langle \Gamma f, v \rangle = \int_0^T [K^*(i^*S^*(\cdot)v)(s), f(s)] \, ds \quad \text{for all } f \in L^2([0,T];V), \; v \in V$$

is $\gamma$-radonifying. Since $V$ is a separable Hilbert space, the operator $\Gamma$ is $\gamma$-radonifying if and only if it is Hilbert-Schmidt.

If $H > \frac{1}{2}$ then $S(\cdot)$ is in $\mathcal{I}$ and equality (7.2) yields for each $k \in \mathbb{N}$

$$\|i^*S^*(\cdot)e_k\|^2_{\mathcal{M}} = H(2H - 1) \int_0^T \int_0^T \|i^*S^*(t)e_k\| \|i^*S^*(s)e_k\| |s - t|^{2H-2} \, ds \, dt$$

$$= H(2H - 1)q_k^2 \int_0^T \int_0^T e^{-\lambda_k t} e^{-\lambda_k s} |s - t|^{2H-2} \, ds \, dt. \quad (7.3)$$

The iterated integral can be estimated by

$$\int_0^T e^{-\lambda_k t} \int_0^T e^{-\lambda_k s} |s - t|^{2H-2} \, ds \, dt = 2 \int_0^T e^{-\lambda_k t} \int_0^t e^{-\lambda_k s} |s - t|^{2H-2} \, ds \, dt$$

$$= 2 \int_0^T e^{-2\lambda_k t} \int_0^t e^{\lambda_k s} s^{2H-2} \, ds \, dt$$

$$= 2 \int_0^T e^{\lambda_k s} s^{2H-2} \int_s^T e^{-2\lambda_k t} \, dt \, ds$$

$$\leq \frac{1}{\lambda_k} \int_0^T e^{-\lambda_k s} s^{2H-2} \, ds$$

$$\leq \frac{1}{\lambda_k^{2H}} \Gamma(2H - 1). \quad (7.4)$$

Since inequality (3.11) guarantees that there exists a constant $c > 0$ such that

$$\sum_{k=1}^{\infty} \|e_k\|^2_{L^2} = \sum_{k=1}^{\infty} \|K^*(i^*S^*(\cdot)e_k)\|^2_{\mathcal{M}} \leq c \sum_{k=1}^{\infty} \|i^*S^*(\cdot)e_k\|^2_{\mathcal{M}},$$

we can conclude from (7.3) and (7.4) that $\Gamma^*$ and thus $\Gamma$ are Hilbert-Schmidt operators.

If $H < \frac{1}{2}$ Proposition 6.4 guarantees that $S(\cdot)$ is in $\mathcal{I}$. As in the proof of Proposition 6.4 it follows that there exists a constant $c_1 > 0$ such that for
all \( k \in \mathbb{N} \)

\[
\| K^s(i^*S^s(\cdot)e_k) \|_{L^2}^2 \leq c_1 \left( \int_0^T \left\| i^*S^s(s) e_k \right\|^2 \frac{ds}{(T - s)^{1 - 2H}} + \int_0^T \left\| i^*S^s(s) e_k \right\|^2 \frac{ds}{s^{1 - 2H}} \right.
\]
\[
+ \left. \int_0^T \left( \int_s^T \left\| i^*S^s(t) e_k - i^*S^s(s) e_k \right\| \frac{dt}{(t - s)^{\frac{3}{2} - H}} \right)^2 \frac{ds}{s^\lambda} \right) \quad (7.5)
\]

Equality (7.2) implies for the first integral the estimate

\[
\int_0^T \left\| i^*S^s(s) e_k \right\|^2 \frac{ds}{(T - s)^{1 - 2H}} \leq q_k^2 \int_0^T e^{-2\lambda s} \frac{ds}{(T - s)^{1 - 2H}} = \frac{q_k^2}{(2\lambda T)^{2H}} \int_0^{2\lambda T} e^{-s} \frac{ds}{(2\lambda T - s)^{1 - 2H}} \leq \frac{q_k^2}{(2\lambda T)^{2H}} \left( 1 + \frac{1}{2H} \right). \quad (7.6)
\]

Here, the estimate of the integral follows from the fact that if \( 2\lambda k T \leq 1 \) then

\[
\int_0^{2\lambda k T} e^{-s} \frac{ds}{(2\lambda k T - s)^{1 - 2H}} \leq \int_0^{2\lambda k T} \frac{1}{(2\lambda k T - s)^{1 - 2H}} ds \leq \frac{1}{2H},
\]

and if \( 2\lambda k T > 1 \) then

\[
\int_0^{2\lambda k T} e^{-s} \frac{ds}{(2\lambda k T - s)^{1 - 2H}} \leq \int_0^{2\lambda k T - 1} e^{-s} ds + \int_{2\lambda k T - 1}^{2\lambda k T} (2\lambda k T - s)^{2H - 1} ds \leq 1 + \frac{1}{2H}.
\]

The second integral in (7.5) can be bounded by

\[
\int_0^T \left\| i^*S^s(s) e_k \right\|^2 \frac{ds}{s^{1 - 2H}} \leq q_k^2 \int_0^T e^{-2\lambda s} \frac{ds}{s^{1 - 2H}} \leq \Gamma(2H) \frac{q_k^2}{(2\lambda k)^{2H}}. \quad (7.7)
\]

Another application of equality (7.2) yields for the third term in (7.5)

\[
\int_0^T \left( \int_s^T \left\| i^*S^s(t) e_k - i^*S^s(s) e_k \right\| \frac{dt}{(t - s)^{\frac{3}{2} - H}} \right)^2 \frac{ds}{s^\lambda} = q_k^2 \int_0^T \left( \int_s^T \frac{e^{-\lambda t} - e^{-\lambda s}}{(t - s)^{\frac{3}{2} - H}} dt \right)^2 \frac{ds}{s^\lambda} = q_k^2 \int_0^T e^{-2\lambda s} \left( \int_0^{T - s} 1 - e^{-\lambda t} \frac{dt}{t^{\frac{3}{2} - H}} \right)^2 ds. \quad (7.8)
\]
Applying the changes of variables \( \lambda_k s = x \) and \( \lambda_k t = y \) yields

\[
\int_0^T e^{-2\lambda_k s} \left( \int_0^{T-s} \frac{1 - e^{-\lambda_k t}}{t^{\frac{3}{2} - H}} \, dt \right)^2 \, ds = \frac{1}{\lambda_k^{2H}} \int_0^{\lambda_k T} e^{-2x} \left( \int_0^{\lambda_k T-x} \frac{1 - e^{-y}}{y^{\frac{3}{2} - H}} \, dy \right)^2 \, dx
\]

\[
= \frac{1}{\lambda_k^{2H}} \int_0^{\lambda_k T} e^{-2(\lambda_k T - x)} \left( \int_0^{x} \frac{1 - e^{-y}}{y^{\frac{3}{2} - H}} \, dy \right)^2 \, dx
\]

\[
\leq \frac{1}{\lambda_k^{2H}} c_2, \hspace{1cm} \text{(7.9)}
\]

where \( c_2 > 0 \) denotes a constant only depending on \( H \) but not on \( \lambda_k \). The finiteness of the constant \( c_2 \) and its independence of \( \lambda_k \) follow from the following three estimates:

\[
\int_0^1 e^{-2(\lambda_k T - x)} \left( \int_0^{x} \frac{1 - e^{-y}}{y^{\frac{3}{2} - H}} \, dy \right)^2 \, dx \leq \int_0^1 \left( \int_0^{1} \frac{1 - e^{-y}}{y^{\frac{3}{2} - H}} \, dy \right)^2 \, dx,
\]

\[
\int_1^\lambda_k T e^{-2(\lambda_k T - x)} \left( \int_0^{1} \frac{1 - e^{-y}}{y^{\frac{3}{2} - H}} \, dy \right)^2 \, dx \leq \frac{1}{(H + \frac{1}{2})^2} \int_1^{\lambda_k T} e^{-2(\lambda_k T - x)} \, dx,
\]

\[
\int_1^{\lambda_k T} e^{-2(\lambda_k T - x)} \left( \int_1^{\frac{1}{1 - H}} \frac{1 - e^{-y}}{y^{\frac{3}{2} - H}} \, dy \right)^2 \, dx \leq \frac{1}{(H - \frac{1}{2})^2} \int_1^{\lambda_k T} e^{-2(\lambda_k T - x)} x^{2H-1} \, dx
\]

\[
\leq \frac{1}{(H - \frac{1}{2})^2} \int_1^{\lambda_k T} e^{-2(\lambda_k T - x)} \, dx.
\]

By applying the estimates (7.6)–(7.9) to (7.5), it follows that there exists a constant \( c_3 > 0 \) such that

\[
\left\| \Gamma^* e_k \right\|_{L^2}^2 = \left\| K^* (i^* S^* (\cdot) e_k) \right\|_{L^2}^2 \leq c_3 \frac{q_k^2}{\lambda_k^{2H}} \hspace{1cm} \text{for all } k \in \mathbb{N}.
\]

As before we can conclude that \( \Gamma \) is Hilbert-Schmidt. \( \Box \)

Consider now the special case of the heat equation with Dirichlet boundary conditions driven by a cylindrical fractional noise with independent components, that is with \( Q = \text{Id} \). In this case \( q_k \equiv 1 \) and the eigenvalues of the Laplacian behave like \( \lambda_k \sim k^{2/n} \) so that the condition for the existence of a weak solution becomes the well known \( n/4 < H < 1 \). This result is in line with the literature, see for example [7, 12, 17].

\section*{References}

[1] E. Alòs, O. Mazet, and D. Nualart. Stochastic calculus with respect to Gaussian processes. \textit{Ann. Probab.}, 29(2):766–801, 2001.

[2] D. Applebaum and M. Riedle. Cylindrical Lévy processes in Banach spaces. \textit{Proc. Lond. Math. Soc.}, 101:697–726, 2010.

[3] A. Badrikian. Séminaire sur les fonctions aléatoires linéaires et les mesures cylindriques. Lecture Notes in Mathematics 139. Berlin: Springer, 1970.

[4] R. M. Balan. \( L_p \)-theory for the stochastic heat equation with infinite-dimensional fractional noise. \textit{ESAIM Probab. Stat.}, 15:110–138, 2011.
[5] F. Biagini, Y. Hu, B. Øksendal, and T. Zhang. Stochastic calculus for fractional Brownian motion and applications. Springer, London, 2008.
[6] Z. Bavna. On stochastic convolution in Banach spaces and applications. Stochastics Stochastics Rep., 61(3-4):245–295, 1997.
[7] Z. Bavna, J. M. A. M. van Neerven, and D. Salo. Stochastic evolution equations driven by a Liouville fractional Brownian motion. Czech. Math. J., 62:1–27, 2012.
[8] G. Da Prato and J. Zabczyk. Stochastic equations in infinite dimensions. Cambridge University Press, Cambridge, 1992.
[9] T. E. Duncan, J. Jakubowski, and B. Pasik-Duncan. Stochastic integration for fractional Brownian motion in a Hilbert space. Stoch. Dyn., 6:53–75, 2006.
[10] T. E. Duncan, B. Maslowski, and B. Pasik-Duncan. Stochastic equations in Hilbert space with a multiplicative fractional Gaussian noise. Stochastic Process. Appl., 115(8):1357–1383, 2005.
[11] T. E. Duncan, B. Maslowski, and B. Pasik-Duncan. Semilinear stochastic equations in a Hilbert space with a fractional Brownian motion. SIAM J. Math. Anal., 40:2286–2315, 2009.
[12] T. E. Duncan, B. Pasik-Duncan, and B. Maslowski. Fractional Brownian motion and stochastic equations in Hilbert spaces. Stoch. Dyn., 2(2):225–250, 2002.
[13] W. Grecksch and V. V. Anh. A parabolic stochastic differential equation with fractional Brownian motion input. Statist. Probab. Lett., 41(4):337–346, 1999.
[14] M. Gubinelli, A. Lejay, and S. Tindel. Young integrals and SPDEs. Potential Anal., 25:307–326, 2006.
[15] M. Hinz, M. and Zähle. Gradient-type noises. I. Partial and hybrid integrals. Complex Var. Elliptic Equ., 54(6):561–583, 2009.
[16] G. Kallianpur and J. Xiong. Stochastic differential equations in infinite dimensional spaces. Hayward, CA: IMS, Inst. of Math. Statistics, 1996.
[17] B. Maslowski and D. Nualart. Evolution equations driven by a fractional Brownian motion. J. Funct. Anal., 202(1):277–305, 2003.
[18] M. Métivier and J. Pellaumail. Stochastic integration. Academic Press, New York, 1980.
[19] Y. Mishura. Stochastic calculus for fractional Brownian motion and related processes, volume 1929 of Lecture Notes in Mathematics. Springer, Berlin, 2008.
[20] D. Nualart. Fractional Brownian motion: stochastic calculus and applications. In International Congress of Mathematicians. Vol. III, pages 1541–1562. Eur. Math. Soc., Zürich, 2006.
[21] D. Nualart. The Malliavin calculus and related topics. Springer, Berlin, 2006.
[22] B. Pasik-Duncan, T. E. Duncan, and B. Maslowski. Linear stochastic equations in a Hilbert space with a fractional Brownian motion. Stochastic processes, optimization, and control theory: applications in financial engineering, queuing networks, and manufacturing systems, Berlin: Springer. Internat. Ser. Oper. Res. Management Sci. 94, 201-221, 2006.
[23] V. Pipiras and M. S. Taqqu. Integration questions related to fractional Brownian motion. Probab. Theory Related Fields, 118:251–291, 2000.
[24] M. Riedle. Cylindrical Wiener processes. Séminaire de Probabilités XLIII, Berlin: Springer. Lecture Notes in Mathematics 2006, 191-214, 2011.
[25] M. Riedle and O. van Gaans. Stochastic integration for Lévy processes with values in Banach spaces. Stochastic Process. Appl., 119(6):1952–1974, 2009.
[26] S. G. Samko, A. A. Kilbas, and O. I. Marichev. Fractional integrals and derivatives. Gordon and Breach Science Publishers, New York, NY, 1993.
[27] L. Schwartz. Radon measures on arbitrary topological spaces and cylindrical measures. London: Oxford University Press, London, 1973.
[28] S. Tindel, C. A. Tudor, and F. Viens. Stochastic evolution equations with fractional Brownian motion. Probab. Theory Related Fields, 127(2):186–204, 2003.
[29] N. N. Vakhania, V. I. Tarieladze, and S. A. Chobanyan. Probability distributions on Banach spaces. D. Reidel Publishing Company, Dordrecht, 1987.
[30] J. M. A. M. van Neerven, M. C. Veraar, and L. Weis. Stochastic integration in UMD Banach spaces. Ann. Probab., 35(4):1438–1478, 2007.
[31] J.M.A.M. van Neerven and L. Weis. Stochastic integration of functions with values in a Banach space. *Stud. Math.*, 166:131–170, 2005.

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