# SCALING LIMIT OF A LONG-RANGE RANDOM WALK IN TIME-CORRELATED RANDOM ENVIRONMENT

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ABSTRACT. This paper concerns a long-range random walk in random environment in dimension 1 + 1, where the environmental disorder is independent in space but has long-range correlations in time. We prove that two types of rescaled partition functions converge weakly to the Stratonovich solution and the Itô-Skorohod solution respectively of a fractional stochastic heat equation with multiplicative Gaussian noise which is white in space and colored in time.

# 1. INTRODUCTION

The model of directed polymer in random environment was first introduced by Huse and Henley [24] in the study of the Ising model and became a canonical model for disordered systems (see e.g. the lecture notes by Comets [11]). In recent years, it has attracted much attention in particular due to its intimate connections to the stochastic heat equation (parabolic Anderson model), the

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stochastic Burgers equation, and the Kardar-Parisi-Zhang (KPZ) equation and its universality class (see [11] and [12] for a review).

For a simple symmetric random walk in i.i.d. random environment on  $\mathbb{N} \times \mathbb{Z}$ , Alberts et al. [1] proved that the rescaled partition function converges weakly to the *Itô-Skorohod* solution of the stochastic heat equation (SHE) with multiplicative space-time Gaussian white noise. This result was extended by Caravenna et al. [8] to a long-range random walk in i.i.d. random environment and to other disordered models (disordered pinning model and random field Ising model). Later on, Rang [37] proved that, for a simple random walk in random environment which is white in time but correlated in space, the rescaled partition function converges weakly to the *Itô-Skorohod* solution of SHE with Gaussian noise white in time and colored in space; this result was extended to long-range random walks by Chen and Gao [9]; if the random environment is given by the occupation field of a Poisson system of independent random walks on  $\mathbb{Z}$  which is now correlated in both time and space, Shen et al. [41] showed that the scaling limit is the *Stratonovich* solution of SHE with space-time colored Gaussian noise whose covariance coincides with the heat kernel. It is suggested by the results and methodologies in the above-mentioned papers that the *temporal independence* of the random environment plays a critical role when identifying the scaling limits for partition functions, which will be further discussed in Section 1.3.

In this paper, we aim to study the scaling limit of the partition function for a long-range random walk in random environment on  $\mathbb{N} \times \mathbb{Z}$  where the disorder is independent in space but has long-range correlations in time.

We remark that the model of directed polymer in space-time correlated random environment whose covariance has a power-law decay was first considered in physics literature by Medina et al. [32] where the Burgers equation with colored noise was studied and then applied to analyse directed polymer and interface growth. We also would like to mention that Rovira and Tindel [39] introduced Brownian polymer in a centered Gaussian field that is white in time and correlated in space on  $\mathbb{R}_+ \times \mathbb{R}$  and studied the asymptotic behavior of the partition function; in the two subsequent papers, Bezerra et al. [6] obtained the superdiffusivity and Lacoin [30] investigated the effect of strong spatial correlation. Finally, we remind the reader that the model of long-range directed polymer has been studied by Comets [10], and more recently by Wei [44].

1.1. Notations and known results. For the reader's convenience, we collect the mathematical notations that will be used throughout this article.

**Notations 1.1.** Let  $\mathbb{N}$  denote the set of natural numbers without 0, i.e.,  $\mathbb{N} \stackrel{\text{def}}{=} \{1, 2, \ldots\}$ ; for  $N \in \mathbb{N}$ ,  $[\![N]\!] \stackrel{\text{def}}{=} \{1, 2, \ldots, N\}$ ; for  $a \in \mathbb{R}$ , [a] means the greatest integer that is not greater than  $a; \|\cdot\|$  is used for the Euclidean norm; let  $\mathbf{k} \stackrel{\text{def}}{=} (k_1, \ldots, k_d), \mathbf{l} \stackrel{\text{def}}{=} (l_1, \ldots, l_d), \mathbf{x} \stackrel{\text{def}}{=} (x_1, \ldots, x_d), \mathbf{y} \stackrel{\text{def}}{=} (y_1, \ldots, y_d)$  etc stand for vectors in  $\mathbb{Z}^d$  or  $\mathbb{R}^d$  depending on the context; we use C to denote a generic positive constant that may change from line to line; we say  $f(x) \leq g(x)$  if  $f(x) \leq Cg(x)$  for all x; we write  $f(x) \sim g(x)$  (as  $x \to \infty$ ), if  $\lim_{x \to \infty} f(x)/g(x) = 1$ . We use  $\stackrel{d}{\to}$  to denote the convergence in distribution (also called weak convergence) for random variables/vectors. For a random variable X,  $\|X\|_{L^p} = (\mathbb{E}[|X|^p])^{1/p}$  for  $p \geq 1$ .

Let  $S = \{S_i, i \in \mathbb{N}_0\}$  be a random walk in  $\mathbb{Z}$  and  $\omega = \{\omega(i, k), (i, k) \in \mathbb{N} \times \mathbb{Z}\}$  be a family of random variables independent of S serving as the random environment (disorder). We shall use  $\mathbb{P}_S$ and  $\mathbb{E}_S$  (resp.  $\mathbb{P}_{\omega}$  and  $\mathbb{E}_{\omega}$ ) to denote the probability and expectation in the probability space of S(resp.  $\omega$ ), respectively. The probability and expectation in the product probability space of  $(S, \omega)$ is denoted by  $\mathbb{P}$  and  $\mathbb{E}$ , respectively. Given  $N \in \mathbb{N}$  and  $k \in \mathbb{Z}$ , let  $S^{(N+1,k)} = \{S_i^{(N+1,k)}, i \in [N+1]\}$  be a *backward* random walk in  $\mathbb{Z}$  with  $S_{N+1}^{(N+1,k)} = k$ , and the *partition function* is defined by

(1.1) 
$$Z_{\omega}^{(N)}(\beta,k) \stackrel{\text{def}}{=} \sum_{S} e^{\beta \sum_{i=1}^{N} \omega \left(i, S_{i}^{(N+1,k)}\right)} \mathbb{P}(S) = \mathbb{E}_{S} \left[ e^{\beta \sum_{i=1}^{N} \omega \left(i, S_{i}^{(N+1,k)}\right)} \right]$$

where  $\beta = 1/T > 0$  is the inverse temperature. We stress that the random walk is backward on the time interval [1, N + 1] in the sense that  $S_{N+1}^{(N+1,k)} = k$  while there is no restriction on  $S_1^{(N+1,k)}$ . Hence,  $Z_{\omega}^{(N)}(\beta, k)$  given by (1.1) is indeed a point-to-line partition function, and it corresponds to directly a discrete version of the solution to stochastic heat equation (see Proposition 2.10), which facilitates our calculations. Throughout the rest of the article, we shall omit the superscript (N+1,k) for the backward random walk to simplify the notation.

When S is a simple symmetric random walk with i.i.d. increments and  $\omega(i, k)$  are i.i.d. random variables with

$$\lambda(\beta) \stackrel{\text{def}}{=} \log \mathbb{E}[\mathrm{e}^{\beta\omega(i,k)}] < \infty$$

for some  $\beta$  sufficiently small, Alberts et al. [1] introduced the so-called *intermediate disorder regime*. More precisely, if  $\beta$  is scaled in the way

$$\beta \to \hat{\beta}_N \stackrel{\text{def}}{=} \beta N^{-\frac{1}{4}},$$

one has the following weak convergence

$$e^{-N\lambda(\hat{\beta}_N)}Z^{(N)}_{\omega}(\hat{\beta}_N, N^{1/2}x) \stackrel{d}{\longrightarrow} u(1,x) \text{ as } N \to \infty_{\mathbb{R}}$$

where u(t, x) is the Itô-Skorohod solution of the equation

(1.2) 
$$\begin{cases} \frac{\partial u(t,x)}{\partial t} = \frac{1}{2}\Delta u(t,x) + \sqrt{2}\beta u(t,x)\dot{W}(t,x), \ t > 0, x \in \mathbb{R}, \\ u(0,x) = 1. \end{cases}$$

Here,  $\dot{W}$  is space-time Gaussian white noise and Itô-Skorohod solution means that the product in  $u(t,x)\dot{W}(t,x)$  is a Wick product, or equivalently the associated stochastic integral is an Itô-Skorohod integral.

Under the same setting except that the random walk S may be long-range (i.e., the increments of S now lie in the domain of attraction of a  $\rho$ -stable law with  $\rho \in (1, 2]$ ), the result of [1] was extended in Caravenna et al. [8] where a unified framework based on the Lindeberg principle for polynomial chaos expansion was developed to study the scaling limits for pinning model, directed polymer model, and Ising model. Still using the setting of [1] but assuming that the disorder  $\{\omega(i,k), (i,k) \in \mathbb{N} \times \mathbb{Z}\}$  is correlated in space (but still independent in time), Rang [37] and Chen-Gao [9] obtained the weak convergence of the rescaled partition function to the Itô-Skorohod solution of (1.2) with the Gaussian noise  $\dot{W}(t, x)$  being colored in space (still white in time).

1.2. Model description. Motivated by the above-mentioned works, we aim to study the weak convergence of the rescaled partition function for a long-range random walk  $\{S_n, n \in \mathbb{N}\}$  in random environment on  $\mathbb{N} \times \mathbb{Z}$ , where the disorder  $\{\omega(i,k), i \in \mathbb{N}, k \in \mathbb{Z}\}$  is correlated in time and independent in space.

In our model, the directed polymer (random walk) is given by  $S_n = \sum_{i=0}^n Y_i$ , where  $\{Y_i\}_{i \in \mathbb{N}_0}$  are independent and identically distributed random variables with mean zero which have a 1-lattice

distribution belonging to the domain of attraction of a  $\rho$ -stable distribution with density function denoted by  $\mathfrak{g}_{\rho}(\cdot)$ . We assume  $1 < \rho \leq 2$  and

(1.3) 
$$\begin{cases} \mathbb{P}(Y_i = k) \lesssim |k|^{-1-\rho} \text{ for } k \in \mathbb{Z} \setminus \{0\}, & \text{ if } \rho \in (1,2), \\ \mathbb{E}[Y_i] = 0 \text{ and } \mathbb{E}[Y_i^2] = 1, & \text{ if } \rho = 2. \end{cases}$$

Let  $\psi(u) \stackrel{\text{def}}{=} \mathbb{E}[e^{iuY_i}]$  be the characteristic function of  $Y_i$ . Then the 1-lattice distribution of  $Y_i$  implies that  $\psi(u)$  is periodic with period  $2\pi$ , and furthermore,  $|\psi(u)| < 1$  for all  $u \in [-\pi, \pi] \setminus \{0\}$  (see, e.g., [25, Theorem 1.4.2]); this property of  $\psi(u)$  usually makes analysis easier. The assumption of 1-lattice distribution on  $Y_i$  also yields that the random walk  $S_n = \sum_{i=1}^n Y_i$  has period 1 and hence is *aperiodic*. We remind the reader that a simple symmetric random walk has period 2.

Denote by  $P_n(k) = \mathbb{P}(S_n = k)$  for  $n \in \mathbb{N}, k \in \mathbb{Z}$  the probability of S being at k at time n. Then by the local limit theorem for  $\rho$ -stable distribution (see [40, Theorem 6.1]), the convergence

(1.4) 
$$n^{1/\rho} P_n(k) - \mathfrak{g}_\rho(k/n^{1/\rho}) \to 0, \text{ as } n \to \infty$$

holds uniformly in k. Noting that  $\mathfrak{g}_{\rho}$  is a bounded function, (1.4) yields

(1.5) 
$$P_n(k) \lesssim n^{-1/\rho}, \text{ for } n \in \mathbb{N}, k \in \mathbb{Z}.$$

Also under the condition of (1.3), we have a local deviation estimation (see, e.g., [5, Theorem 2.6]):

(1.6) 
$$P_n(k) \lesssim n|k|^{-1-\rho}, \text{ for } n \in \mathbb{N}, k \in \mathbb{Z}.$$

This together with (1.5), we get

(1.7) 
$$P_n(k) \lesssim \left( n|k|^{-1-\rho} \right) \wedge n^{-1/\rho} = n^{-1/\rho} \left( |n^{-1/\rho}k|^{-1-\rho} \wedge 1 \right), \text{ for } n \in \mathbb{N}, k \in \mathbb{Z},$$

where  $a \wedge b := \min\{a, b\}$  for  $a, b \in \mathbb{R}$ .

It is well known that

$$\frac{S_n}{n^{1/\rho}} \stackrel{d}{\longrightarrow} \xi, \text{ as } n \to \infty,$$

where  $\xi$  has the symmetric  $\rho$ -stable distribution with characteristic function  $\exp\{-c_{\rho}|\eta|^{\rho}\}$  for some  $c_{\rho} > 0$ . Letting  $\mathfrak{g}_{\rho}(t, x)$  be the density function of the corresponding  $\rho$ -stable process, we have

$$\int_{\mathbb{R}} e^{i\eta x} \mathfrak{g}_{\rho}(t, x) dx = e^{-c_{\rho}t|\eta|^{\rho}}$$

Throughout the rest of the paper we will omit the subscript  $\rho$  and use  $\mathfrak{g}(t, x)$  exclusively to denote the density function of the  $\rho$ -stable process X. Note that  $\mathfrak{g}$  has the following scaling property

(1.8) 
$$\mathfrak{g}(t,x) = t^{-\frac{1}{\rho}} \mathfrak{g}(1,t^{-\frac{1}{\rho}}x),$$

and the upper bound in parallel with (1.7)

(1.9) 
$$\mathfrak{g}(t,x) \lesssim (t|x|^{-1-\rho}) \wedge t^{-1/\rho} = t^{-1/\rho} \Big( |t^{-1/\rho}x|^{-1-\rho} \wedge 1 \Big), \quad t \in \mathbb{R}_+, \ x \in \mathbb{R}.$$

We assume that the disorder in the environment is given by a family of Gaussian random variables  $\{\omega(n,k), n \in \mathbb{N}, k \in \mathbb{Z}\}$  with mean zero and covariance

(1.10) 
$$\mathbb{E}[\omega(n,k)\omega(n',k')] = \gamma(n-n')\delta_{kk'},$$

where  $\delta_{kk'}$  is the Kronecker delta function, i.e.,  $\delta_{kk'} = 1$  if k = k' and  $\delta_{kk'} = 0$  otherwise, and  $\gamma(n)$  has a power law decay:

(1.11) 
$$\gamma(n) \lesssim |n|^{2H-2} \wedge 1, \text{ for } n \in \mathbb{Z},$$

where  $H \in (1/2, 1]$  denotes a fixed constant throughout this paper. We further assume that for all  $t \in \mathbb{R} \setminus \{0\}$ ,

(1.12) 
$$\lim_{N \to \infty} N^{2-2H} \gamma([Nt]) = |t|^{2H-2}.$$

1.3. Main result, strategy and discussions. Consider the stochastic fractional heat equation on  $\mathbb R$ 

(1.13) 
$$\begin{cases} \frac{\partial u(t,x)}{\partial t} = -c_{\rho}(-\Delta)^{\frac{\rho}{2}}u(t,x) + \beta u(t,x)\dot{W}(t,x),\\ u(0,x) = 1, \end{cases}$$

where  $\rho \in (1, 2]$  and W(t, x) is Gaussian noise with covariance function given by

(1.14) 
$$\mathbb{E}\left[\dot{W}(t,x)\dot{W}(s,y)\right] = K(t-s,x-y) \stackrel{\text{def}}{=} |t-s|^{2H-2}\delta(x-y),$$

with  $H \in (1/2, 1]$  and  $\delta(\cdot)$  being the Dirac delta function. In particular,  $\dot{W}$  is a spatial white noise independent of time if H = 1.

We consider two types of solutions of (1.13)—the *Stratonovich solution* if the product  $u(t, x)\dot{W}(t, x)$  is an ordinary product (i.e., the associated stochastic integral is a Stratonovich integral) and the *Itô-Skorohod solution* if  $u(t, x)\dot{W}(t, x)$  is a Wick product (i.e., the associated stochastic integral is a Skorohod integral). See Section 2.1 for details.

Now we are ready to present our main results. For the first type of partition function  $Z_{\omega}^{(N)}$  which is given in (1.1), we have the following result.

**Theorem 1.1.** Let the random walk S and the disorder  $\omega$  be given as in Section 1.2. Let H and  $\rho$  be parameters satisfying

(1.15) 
$$H \in (1/2, 1] \text{ and } \theta \stackrel{\text{def}}{=} H - \frac{1}{2\rho} > \frac{1}{2}$$

Consider the rescaled partition function  $Z^{(N)}_{\omega}(\hat{\beta}_N, N^{1/\rho}x_0)$  given in (1.1) under the scaling

$$\beta \to \hat{\beta}_N \stackrel{\text{def}}{=} \beta N^{-\theta}.$$

Let u(t, x) be the Stratonovich solution of (1.13). Then we have

$$Z^{(N)}_{\omega}(\hat{\beta}_N, N^{1/\rho}x_0) \xrightarrow{d} u(1, x_0), \text{ as } N \to \infty.$$

In comparison with (1.1), another type of point-to-line partition function is given by

(1.16) 
$$\tilde{Z}_{\omega}^{(N)}(\beta,k) \stackrel{\text{def}}{=} \mathbb{E}_{S}\left[\mathrm{e}^{\beta\sum_{i=1}^{N}\omega(i,S_{i})-\frac{\beta^{2}}{2}\sum_{i,j=1}^{N}\gamma(i-j)\mathbf{1}_{\{S_{i}=S_{j}\}}}\right],$$

where  $S = S^{(N+1,k)} = \{S_i, i \in [N+1]\}$  is a backward random walk with  $S_{N+1} = k$ . The extra term in the exponential is half of the variance of  $\beta \sum_{i=1}^{N} \omega(i, S_i)$  conditional on the random walk

S. Indeed, we have

(1.17)
$$\mathbb{E}_{\omega}\left[\left(\sum_{i=1}^{N}\omega(i,S_{i})\right)^{2}\right] = \mathbb{E}_{\omega}\left[\sum_{i=1}^{N}\sum_{k\in\mathbb{Z}}\omega(i,k)\mathbf{1}_{\{S_{i}=k\}}\sum_{j=1}^{N}\sum_{l\in\mathbb{Z}}\omega(j,l)\mathbf{1}_{\{S_{j}=l\}}\right]$$
$$=\sum_{i,j=1}^{N}\sum_{k,l\in\mathbb{Z}}\gamma(i-j)\delta_{kl}\mathbf{1}_{\{S_{i}=k\}}\mathbf{1}_{\{S_{j}=l\}}$$
$$=\sum_{i,j=1}^{N}\gamma(i-j)\mathbf{1}_{\{S_{i}=S_{j}\}},$$

which can be viewed as a weighted intersection local time of S.

For the scaling limit of  $\tilde{Z}_{\omega}^{(N)}$  given in (1.16), we have the following result in parallel with Theorem 1.1.

**Theorem 1.2.** Let the random walk S and the disorder  $\omega$  be given as in Section 1.2. Let H and  $\rho$  be parameters satisfying

(1.18) 
$$H \in (1/2, 1], \ \rho \in (1, 2], \ and \ \theta = H - \frac{1}{2\rho} > 0.$$

Consider the rescaled partition function  $\tilde{Z}^{(N)}_{\omega}(\hat{\beta}_N, N^{1/\rho}x_0)$  given in (1.16) under the scaling

$$\beta \to \hat{\beta}_N \stackrel{\text{def}}{=} \beta N^{-\theta}$$

Let  $\tilde{u}(t,x)$  be the Itô-Skorohod solution of (1.13). Then we have

$$\tilde{Z}^{(N)}_{\omega}(\hat{\beta}_N, N^{1/\rho}x_0) \stackrel{d}{\longrightarrow} \tilde{u}(1, x_0), \ as \ N \to \infty.$$

**Remark 1.2.** In Theorem 1.2, the condition  $\rho > 1$  in (1.18) arises to guarantee the existence and uniqueness of the Itô-Skorohod solution of (1.13) (see Remark 2.13 in Section 2.3). In Theorem 1.1, the condition (1.15) also implies  $\rho > \frac{1}{2H-1} \ge 1$ . This is why we restrict us on the case  $\rho \in (1,2]$  in this article.

**Remark 1.3.** For (1.13), it requires less restrictive conditions to have an Itô-Skorohod solution than to have a Stratonovich solution (see Remark 2.17 and see also [42] for a more general class of SPDEs). This explains why condition (1.18) for the Itô-Skorohod case is weaker than condition (1.15) for the Stratonovich case.

The proofs of Theorems 1.1 and 1.2 will be presented in Section 3.2. For the reader's convenience, here we explain briefly the strategy for Theorem 1.1 (the proof of Theorem 1.2 is similar but easier). By Taylor's expansion, the rescaled partition function can be written as

$$Z_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho} x_0) = \sum_{m=0}^{\infty} \frac{1}{m!} \mathbb{S}_m^{(N)},$$

where

(1.19) 
$$\mathbb{S}_{m}^{(N)} = \hat{\beta}_{N}^{m} \sum_{n_{1},\dots,n_{m} \in [[N]]} \sum_{k_{1},\dots,k_{m} \in \mathbb{Z}} \omega(n_{1},k_{1}) \cdots \omega(n_{m},k_{m}) P_{\boldsymbol{n}^{*}}(N^{1/\rho}x_{0};k_{1},\dots,k_{m}),$$

with  $P_{n^*}$  (see eq. (3.4)) being the product of the transition densities of the random walk S. Meanwhile, the Stratonovich solution of the continuum equation (1.13) has the following series representation (see eq. (2.27)):

$$u(1,x_0) = \sum_{m=0}^{\infty} \beta^m \mathbb{I}_m(\mathfrak{g}_m(\cdot;1,x_0)),$$

where  $\mathbb{I}_m(\cdot)$  is a multiple Stratonovich integral (see Section 2.1.2) and  $\mathfrak{g}_m$  (see eq. (2.28)) is the product of the transition densities of the stable process X. We remind the reader that the continuum multiple Stratonovich integral  $\mathbb{I}_m(\mathfrak{g}_m(\cdot; 1, x_0))$  resembles the discrete sum  $\mathbb{S}_m^{(N)}$ .

Under condition (1.15), one can obtain the  $L^1$ -convergence of the series of u(t, x) (see Proposition 2.12) and the uniform (in N)  $L^1$ -convergence of the series of  $Z_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0)$  (see (3.40)). Thus, in order to prove the weak convergence  $Z_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0) \xrightarrow{d} u(1, x_0)$ , in light of Lemma B.1, it suffices to obtain joint weak convergences for  $\mathbb{S}_m^{(N)}$ 's, that is, for  $k \in \mathbb{N}$  and  $l_1, \ldots, l_k \in \mathbb{N}$ ,

$$\left(\frac{1}{l_1!}\mathbb{S}_{l_1}^{(N)},\ldots,\frac{1}{l_k!}\mathbb{S}_{l_k}^{(N)}\right) \stackrel{d}{\longrightarrow} \left(\beta^{l_1}\mathbb{I}_{l_1}(\mathfrak{g}_{l_1}(\cdot;1,x_0)),\ldots,\beta^{l_k}\mathbb{I}_{l_k}(\mathfrak{g}_{l_k}(\cdot;1,x_0))\right), \text{ as } N \to \infty,$$

which is proved in Proposition 3.4.

Multiple Wiener integrals  $\mathbf{I}_m(f)$  and multiple Stratonovich integrals  $\mathbb{I}_m(f)$  are linked via the celebrated Hu-Meyer's formula (2.19). Usually it is more convenient to deal with multiple Wiener integrals  $\mathbf{I}_m(f)$  whose second moment is easier to calculate. For this reason, in Section 3.1 we first consider U-statistics given in (3.8) which are multi-linear Wick polynomials of  $\omega(n_i, k_i)$ 's, and prove that they converge weakly to multiple Wiener integrals (see Proposition 3.3). This together with Hu-Meyer's formula (2.19) then yields Proposition 3.4.

Note that the exponential term in the partition function  $\tilde{Z}_{\omega}^{(N)}$  in (1.16) is actually a Wick exponential (see eq. (2.15)) of  $\beta \sum_{i=1}^{N} \omega(i, S_i)$  conditional on the random walk S. Thus, to prove Theorem 1.2, one only needs Proposition 3.3 on the weak convergence of multi-linear Wick polynomials.

As can be seen, our approach is very much inspired by [1] where the random variables of disorder are i.i.d. However, in our setting the disorder is correlated in time, and it turns out that this difference is critical in the sense that it prevents us from using the trick of "modified partition function" (see in [1, eq. (4)], which was also employed in [37]), as this method relies on the temporal independence of the disorder. For the same reason, the Mayer expansion (see [8, equations (1.5) and (1.7)]), which was employed in [8] to linearize the exponential while introducing an error term that comes from the Itô correction, cannot be applied here either, since it works well only if the disorder is independent in both time and space.

To conclude the introduction, we make some remarks on our main results Theorems 1.1 and 1.2.

(i) In the polymer model, the temporal independence of the disorder plays a critical role. Recently, a directed polymer in time-space correlated random environment was studied in [41] and it was shown that the rescaled partition function converges weakly to a Stratonovich solution. Assuming that the disorder is correlated in time and white in space in our setting, the rescaled partition functions  $Z_{\omega}^{(N)}$  and  $\tilde{Z}_{\omega}^{(N)}$  converge weakly to Stratonovich and Itô-Skorohod solutions, respectively. In contrast, if the disorder possesses temporal independence, after a proper scaling, the limit of the rescaled partition function is an Itô-Skorohod solution but not a Stratonovich solution, as has been shown in [1, 8, 37, 9]. Among others, a technical explanation for the critical role of the temporal independence of the disorder is the following: when the temporal independence of disorder appears, by using the "modified partition function" trick, the term corresponding to  $\mathbb{S}_m^{(N)}$  given by (1.19) becomes a multi-linear polynomial of  $\omega(n_i, k_i)$  for  $i = 1, \ldots, m$  with  $n_1 < n_2 < \cdots < n_m$ , which converges weakly to a multiple Wiener integral due to the independence (see [34, 8]).

(ii) In contrast to the case that the disorder is independent in time where one can employ the "modified partition function" trick (or Mayer's expansion for the case that the disorder is independent both in time and space) to make time indices in the summation distinct from each other, we expand the partition function directly via Taylor's expansion, and as a consequence the time

indices  $n_1, \ldots, n_m$  in the summation (1.19) can be repeated. It turns out that this in fact gives negligible contribution for the Stratonovich case, due to the condition  $\theta > \frac{1}{2}$  in (1.15) which is more restrictive than the condition  $\theta > 0$  in (1.18) for the Skorohod case (see also [1, 8, 37]). For instance, when m = 2, the expectation of the sum of the diagonal terms in (1.19) is  $\beta \gamma(0) N^{1-2\theta}$ which converges to zero as N tends to infinity if we assume  $\theta > 1/2$ .

(iii) The study of the scaling limit of partition functions was initiated in [1] in order to understand the polymer behavior in the so-called *intermediate disorder regime* which sits between weak and strong disorder regimes. Meanwhile, as pointed out in [8], the fact that the rescaled partition functions converge weakly to a non-trivial limit indicates that the directed polymer model is *disorder relevant*, since it implies that the presence of disorder, no matter how small it is, changes the qualitative features of the underlying homogeneous model.

(iv) As pointed out in [32], it is of interest to study directed polymer in random environment which has long-range correlations in time and space. In light of the work of [41], we expect that our results and methodology can be extended to the model in dimension 1 + d with space-time long-range correlated disorder. Note that the corresponding continuum SPDEs have been investigated in [23, 42]. For instance, consider the limiting Gaussian noise with covariance

$$\mathbb{E}\Big[\dot{W}(t,x)\dot{W}(s,y)\Big] = |t-s|^{-\alpha_0}|x-y|^{-\alpha},$$

where  $\alpha_0 \in (0, 1)$  and  $\alpha \in (0, d)$ . Then (1.2) has a Stratonovich solution if  $\alpha < \rho(1 - \alpha_0)$  (see Remark 3.1 in [42]) and an Itô-Skorohod solution if  $\alpha < \rho$  (see Remark 5.1 and Theorem 5.3 in [42]). In particular, this allows to consider the case  $d > 1, \rho \in (0, 1]$  (recall that the spatial independence forces us to consider only the case  $d = 1, \rho \in (1, 2]$ , see Remarks 1.2 and 2.13).

(v) In our model, we assume the Gaussianity of the disorder  $\omega$  for technical reasons. The loss of Gaussian property in  $\omega$  shall cause a lot more complexity in computations (see e.g. (A.7) and (A.8) for a comparison of the computations of moments for Gaussian and non-Gaussian random variables). The functional analytic approach developed in [41] might be a choice to circumvent this difficulty. Nevertheless, we conjecture that our result still holds for sub-Gaussian disorder.

(vi) Assume the disorder  $\{\omega(i,k), i \in \mathbb{N}, k \in \mathbb{Z}\}$  is a family of i.i.d. standard Gaussian random variables. In this situation, the exponent in the correction term of the rescaled partition function  $\tilde{Z}_{\omega}^{(N)}(\hat{\beta}_N, k)$  becomes, recalling (1.17) and replacing  $\gamma(i-j)$  by  $\delta_{ij}$  therein,

$$\frac{1}{2}\hat{\beta}_N^2 \sum_{i,j=1}^N \delta_{ij} \mathbf{1}_{\{S_i=S_j\}} = \frac{1}{2}\hat{\beta}_N^2 N = \frac{\beta^2}{2} N^{\frac{1}{\rho}},$$

which is now independent of the random walk S. Thus, the rescaled partition function for the Itô-Skorohod case can be written as

$$\tilde{Z}^{(N)}_{\omega}(\hat{\beta}_N,k) = e^{-\frac{\beta^2}{2}N^{\frac{1}{\rho}}} Z^{(N)}_{\omega}(\hat{\beta}_N,k),$$

which converges weakly to the Itô-Skorohod solution of (1.13) with space-time white noise. This is consistent with [1, Theorem 2.1], noting that  $N \log \mathbb{E}[e^{\hat{\beta}_N \omega(i,k)}] = \frac{1}{2}\hat{\beta}_N^2 N$  if we assume the disorder is Gaussian.

(vii) If the disorder  $\omega$  and the limiting noise  $\dot{W}$  live in the same probability space such that  $\dot{W}$  is the scaling limit of  $\omega$ , we can get strong convergence (a.s. convergence) in Theorem 1.1 and Theorem 1.2. In this sense, the rescaled partition functions  $Z_{\omega}^{(N)}$  and  $\tilde{Z}_{\omega}^{(N)}$  can be viewed as approximations for the Stratonovich solution and the Itô-Skorohod solution of equation (1.13), respectively. See e.g. Foondun at el. [15] and Joseph et al. [28] for related results.

(viii) As a byproduct, we obtain a Jensen type inequality for the integrals induced by fractional Brownian motion (see Lemmas 2.2 and C.1) which is new in the literature to our best knowledge. We provide a new approach to prove the exponential integrability of the weighted self-intersection local time of the  $\rho$ -stable process in Proposition 2.7 which plays a critical role in obtaining the Feynman-Kac formula for the solution of (1.13); this approach also simplifies the proof in [23] which works exclusively for the weighted self-intersection local time of Brownian motion.

The rest of the paper is organized as follows. In Section 2, we provide some preliminaries on Gaussian spaces and then study the limiting SPDE (1.13). The main results Theorem 1.1 and Theorem 1.2 are proved in Section 3. In Appendices A, B, and C, we collect some preliminaries on Wick products, convergence of probability measures, and some other miscellaneous results that are used in this article.

### 2. On the SPDE

In this section, we first recall some preliminaries on Gaussian spaces and then we study the limiting continuum SPDE (1.13).

2.1. Preliminaries on Gaussian spaces. In this subsection, we provide some preliminaries on Gaussian spaces. We refer to [21, 26, 35] for more details.

2.1.1. Banach spaces associated with  $\dot{W}$ . On probability space  $(\Omega, \mathcal{F}, P)$  satisfying usual conditions, let  $\dot{W} = \{\dot{W}(t, x) : t \in [0, 1], x \in \mathbb{R}\}$  be real-valued centered Gaussian noise with covariance

(2.1) 
$$\mathbb{E}[\dot{W}(t,x)\dot{W}(s,y)] = K(t-s,x-y) = |t-s|^{2H-2}\delta(x-y),$$

where  $\delta$  is the Dirac delta function. The Hilbert space  $\mathscr{H}$  associated with  $\dot{W}$  is the completion of smooth functions with compact support under the inner product

(2.2) 
$$\langle f,g \rangle_{\mathscr{H}} = \int_0^1 \int_0^1 \int_{\mathbb{R}} \int_{\mathbb{R}} f(s,x) K(s-t,x-y) g(t,y) \mathrm{d}x \mathrm{d}y \mathrm{d}s \mathrm{d}t \\ = \int_0^1 \int_0^1 \int_{\mathbb{R}} |s-t|^{2H-2} f(s,x) g(t,x) \mathrm{d}x \mathrm{d}s \mathrm{d}t.$$

We assume  $H \in (1/2, 1]$ . For the case H = 1, the noise  $\dot{W}$  is indeed a spatial white noise which does not depend on time.

We remark that the Hilbert space  $\mathscr{H}$  contains distributions (see [36]). For our purpose, it suffices to just consider classical measurable functions. We introduce the following Banach space  $\mathscr{B}$  which is a subset of the set  $\mathscr{B}([0,1] \times \mathbb{R})$  of measurable functions on  $[0,1] \times \mathbb{R}$ .

**Definition 2.1.** We define

(2.3)  
$$\mathcal{B} \stackrel{\text{def}}{=} \left\{ f \in \mathscr{B}([0,1] \times \mathbb{R}) : \|f\|_{\mathcal{B}} \stackrel{\text{def}}{=} \left( \int_{0}^{1} \int_{0}^{1} \int_{\mathbb{R}} |s-t|^{2H-2} |f(s,x)| |f(t,x)| \mathrm{d}x \mathrm{d}s \mathrm{d}t \right)^{\frac{1}{2}} < \infty \right\}.$$

Clearly  $\mathcal{B}$  is a dense subset of  $\mathcal{H}$ , and if  $f \in \mathcal{H}$  is a nonnegative function, f also belongs to  $\mathcal{B}$ .

For  $N \in \mathbb{N}$  and  $f \in \mathcal{B}^{\otimes m}$ , we denote by  $\mathcal{A}_N(f)$  the conditional expectation of f with respect to the  $\sigma$ -algebra  $\mathscr{B}_N^m = \sigma(\mathfrak{R}_N^m)$  generated by the set  $\mathfrak{R}_N^m$  of rectangles

(2.4) 
$$\mathfrak{R}_{N}^{m} := \left\{ \left[ \frac{\boldsymbol{i}}{N}, \frac{\boldsymbol{i} + \overrightarrow{1}_{m}}{N} \right] \times \left[ \frac{\boldsymbol{k}}{N^{1/\rho}}, \frac{\boldsymbol{k} + \overrightarrow{1}_{m}}{N^{1/\rho}} \right] : \boldsymbol{i} \in [\![N]\!]^{m}, \boldsymbol{k} \in \mathbb{Z}^{m} \right\},$$

where  $1'_m$  is the *m*-dimensional vector of all ones. That is,  $\mathcal{A}_N(f)$  is defined by the average values of f on the blocks  $B \in \mathfrak{R}_N^m$ :

(2.5) 
$$\mathcal{A}_N(f)(\boldsymbol{t},\boldsymbol{x}) = \frac{1}{|B|} \int_B f(\boldsymbol{s},\boldsymbol{y}) \mathrm{d}\boldsymbol{s} \mathrm{d}\boldsymbol{y} \cdot \mathbf{1}_B(\boldsymbol{t},\boldsymbol{x}),$$

where  $\mathbf{1}_B$  is an indicator function and |B| is the Lebesgue measure of B.

The following Jensen type of inequality will be used to prove the weak convergence of U-statistics in Section 3.1.

**Lemma 2.2.** For  $m, N \in \mathbb{N}$ , consider  $f \in \mathcal{B}^{\otimes m}$  and let  $\mathcal{A}_N(f)$  be given by (2.5). Then, we have (2.6)  $\|\mathcal{A}_N(f)\|_{\mathcal{B}^{\otimes m}} \leq C^m \|f\|_{\mathcal{B}^{\otimes m}}$ ,

where C is a constant depending only on H.

*Proof* We prove (2.6) for m = 1 and the general case can be proved in a similar way. Note that by Fubini's theorem,  $\mathcal{A}_N(f) = \mathcal{A}_N^{t}(\mathcal{A}_N^{x}(f)) = \mathcal{A}_N^{x}(\mathcal{A}_N^{t}(f))$ , where  $\mathcal{A}_N^{t}$  (resp.  $\mathcal{A}_N^{x}$ ) means taking average in time (resp. space) only. Therefore, we have

$$\|\mathcal{A}_N(f)\|_{\mathcal{B}} = \|\mathcal{A}_N^{\mathrm{t}}(\mathcal{A}_N^{\mathrm{x}}(f))\|_{\mathcal{B}} \le C\|\mathcal{A}_N^{\mathrm{x}}(f)\|_{\mathcal{B}},$$

for some C depending on H only, where the inequality follows from Lemma C.1.

Now, it suffices to prove  $\|\mathcal{A}_N^{\mathbf{x}}(f)\|_{\mathcal{B}} \leq \|f\|_{\mathcal{B}}$ . Using the identity, for  $H \in (\frac{1}{2}, 1)$ ,

$$|s-t|^{2H-2} = c_H \int_{\mathbb{R}} |s-\tau|^{H-\frac{3}{2}} |t-\tau|^{H-\frac{3}{2}} d\tau$$

where  $c_H$  is a finite number only depending on H, we have

$$\begin{split} \|\mathcal{A}_{N}^{\mathbf{x}}(f)\|_{\mathcal{B}}^{2} &= c_{H} \int_{\mathbb{R}} \mathrm{d}x \int_{0}^{1} \mathrm{d}s \int_{0}^{1} \mathrm{d}t \int_{\mathbb{R}} \mathrm{d}\tau |s-\tau|^{H-\frac{3}{2}} |t-\tau|^{H-\frac{3}{2}} |\mathcal{A}_{N}^{\mathbf{x}}(f)| \, (s,x) \, |\mathcal{A}_{N}^{\mathbf{x}}(f)| \, (t,x) \\ &= c_{H} \int_{\mathbb{R}} \mathrm{d}x \int_{\mathbb{R}} \mathrm{d}\tau \left( \int_{0}^{1} \mathrm{d}s |s-\tau|^{H-\frac{3}{2}} |\mathcal{A}_{N}^{\mathbf{x}}(f)| \, (s,x) \right)^{2} \\ &\leq c_{H} \int_{\mathbb{R}} \mathrm{d}x \int_{\mathbb{R}} \mathrm{d}\tau \left( \mathcal{A}_{N}^{\mathbf{x}} \left( \int_{0}^{1} \mathrm{d}s |s-\tau|^{H-\frac{3}{2}} |f|(s,\cdot) \right) (\tau,x) \right)^{2} \\ &\leq c_{H} \int_{\mathbb{R}} \mathrm{d}x \int_{\mathbb{R}} \mathrm{d}\tau \left( \int_{0}^{1} \mathrm{d}s |s-\tau|^{H-\frac{3}{2}} |f|(s,x) \right)^{2} \\ &= \|f\|_{\mathcal{B}}^{2}, \end{split}$$

where the last inequality follows from the classical Jensen's inequality.

2.1.2. Chaos expansion, Wick product, multiple integrals, etc. Recall that  $\mathscr{H}$  is the Hilbert space associated with Gaussian noise  $\dot{W}$  with covariance (2.1). Let  $\{W(f), f \in \mathscr{H}\}$  be an isonormal Gaussian process with covariance

(2.7) 
$$\mathbb{E}[W(f)W(q)] = \langle f, q \rangle_{\mathscr{H}}.$$

In particular, if  $f = \mathbf{1}_{[0,t] \times [0,x]}$ , we denote

$$W(t,x) \stackrel{\text{def}}{=} W(\mathbf{1}_{[0,t]\times[0,x]}).$$

Then the fractional noise  $\dot{W}(t,x)$  can be viewed as the partial derivative  $\frac{\partial^2}{\partial t \partial x} W(t,x)$  in the sense of distribution. For  $f \in \mathscr{H}$ , we also use the integral form to denote the Wiener integral W(f):

$$\int_0^1 \int_{\mathbb{R}} f(s, y) W(\mathrm{d}s, \mathrm{d}y) \stackrel{\text{def}}{=} W(f).$$

For  $m \in \mathbb{N} \cup \{0\}$ , let

(2.8) 
$$H_m(x) \stackrel{\text{def}}{=} (-1)^m \mathrm{e}^{x^2/2} \frac{\mathrm{d}^m}{\mathrm{d}x^m} \mathrm{e}^{-x^2/2}, \quad x \in \mathbb{R}$$

be the *m*th Hermite polynomial. For  $g \in \mathscr{H}$ , the multiple Wiener integral of  $g^{\otimes m} \in \mathscr{H}^{\otimes m}$  can be defined via (see e.g. [20, 35])

(2.9) 
$$\mathbf{I}_m(g^{\otimes m}) \stackrel{\text{def}}{=} \|g\|_{\mathscr{H}}^m H_m\left(W(g)\|g\|_{\mathscr{H}}^{-1}\right).$$

In particular, we have  $W(g) = \mathbf{I}_1(g)$  and  $\mathbf{I}_m(g^{\otimes m}) =: W(g)^m$ :, where  $: W(g)^m$ : is a physical Wick product (see Section A).

For  $f \in \mathscr{H}^{\otimes m}$ , let  $\hat{f}$  be its symmetrization, i.e.,

$$\hat{f}(t_1, x_1, \dots, t_m, x_m) = \frac{1}{m!} \sum_{\sigma \in \mathcal{P}_m} f(t_{\sigma(1)}, x_{\sigma(1)}, \dots, t_{\sigma(m)}, x_{\sigma(m)}),$$

where  $\mathcal{P}_m$  is the set of all permutations of  $\llbracket m \rrbracket = \{1, 2, \ldots, m\}$ . Let  $\mathscr{H}^{\hat{\otimes}m}$  be the symmetrization of  $\mathscr{H}^{\otimes m}$ . Then for  $f \in \mathscr{H}^{\hat{\otimes}m}$ , one can define the *m*th multiple Wiener integral  $\mathbf{I}_m(f)$  via (2.9) by a limiting argument. Moreover, for  $f \in \mathscr{H}^{\hat{\otimes}m}$  and  $g \in \mathscr{H}^{\hat{\otimes}n}$ , we have

(2.10) 
$$\mathbb{E}[\mathbf{I}_m(f)\mathbf{I}_n(g)] = m! \langle f, g \rangle_{\mathscr{H}^{\otimes m}} \delta_{mn}$$

where we recall that  $\delta_{mn}$  is the Kronecker delta function. For  $f \in \mathscr{H}^{\otimes m}$  which is not necessarily symmetric, we simply let  $\mathbf{I}_m(f) \stackrel{\text{def}}{=} \mathbf{I}_m(\hat{f})$ . We also take the following notation for multiple Wiener integrals:

(2.11) 
$$\int_{([0,1]\times\mathbb{R})^m} f(\boldsymbol{t},\boldsymbol{x}) W(\mathrm{d}t_1,\mathrm{d}x_1) \diamond \cdots \diamond W(\mathrm{d}t_m,\mathrm{d}x_m) \stackrel{\mathrm{def}}{=} \mathbf{I}_m(f), \text{ for } f \in \mathscr{H}^{\otimes m}$$

For  $f \in \mathscr{H}^{\hat{\otimes}m}$  and  $g \in \mathscr{H}^{\hat{\otimes}n}$ , their contraction of r indices for  $1 \leq r \leq m \wedge n$  is defined by

$$(f \otimes_r g)(t_1, x_1, \dots, t_{m+n-2r}, x_{m+n-2r}) \stackrel{\text{def}}{=} \int_{[0,1]^{2r}} \int_{\mathbb{R}^{2r}} f(t_1, x_1, \dots, t_{n-r}, x_{n-r}, u_1, y_1, \dots, u_r, y_r) \prod_{i=1}^r K(u_i - v_i, y_i - z_i) \times g(t_{n-r+1}, x_{n-r+1}, \dots, t_{n+m-2r}, x_{n+m-2r}, v_1, z_1, \dots, v_r, z_r) \mathrm{d}\boldsymbol{y} \mathrm{d}\boldsymbol{z} \mathrm{d}\boldsymbol{u} \mathrm{d}\boldsymbol{v},$$

and we have the following recursive formulas:

(2.12) 
$$\mathbf{I}_m(f)\mathbf{I}_n(g) = \sum_{r=0}^{m \wedge n} r! \binom{n}{r} \binom{m}{r} \mathbf{I}_{m+n-2r} (f \otimes_r g).$$

In particular, when n = 1 we have,

(2.13) 
$$\mathbf{I}_m(f)\mathbf{I}_1(g) = \mathbf{I}_{m+1}(f \otimes g) + m\mathbf{I}_{m-1}(f \otimes_1 g).$$

Square integrable random variables have a unique chaos expansion as stated below.

**Proposition 2.3.** Let  $\mathcal{G}$  be the  $\sigma$ -field generated by W. Then for any  $F \in L^2(\Omega, \mathcal{G}, P)$ , it admits a unique chaos expansion

$$F = \sum_{m=0}^{\infty} \mathbf{I}_m(f_m) \quad with \quad f_m \in \mathscr{H}^{\hat{\otimes}m},$$

where the series converges in  $L^2$ . Moreover,

$$\mathbb{E}[F^2] = \sum_{m=0}^{\infty} m! \|f_m\|_{\mathscr{H}^{\otimes m}}^2.$$

For  $f \in \mathscr{H}^{\otimes m}$  and  $g \in \mathscr{H}^{\otimes n}$ , the probabilistic Wick product (or Wick product) of  $\mathbf{I}_m(f)$  and  $\mathbf{I}_n(g)$  is defined by

(2.14) 
$$\mathbf{I}_m(f) \diamond \mathbf{I}_n(g) \stackrel{\text{def}}{=} \mathbf{I}_{m+n}(f \otimes g).$$

Unlike the *physical* Wick product, the Wick product defined by (2.14) has the same properties as the ordinary product. As an example, by (2.14) and (2.13), we can show that for  $f \in \mathscr{H}^{\otimes m}$  and  $g \in \mathscr{H}$ ,

$$\mathbf{I}_m(f)\diamond\mathbf{I}_1(g)=\mathbf{I}_m(f)\mathbf{I}_1(g)-m\mathbf{I}_{m-1}(f\otimes_1 g).$$

For two square integrable random variables  $F = \sum_{m=0}^{\infty} \mathbf{I}_m(f_m)$  and  $G = \sum_{n=0}^{\infty} \mathbf{I}_n(g_n)$ , their Wick product is given by

$$F \diamond G = \sum_{m=0}^{\infty} \sum_{n=0}^{\infty} \mathbf{I}_{m+n}(f_m \otimes g_n),$$

as long as the right-hand side is well defined. For instance, let  $\varepsilon(u)$  be the *exponential vector* of u for  $u \in \mathscr{H}$ :

$$\varepsilon(u) \stackrel{\text{def}}{=} \exp\left(W(u) - \frac{\|u\|_{\mathscr{H}}^2}{2}\right)$$

Then  $\varepsilon(u) \diamond \varepsilon(v) = \varepsilon(u+v)$  for  $u, v \in \mathscr{H}$ . We also recall that for a centered Gaussian random variable F, its Wick exponential is given by

(2.15) 
$$\exp^{\diamond}(F) \stackrel{\text{def}}{=} \exp\left(F - \frac{1}{2}\mathbb{E}[F^2]\right) = \sum_{m=0}^{\infty} \frac{1}{m!} F^{\diamond m}.$$

As mentioned in Section A, the physical and probabilistic Wick products of a Gaussian vector coincide. In particular, for  $f_i \in \mathcal{H}, i = 1, ..., n$ , we have

(2.16) 
$$:\mathbf{I}_1(f_1)\cdots\mathbf{I}_1(f_n):=\mathbf{I}_1(f_1)\diamond\cdots\diamond\mathbf{I}_1(f_n).$$

In contrast to the multiple Wiener integral given in (2.9), the multiple Stratonovich integral of  $g^{\otimes m}$  for  $g \in \mathscr{H}$  is defined by ([19, 21]):

(2.17) 
$$\mathbb{I}_m(g^{\otimes m}) \stackrel{\text{def}}{=} W(g)^m.$$

By a limiting argument, one can define  $\mathbb{I}_m(g)$  for g in some appropriate space. We also take the following notation for  $\mathbb{I}_m(f)$  (in comparison with  $\mathbf{I}_m(f)$  in (2.11)):

$$\mathbb{I}_m(f) = \int_{([0,1]\times\mathbb{R})^m} f(\boldsymbol{t},\boldsymbol{x}) W(\mathrm{d}t_1,\mathrm{d}x_1)\cdots W(\mathrm{d}t_m,\mathrm{d}x_m)$$
$$= \int_{([0,1]\times\mathbb{R})^m} f(\boldsymbol{t},\boldsymbol{x}) \prod_{i=1}^m W(\mathrm{d}t_i,\mathrm{d}x_i).$$

For 
$$f \in \mathscr{H}^{\otimes m}$$
, define the *k*th order trace  $\operatorname{Tr}^{k} f$  of  $f$  by  
(2.18)  $\operatorname{Tr}^{k} f(t_{1}, x_{1}, \dots, t_{m-2k}, x_{m-2k})$   
 $\stackrel{\text{def}}{=} \int_{([0,1] \times \mathbb{R})^{2k}} f(s_{1}, y_{1}, \dots, s_{2k}, y_{2k}, t_{1}, x_{1}, \dots, t_{m-2k}, x_{m-2k})$   
 $\times K(s_{1} - s_{2}, y_{1} - y_{2}) \cdots K(s_{2k-1} - s_{2k}, y_{2k-1} - y_{2k}) \mathrm{d}s \mathrm{d}s.$ 

The following Hu-Meyer's formula (see [19], [14], [27]) connects multiple Stratonovich integrals  $\mathbb{I}_m(f)$  with multiple Wiener integrals  $\mathbf{I}_m(f)$ ,

(2.19) 
$$\mathbb{I}_{m}(f) = \sum_{k=0}^{\left[\frac{m}{2}\right]} \frac{m!}{k!(m-2k)!2^{k}} \,\mathbf{I}_{m-2k}(\mathrm{Tr}^{k}\hat{f}), \ f \in \mathscr{H}^{\otimes m},$$

as long as the right-hand side is well-defined, i.e.,  $\operatorname{Tr}^k \hat{f} \in \mathscr{H}^{\otimes (m-2k)}$  for  $k = 0, 1, \dots [m/2]$ . In this case, we have

(2.20) 
$$\|f\|_{\mathbf{S}_m}^2 \stackrel{\text{def}}{=} \mathbb{E}[|\mathbb{I}_m(f)|^2] = \sum_{k=0}^{\left\lfloor\frac{m}{2}\right\rfloor} \frac{1}{(m-2k)!} \left(\frac{m!}{k!2^k}\right)^2 \|\operatorname{Tr}^k \hat{f}\|_{\mathscr{H}^{\otimes (m-2k)}}^2.$$

To end this section, we introduce the stochastic Fubini theorem which will be used in the proof of Proposition 2.10. Note that stochastic Fubini theorem has been proved in different contexts (see e.g. [13, 29]), and here we provide a version working for multiple Wiener and Stratonovich integrals.

Let  $(\mathbf{X}, \mathcal{X}, \mu)$  be a measurable space with  $\mu(\mathbf{X}) < \infty$ . For a measurable function  $f : \mathbf{X} \to \mathbb{R}$ , we take the notation

$$\mu(f) \stackrel{\text{def}}{=} \int_{\mathbf{X}} f(x)\mu(\mathrm{d}x).$$

The following result holds for general Gaussian spaces associated with a *separable* Hilbert space.

**Proposition 2.4.** For  $m \in \mathbb{N}$ , let  $h : \mathbf{X} \to \mathscr{H}^{\otimes m}$  be a measurable mapping such that

$$\mu(\mathbb{E}[|\mathbf{I}_m(h)|^2]) = \mu(||h||^2_{\mathscr{H}^{\otimes m}}) < \infty.$$

Then the following stochastic Fubini theorem holds,

(2.21) 
$$\mu(\mathbf{I}_m(h)) = \mathbf{I}_m(\mu(h)).$$

Similarly, if we assume

$$\mu\left(\mathbb{E}[|\mathbb{I}_m(h)|^2]\right) = \mu(\|h\|_{\mathbf{S}_m}^2) = \sum_{k=0}^{\left\lfloor\frac{m}{2}\right\rfloor} \frac{1}{(m-2k)!} \left(\frac{m!}{k!2^k}\right)^2 \mu\left(\|\mathrm{Tr}^k\hat{h}\|_{\mathscr{H}^{\otimes(m-2k)}}^2\right) < \infty,$$

we also have

(2.22) 
$$\mu(\mathbb{I}_m(h)) = \mathbb{I}_m(\mu(h)).$$

*Proof* Let  $\{e_i, i \in \mathbb{N}\}$  be an orthonormal basis of  $\mathcal{H}$ . Suppose

$$h(x) = \sum_{i_1,\dots,i_m=1}^{\infty} \alpha_{i_1,\dots,i_m}(x) e_{i_1} \otimes \dots \otimes e_{i_m}.$$

Then the condition  $\mu(\|h\|_{\mathscr{H}^{\otimes m}}^2) < \infty$  implies  $\sum_{i_1,\ldots,i_m} \mu(\alpha_{i_1,\ldots,i_m}^2) < \infty$ . This further implies

(2.23) 
$$\mu(h) = \sum_{i_1,\dots,i_m} \mu(\alpha_{i_1,\dots,i_m}) e_{i_1} \otimes \dots \otimes e_{i_m},$$

where the series on the right-hand side converges in the Hilbert space  $\mathscr{H}^{\otimes m}$  due to Jensen's inequality and the assumption  $\mu(\mathbf{X}) < \infty$ .

Consider  $g \in \mathscr{H}^{\otimes m}$  and suppose

$$g = \sum_{i_1,\dots,i_m=1}^{\infty} \beta_{i_1,\dots,i_m} e_{i_1} \otimes \dots \otimes e_{i_m}$$

with  $||g||_{\mathscr{H}^{\otimes m}}^2 = \sum_{i_1,\dots,i_m} \beta_{i_1,\dots,i_m}^2 < \infty$ . We have

(2.24) 
$$\mu(\langle h, g \rangle_{\mathscr{H}^{\otimes m}}) = \mu\left(\sum_{i_1, \dots, i_m} \alpha_{i_1, \dots, i_m} \beta_{i_1, \dots, i_m}\right)$$
$$= \sum_{i_1, \dots, i_m} \mu(\alpha_{i_1, \dots, i_m}) \beta_{i_1, \dots, i_m} = \langle \mu(h), g \rangle_{\mathscr{H}^{\otimes m}},$$

where the second equality follows from classical Fubini's theorem, noting that

$$\mu\left(\sum_{i_1,\dots,i_m} |\alpha_{i_1,\dots,i_m}\beta_{i_1,\dots,i_m}|\right) \le \mu\left(\left(\sum_{i_1,\dots,i_m} \alpha_{i_1,\dots,i_m}^2\right)^{1/2}\right) \left(\sum_{i_1,\dots,i_m} \beta_{i_1,\dots,i_m}^2\right)^{1/2} < \infty,$$

and the last equality follows from (2.23).

By (2.24) we have  $\mathbb{E}[\mu(\mathbf{I}_m(h))\mathbf{I}_m(g)] = \mathbb{E}[\mathbf{I}_m(\mu(h))\mathbf{I}_m(g)]$  for any  $g \in \mathscr{H}^{\otimes m}$ , and thus the desired equality (2.21) holds. Finally, equation (2.22) follows from (2.21) and (2.19).

2.2. Mild Stratonovich solution. In this subsection, we consider the fractional SHE (1.13) with initial value  $u(0,x) = u_0(x), x \in \mathbb{R}$ . We call  $\{u(t,x)\}_{(t,x)\in\mathbb{R}_+\times\mathbb{R}}$  a mild Stratonovich solution of (1.13) if u(t,x) is an adapted process (with respect to the filtration  $(\mathcal{F}_t)_{t\geq 0}$  where  $\mathcal{F}_t = \sigma(W(s,x), 0 \leq s \leq t, x \in \mathbb{R})$ ) such that  $\mathbb{E}[|u(t,x)|^2] < \infty$  for all  $(t,x) \in \mathbb{R}_+ \times \mathbb{R}$  and satisfies the following integral equation

(2.25) 
$$u(t,x) = \int_{\mathbb{R}} \mathfrak{g}(t,x-y)u_0(y)\mathrm{d}y + \beta \int_0^t \int_{\mathbb{R}} \mathfrak{g}(t-r,x-y)u(r,y)W(\mathrm{d}r,\mathrm{d}y)$$

where  $\mathfrak{g}(t, x)$  is the density function of the  $\rho$ -stable process  $\{X_t, t \ge 0\}$  with  $X_0 = 0$  and the integral on the right-hand side is understood in the sense of Stratonovich which will be specified below.

Assuming  $u_0(x) \equiv 1$ , we have

(2.26) 
$$u(t,x) = 1 + \beta \int_0^t \int_{\mathbb{R}} \mathfrak{g}(t-r,x-y)u(r,y)W(\mathrm{d}r,\mathrm{d}y).$$

Iterate the equation and formally we have a series expansion for a mild Stratonovich solution:

(2.27)  
$$u(t,x) = 1 + \sum_{m=1}^{\infty} \beta^m \int_{[0,t]_{<}^m} \int_{\mathbb{R}^m} \prod_{i=1}^m \mathfrak{g}(t_{i+1} - t_i, x_{i+1} - x_i) W(\mathrm{d}t_i, \mathrm{d}x_i)$$
$$= 1 + \sum_{m=1}^{\infty} \beta^m \int_{[0,t]_{<}^m} \int_{\mathbb{R}^m} \mathfrak{g}_m(t, \boldsymbol{x}; t, x) \prod_{i=1}^m W(\mathrm{d}t_i, \mathrm{d}x_i)$$
$$= 1 + \sum_{m=1}^{\infty} \beta^m \mathbb{I}_m(\mathfrak{g}_m(\cdot; t, x)),$$

where

(2.28) 
$$\mathfrak{g}_m(\boldsymbol{t}, \boldsymbol{x}; t, x) \stackrel{\text{def}}{=} \prod_{i=1}^m \mathfrak{g}(t_{i+1} - t_i, x_{i+1} - x_i) \mathbf{1}_{[0,1]^m_{\leq}}(t_1, \dots, t_m)$$

with

$$[0,t]^m \stackrel{\text{def}}{=} \{0 < t_1 < \dots < t_m < t\} \text{ and } t_{m+1} \stackrel{\text{def}}{=} t, x_{m+1} \stackrel{\text{def}}{=} x,$$

and on the right-hand side are *multiple Stratonovich integrals*. The series in (2.27) converges in  $L^1(\Omega)$  under the condition (1.15) (see Proposition 2.12). We also refer to [23, 42] for an equivalent description.

Knowing that u(t, x) solves a stochastic heat equation, we now aim to derive a Feynman-Kac type representation for u(t, x). Given a path of the  $\rho$ -stable process X, the following Wiener integral

(2.29) 
$$\mathcal{I}_{t,x}^{\varepsilon} \stackrel{\text{def}}{=} \int_0^t \int_{\mathbb{R}} p_{\varepsilon} (X_{t-r}^x - y) W(\mathrm{d}r, \mathrm{d}y),$$

is well defined, where  $X_s^x \stackrel{\text{def}}{=} X_s + x$  and  $p_{\varepsilon}(x) \stackrel{\text{def}}{=} \frac{1}{\sqrt{2\pi\varepsilon}} e^{-\frac{x^2}{2\varepsilon}}$  is the heat kernel.

As for the discrete model, we use  $\mathbb{E}_X$  and  $\mathbb{E}_W$  to denote the expectations in the probability space of X and W, respectively, and we abuse the notation  $\mathbb{E} = \mathbb{E}_W \times \mathbb{E}_X$  in this section.

**Proposition 2.5.** Assume the condition (1.15). Then, for each  $\varepsilon > 0$ ,  $p_{\varepsilon}(X_{t-.}^{x} - \cdot)$  belongs to  $\mathscr{H}$  a.s. and the family of random variables  $\mathcal{I}_{t,x}^{\varepsilon}$  defined by (2.29) converges in  $L^{2}$  to a limit denoted by

(2.30) 
$$\mathcal{I}_{t,x} \stackrel{\text{def}}{=} \int_0^t \int_{\mathbb{R}} \delta(X_{t-r}^x - y) W(\mathrm{d}r, \mathrm{d}y),$$

where  $\delta(X_{t-.}^x - \cdot)$  is an  $\mathscr{H}$ -valued random variable given by the  $L^2$ -limit of  $p_{\varepsilon}(X_{t-.}^x - \cdot)$ . Moreover,  $\delta(X_{t-.}^x - \cdot) \in L^2(\Omega_X, \mathcal{G}, P; \mathscr{H})$ , where  $(\Omega_X, \mathcal{G}, P)$  is the probability space of X.

Conditional on X,  $\mathcal{I}_{t,x}$  is a Gaussian random variable with mean 0 and variance

(2.31) 
$$\operatorname{Var}[\mathcal{I}_{t,x}|X] = \int_0^t \int_0^t |r-s|^{2H-2} \delta(X_r - X_s) \mathrm{d}r \mathrm{d}s.$$

*Proof* For  $\varepsilon, \sigma > 0$ , by (2.2) we have for  $t \leq 1$ ,

(2.32) 
$$\left\langle p_{\varepsilon}(X_{t-\cdot}^{x} - \cdot)\mathbf{I}_{[0,t]}, p_{\sigma}(X_{t-\cdot}^{x} - \cdot)\mathbf{I}_{[0,t]} \right\rangle_{\mathscr{H}}$$
$$= \int_{0}^{t} \int_{0}^{t} \int_{\mathbb{R}} |s - r|^{2H-2} p_{\varepsilon}(X_{s}^{x} - y) p_{\sigma}(X_{r}^{x} - y) \mathrm{d}y \mathrm{d}s \mathrm{d}r$$
$$= \int_{0}^{t} \int_{0}^{t} |s - r|^{2H-2} p_{\varepsilon+\sigma}(X_{s} - X_{r}) \mathrm{d}s \mathrm{d}r.$$

By (1.8), we have

$$\mathbb{E}[p_{\varepsilon+\sigma}(X_s - X_r)] = \int_{\mathbb{R}} p_{\varepsilon+\sigma}(y)\mathfrak{g}(|r-s|, y) \mathrm{d}y \lesssim |r-s|^{-1/\rho}.$$

Thus, the condition (1.15) yields

$$\mathbb{E}\left\langle p_{\varepsilon}(X_{t-\cdot}^{x}-\cdot), p_{\sigma}(X_{t-\cdot}^{x}-\cdot)\right\rangle_{\mathscr{H}} \lesssim \int_{0}^{t} \int_{0}^{t} |s-r|^{2H-2-\frac{1}{\rho}} \mathrm{d}s \mathrm{d}r < \infty,$$

hence  $p_{\varepsilon}(X_{t-\cdot}^x - \cdot)$  belongs to  $\mathscr{H}$  for all  $\varepsilon > 0$  almost surely and

(2.33) 
$$\mathbb{E}[\mathcal{I}_{t,x}^{\varepsilon}\mathcal{I}_{t,x}^{\sigma}] = \mathbb{E}\left\langle p_{\varepsilon}(X_{t-\cdot}^{x} - \cdot), p_{\sigma}(X_{t-\cdot}^{x} - \cdot)\right\rangle_{\mathscr{H}}$$
$$= \mathbb{E}\int_{0}^{t}\int_{0}^{t}|s-r|^{2H-2}p_{\varepsilon+\sigma}(X_{s}-X_{r})\mathrm{d}s\mathrm{d}r$$

As  $(\varepsilon, \sigma) \to 0$ , by the dominated convergence theorem we have

$$\lim_{(\varepsilon,\sigma)\to 0} \mathbb{E}[\mathcal{I}_{t,x}^{\varepsilon}\mathcal{I}_{t,x}^{\sigma}] = \mathbb{E}\int_{0}^{t}\int_{0}^{t}|s-r|^{2H-2}\delta(X_{s}-X_{r})\mathrm{d}s\mathrm{d}r$$
$$= C\int_{0}^{t}\int_{0}^{t}|s-r|^{2H-2-\frac{1}{\rho}}\mathrm{d}s\mathrm{d}r < \infty$$

for some proper constant C. As a consequence, we have  $\lim_{(\varepsilon,\sigma)\to 0} \mathbb{E}\left[\left(\mathcal{I}_{t,x}^{\varepsilon} - \mathcal{I}_{t,x}^{\sigma}\right)^2\right] = 0.$ 

This implies that, as  $\varepsilon$  converges to zero,  $p_{\varepsilon}(X_{t-.}^x - \cdot)$  is a Cauchy sequence in  $L^2(\Omega_X, \mathcal{G}, P; \mathcal{H})$ , and the limit is denoted by  $\delta(X_{t-.}^x - \cdot)$ . Therefore,  $\mathcal{I}_{t,x}^{\varepsilon}$  converges to  $\mathcal{I}_{t,x}$  given in (2.30) in  $L^2$ . Finally, (2.31) can be proven by a similar argument.

**Remark 2.6.** By the computation above, it is clear that we have

$$\mathbb{E}[|\mathcal{I}_{t,x}|^2] = C \int_0^t \int_0^t |s-r|^{2H-2-\frac{1}{\rho}} \mathrm{d}s \mathrm{d}r,$$

which is finite iff (1.15) holds. It turns out that (1.15) is also sufficient (and necessary, of course) for  $\mathbb{E}[e^{\mathcal{I}_{t,x}}] < \infty$  by (2.31) and Proposition 2.7 below.

**Proposition 2.7.** Under the condition (1.15), we have

(2.34) 
$$\mathbb{E}\left[\exp\left(\beta \int_0^1 \int_0^1 |r-s|^{2H-2} \delta(X_r - X_s) \mathrm{d}r \mathrm{d}s\right)\right] < \infty \text{ for all } \beta > 0$$

*Proof* The proof essentially follows from that of [42, Theorem 3.3] with some minor modifications. Taylor expansion and Fubini's theorem yield

$$(2.35) \qquad \begin{split} \mathbb{E}\left[\exp\left(\beta\int_{0}^{1}\int_{0}^{1}|r-s|^{2H-2}\delta(X_{r}-X_{s})\mathrm{d}r\mathrm{d}s\right)\right] \\ &=\sum_{m=0}^{\infty}\frac{1}{m!}\left(\frac{\beta}{2\pi}\right)^{m}\int_{[0,1]^{2m}}\int_{\mathbb{R}^{m}}\prod_{i=1}^{m}|s_{2i}-s_{2i-1}|^{2H-2}\mathbb{E}\left[\mathrm{e}^{i\sum_{i=1}^{m}\xi_{i}(X_{s_{2i}}-X_{s_{2i-1}})}\right]\mathrm{d}\boldsymbol{\xi}\mathrm{d}\boldsymbol{s} \\ &=\sum_{m=0}^{\infty}\frac{1}{m!}\left(\frac{\beta}{2\pi}\right)^{m}\sum_{\sigma\in\mathcal{P}_{2m}}\int_{[0,1]^{2m}}\int_{\mathbb{R}^{m}}\prod_{i=1}^{m}|s_{\sigma(2i)}-s_{\sigma(2i-1)}|^{2H-2} \\ &\times\mathbb{E}\left[\mathrm{e}^{i\sum_{i=1}^{m}\xi_{i}(X_{s_{\sigma(2i}}-X_{s_{\sigma(2i-1)}})}\right]\mathrm{d}\boldsymbol{\xi}\mathrm{d}\boldsymbol{s}, \end{split}$$

where we recall that  $[0,1]^{2m}_{<} = \{0 < s_1 < s_2 < \cdots < s_{2m} < 1\}$  and  $\mathcal{P}_{2m}$  is the set of all permutations on  $[\![2m]\!]$ . Here we use the fact  $\delta(x) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi x} d\xi$ . We remark that the integral appearing in (2.34) is defined as follows

$$\int_0^1 \int_0^1 |r-s|^{2H-2} \delta(X_r - X_s) \mathrm{d}r \mathrm{d}s := \lim_{\varepsilon \to 0} \int_0^1 \int_0^1 |r-s|^{2H-2} p_\varepsilon(X_r - X_s) \mathrm{d}r \mathrm{d}s,$$

where the limit is taken in  $L^2$ , and the computations in (2.35) can be made rigorous by a limiting argument (see, e.g., the proof of [42, Theorem 4.1]).

Let  $\sigma \in \mathcal{P}_{2m}$  be arbitrarily chosen and fixed. Consider  $(s_1, \ldots, s_{2m}) \in [0, 1]_{<}^{2m}$ , and for each pair  $(s_{\sigma(2i-1)}, s_{\sigma(2i)})$ , we let  $t_{2i}^* \stackrel{\text{def}}{=} s_{\sigma(2i-1)} \lor s_{\sigma(2i)}$  and  $t_{2i-1}^*$  be the unique  $s_j$  which is the closest point to  $t_{2i}^*$  from the left. Then clearly, we have

$$\prod_{i=1}^{m} |s_{\sigma(2i)} - s_{\sigma(2i-1)}|^{2H-2} \le \prod_{i=1}^{m} |t_{2i}^* - t_{2i-1}^*|^{2H-2}.$$

Noting that

$$\sum_{i=1}^{m} \xi_i (X_{s_{\sigma(2i)}} - X_{s_{\sigma(2i-1)}}) = \sum_{j=2}^{2m} \eta_j (X_{s_j} - X_{s_{j-1}})$$

where each  $\eta_j$  is a linear combination of  $\xi_i$ 's for i = 1, ..., m. Then we have

$$\mathbb{E}\left[\mathrm{e}^{i\sum_{i=1}^{m}\xi_{i}(X_{s_{\sigma}(2i)}-X_{s_{\sigma}(2i-1)})}\right] = \prod_{j=2}^{2m}\mathrm{e}^{-c_{\rho}|\eta_{j}|^{\rho}(s_{j}-s_{j-1})} \le \prod_{i=1}^{m}\mathrm{e}^{-c_{\rho}|\tilde{\eta}_{i}|^{\rho}(t_{2i}^{*}-t_{2i-1}^{*})},$$

where the inequality holds because we only keep the factors resulting from the characteristic function of  $X_{t_{2i}^*} - X_{t_{2i-1}^*}$  for  $i \in [m]$  and drop all the others. Here, for each  $i \in [m]$ , we define  $\tilde{\eta}_i = \eta_j$  where j is the unique index such that  $s_j = t_{2i}^* = s_{\sigma(2i)} \vee s_{\sigma(2i-1)}$ .

Thus, we have

$$(2.36) \qquad \int_{[0,1]_{<}^{2m}} \int_{\mathbb{R}^{m}} \prod_{i=1}^{m} |s_{\sigma(2i)} - s_{\sigma(2i-1)}|^{2H-2} \mathbb{E} \left[ e^{i \sum_{i=1}^{m} \xi_{i}(X_{s_{\sigma(2i)}} - X_{s_{\sigma(2i-1)}})} \right] d\boldsymbol{\xi} d\boldsymbol{s}$$

$$\leq \int_{[0,1]_{<}^{2m}} \int_{\mathbb{R}^{m}} \prod_{i=1}^{m} |t_{2i}^{*} - t_{2i-1}^{*}|^{2H-2} \prod_{i=1}^{m} e^{-c_{\rho}|\tilde{\eta}_{i}|^{\rho}(t_{2i}^{*} - t_{2i-1}^{*})} d\boldsymbol{\xi} d\boldsymbol{s}$$

$$= C^{m} \int_{[0,1]_{<}^{2m}} \prod_{i=1}^{m} |t_{2i}^{*} - t_{2i-1}^{*}|^{2H-2-\frac{1}{\rho}} d\boldsymbol{s}$$

$$\leq \frac{C^{m}}{\Gamma\left(m(2H - \frac{1}{\rho}) + 1\right)}.$$

where the equality follows from the fact that the Jacobian determinant  $|[\partial \xi_i/\partial \tilde{\eta}_j]_{m \times m}|$  is one and that  $\int_{\mathbb{R}} e^{-a|\xi|^{\rho}t} d\xi = C a^{-1/\rho}$  for a > 0, and the last step follows from Lemma C.3.

Therefore, by (2.35) and (2.36), and noting that  $|\mathcal{P}_{2m}| = (2m)!$ , we have

$$\mathbb{E}\left[\exp\left(\beta \int_0^1 \int_0^1 |r-s|^{2H-2} \delta(X_r - X_s) \mathrm{d}r \mathrm{d}s\right)\right]$$
  
$$\leq \sum_{m=0}^\infty C^m \beta^m \frac{(2m)!}{m! \Gamma\left(m(2H - \frac{1}{\rho}) + 1\right)} < \infty,$$

where the last step follows from Stirling's formula and  $2H - \frac{1}{a} > 1$ .

**Remark 2.8.** In the proof of [42, Theorem 3.3], the Markov property instead of the independent increment property of X was invoked, and as a consequence, [42, Lemma 2.2] was needed therein. As shown in the proof of Proposition 2.7, if we utilise the independent increment property of X, [42, Lemma 2.2] is no longer needed, and this is important for the proof of uniform  $L^1$ -bound for rescaled partition function  $\mathbb{Z}^{(N)}_{\omega}(\hat{\beta}_N, k)$  in Section 3.3.

**Remark 2.9.** Note that when  $\rho = 2$ , the condition  $\theta > \frac{1}{2}$  in (1.15) becomes  $H > \frac{3}{4}$ . Thus, Proposition 2.7 is consistent with [23, Theorem 6.2].

**Proposition 2.10.** Under the condition (1.15), the mild Stratonovich solution given in (2.27) has the following Feynman-Kac representation:

(2.37) 
$$u(t,x) = \mathbb{E}_X \left[ \exp\left(\beta \int_0^t \int_{\mathbb{R}} \delta(X_{t-r}^x - y) W(\mathrm{d}r, \mathrm{d}y) \right) \right].$$

*Proof* First, we prove the integrability of the right-hand side of (2.37). For all  $\beta \in \mathbb{R}$ ,

(2.38)  

$$\mathbb{E}_{W}\left[\mathbb{E}_{X}\left[\exp\left(\beta\int_{0}^{t}\int_{\mathbb{R}}\delta(X_{t-r}^{x}-y)W(\mathrm{d}r,\mathrm{d}y)\right)\right]\right] \\
=\mathbb{E}_{X}\left[\mathbb{E}_{W}\left[\exp\left(\beta\int_{0}^{t}\int_{\mathbb{R}}\delta(X_{t-r}^{x}-y)W(\mathrm{d}r,\mathrm{d}y)\right)\right]\right] \\
=\mathbb{E}_{X}\left[\exp\left(\frac{1}{2}\beta^{2}t^{2H-\frac{1}{\rho}}\int_{0}^{1}\int_{0}^{1}|r-s|^{2H-2}\delta(X_{r}-X_{s})\mathrm{d}r\mathrm{d}s\right)\right] \\
<\infty,$$

where the second equality follows from (2.31) and the self-similarity of X, and the last step follows from Proposition 2.7. Thus u(t, x) given by (2.37) is well-defined and is  $L^p$ -integrable for all p > 0.

The Feynman-Kac formula (2.37) now follows from (2.27) and the following equation:

(2.39) 
$$\mathbb{E}_{X}\left[\left(\int_{0}^{t}\int_{\mathbb{R}}\delta(X_{t-r}^{x}-y)W(\mathrm{d}r,\mathrm{d}y)\right)^{m}\right]$$
$$=\int_{[0,t]^{m}}\int_{\mathbb{R}^{m}}\mathbb{E}_{X}\left[\prod_{i=1}^{m}\delta(X_{t-r_{i}}^{x}-y_{i})\right]W(\mathrm{d}r_{1},\mathrm{d}y_{1})\ldots W(\mathrm{d}r_{m},\mathrm{d}y_{m})$$
$$=m!\int_{[0,t]^{m}}\int_{\mathbb{R}^{m}}\mathfrak{g}_{m}(\boldsymbol{r},\boldsymbol{y};t,x)\prod_{i=1}^{m}W(\mathrm{d}r_{i},\mathrm{d}y_{i}),$$

where the first equality follows from (2.17) and the stochastic Fubini theorem (Proposition 2.4).

Remark 2.11. A direct corollary of Propositions 2.7 and 2.10 is

$$\|\mathfrak{g}_m\|_{\mathbf{S}_m} < \infty,$$

where  $\mathfrak{g}_m$  is given in (2.28) and  $\|\cdot\|_{\mathbf{S}_m}$  in (2.20). Indeed,

$$\|\mathfrak{g}_m\|_{\mathbf{S}_m}^2 = \mathbb{E}[|\mathbb{I}_m(\mathfrak{g}_m)|^2] = \left(\frac{1}{m!}\right)^2 \mathbb{E}_W\left[\left(\mathbb{E}_X[\mathcal{I}_{t,x}^m]\right)^2\right] \le \left(\frac{1}{m!}\right)^2 \mathbb{E}[\mathcal{I}_{t,x}^{2m}] < \infty,$$

where  $\mathcal{I}_{t,x}$  is given in (2.30), the second equality follows from (2.39), and the last inequality holds since  $\mathbb{E}[e^{\beta \mathcal{I}_{t,x}}] < \infty$  for all  $\beta \in \mathbb{R}$  by (2.38). Thus,  $\operatorname{Tr}^k \hat{\mathfrak{g}}_m \in \mathscr{H}^{\otimes (m-2k)}$  by (2.20). Moreover, by (2.39) we have

$$\sum_{m=1}^{\infty} \mathbb{E}[|\mathbb{I}_m(\mathfrak{g}_m)|] \le \sum_{m=0}^{\infty} \frac{1}{m!} \mathbb{E}[|\mathcal{I}_{t,x}|^m] = \mathbb{E}[\mathrm{e}^{|\mathcal{I}_{t,x}|}] \le 2\mathbb{E}[\mathrm{e}^{\mathcal{I}_{t,x}}] < \infty.$$

This implies that  $\sum_{m=1}^{\infty} \mathbb{I}_m(\mathfrak{g}_m)$  converges in  $L^1$ , which is stated in the following proposition.

**Proposition 2.12.** Under the condition (1.15), the mild Stratonovich solution of (2.26) given in (2.27) converges in  $L^1$ .

2.3. Mild Itô-Skorohod solution. Mild Itô-Skorohod solution is defined similarly as for mild Stratonovich solution in Section 2.2, and the only difference is that the stochastic integral in (2.26) is now understood in the Skorohod sense. More precisely, still assuming  $\tilde{u}_0(x) \equiv 1$ , the Itô-Skorohod solution  $\tilde{u}(t, x)$  of (1.13) satisfies

(2.41) 
$$\tilde{u}(t,x) = 1 + \beta \int_0^t \int_{\mathbb{R}} \mathfrak{g}(t-r,x-y)\tilde{u}(r,y) \diamond W(\mathrm{d}r,\mathrm{d}y),$$

and its chaos expansion is given by

(2.42)  
$$\tilde{u}(t,x) = 1 + \sum_{m=1}^{\infty} \beta^m \int_{[0,t]^m} \int_{\mathbb{R}^m} \mathfrak{g}_m(t,x;t,x) W(\mathrm{d}t_1,\mathrm{d}x_1) \diamond \cdots \diamond W(\mathrm{d}t_m,\mathrm{d}x_m)$$
$$= 1 + \sum_{m=1}^{\infty} \beta^m \mathbf{I}_m(\mathfrak{g}_m(\cdot;t,x)),$$

where  $\mathfrak{g}_m$  is given in (2.28) and the stochastic integrals above are in the Skorohod sense. A sufficient and necessary condition for the existence and uniqueness of the Itô-Skorohod solution is (see, e.g., [22, 42]),

(2.43) 
$$\sum_{m=1}^{\infty} \beta^m m! \left\| \hat{\mathfrak{g}}_m(\cdot; t, x) \right\|_{\mathscr{H}^{\otimes m}}^2 < \infty,$$

where  $\hat{\mathfrak{g}}_m$  is the symmetrization of  $\mathfrak{g}_m$ :

(2.44) 
$$\hat{\mathfrak{g}}_m(\boldsymbol{t},\boldsymbol{x};t,x) = \frac{1}{m!} \sum_{\sigma \in \mathcal{P}_m} \mathfrak{g}_m(\boldsymbol{t}_\sigma,\boldsymbol{x}_\sigma;t,x),$$

with  $\boldsymbol{t}_{\sigma} \stackrel{\text{def}}{=} (t_{\sigma(1)}, \ldots, t_{\sigma(m)})$  and  $\boldsymbol{x}_{\sigma} \stackrel{\text{def}}{=} (x_{\sigma(1)}, \ldots, x_{\sigma(m)}).$ 

By [42, Theorem 5.3], there exists a unique mild Itô-Skorohod solution for (1.13) on  $\mathbb{R}^d$  if

(2.45) 
$$\int_{\mathbb{R}^d} \frac{1}{1+|\xi|^{\rho}} \mathrm{d}\xi < \infty \iff \rho > d.$$

**Remark 2.13.** Note that the requirement  $\rho \leq 2$  together with (2.45) forces us to consider the case d = 1 only. By (2.45), we also need to assume  $\rho > 1$  in Theorem 1.2.

We shall provide an alternative proof for (2.43) under the condition  $H \in (1/2, 1], \rho \in (1, 2]$ , which will be easier to be adapted to estimate the moments of the terms arising from the partition function  $\tilde{Z}_{\omega}^{(N)}$  (see Remark 2.16 below).

**Proposition 2.14.** Assume  $H \in (1/2, 1]$  and  $\theta = H - \frac{1}{2\rho} > 0$ . Then there exists a constant C such that for all  $(t, x) \in [0, 1] \times \mathbb{R}$ ,

(2.46) 
$$\|\hat{\mathfrak{g}}_m(\cdot;t,x)\|_{\mathscr{H}^{\otimes m}} \leq C^m (m!)^{H-1} \left(\frac{\Gamma(\theta/H)^m}{\Gamma(m\theta/H+1)}\right)^H,$$

where  $\hat{\mathfrak{g}}_m$  is given in (2.44). Moreover, (2.43) holds under the condition (1.18).

*Proof* Recalling (2.2), we have

$$\begin{aligned} \|\hat{\mathfrak{g}}_{m}(\cdot;t,x)\|_{\mathscr{H}^{\otimes m}}^{2} &= \int_{[0,1]^{2m}} \int_{\mathbb{R}^{m}} \prod_{i=1}^{m} |r_{i} - s_{i}|^{2H-2} \hat{\mathfrak{g}}_{m}(\boldsymbol{r},\boldsymbol{x};t,x) \hat{\mathfrak{g}}_{m}(\boldsymbol{s},\boldsymbol{x};t,x) \mathrm{d}\boldsymbol{x} \mathrm{d}\boldsymbol{r} \mathrm{d}\boldsymbol{s} \\ &\leq C_{H}^{m} \int_{\mathbb{R}^{m}} \left( \int_{[0,1]^{m}} |\hat{\mathfrak{g}}_{m}(\boldsymbol{s},\boldsymbol{x};t,x)|^{\frac{1}{H}} \mathrm{d}\boldsymbol{s} \right)^{2H} \mathrm{d}\boldsymbol{x} \\ &\leq C_{H}^{m} \left( \int_{[0,1]^{m}} \left( \int_{\mathbb{R}^{m}} |\hat{\mathfrak{g}}_{m}(\boldsymbol{s},\boldsymbol{x};t,x)|^{2} \mathrm{d}\boldsymbol{x} \right)^{\frac{1}{2H}} \mathrm{d}\boldsymbol{s} \right)^{2H} \end{aligned}$$

where the first inequality follows from Lemma C.2 and the second one follows from the Minkowski inequality. Recalling the definition (2.44) of  $\hat{\mathfrak{g}}_m$  and the definition (2.28) of  $\mathfrak{g}_m$ , we have

$$\begin{aligned} |\hat{\mathfrak{g}}_{m}(\boldsymbol{s}, \boldsymbol{x}; t, \boldsymbol{x})|^{2} &= \frac{1}{(m!)^{2}} \left| \sum_{\sigma \in \mathcal{P}_{m}} \mathfrak{g}_{m}(\boldsymbol{s}_{\sigma}, \boldsymbol{x}_{\sigma}; t, \boldsymbol{x}) \right|^{2} \\ &= \frac{1}{(m!)^{2}} \left| \sum_{\sigma \in \mathcal{P}_{m}} \prod_{i=1}^{m} \mathfrak{g}(s_{\sigma(i+1)} - s_{\sigma(i)}, x_{\sigma(i+1)} - x_{\sigma(i)}) \mathbf{1}_{[0,1]_{<}^{m}}(s_{\sigma(1)}, \dots, s_{\sigma(m)}) \right|^{2} \\ &= \frac{1}{(m!)^{2}} \sum_{\sigma \in \mathcal{P}_{m}} \prod_{i=1}^{m} \left| \mathfrak{g}(s_{\sigma(i+1)} - s_{\sigma(i)}, x_{\sigma(i+1)} - x_{\sigma(i)}) \right|^{2} \mathbf{1}_{[0,1]_{<}^{m}}(\boldsymbol{s}_{\sigma}), \end{aligned}$$

where we use the convention  $s_{\sigma(m+1)} = s, x_{\sigma(m+1)} = x$ , and hence

$$\begin{split} &\|\hat{\mathfrak{g}}_{m}(\cdot;t,x)\|_{\mathscr{H}^{\otimes m}}^{2} \\ &\leq C_{H}^{m} \left( \int_{[0,1]^{m}} (m!)^{-\frac{1}{H}} \left( \int_{\mathbb{R}^{m}} \sum_{\sigma \in \mathcal{P}_{m}} \prod_{i=1}^{m} \left| \mathfrak{g}(s_{\sigma(i+1)} - s_{\sigma(i)}, x_{\sigma(i+1)} - x_{\sigma(i)}) \right|^{2} \mathbf{1}_{[0,1]^{m}_{<}}(s_{\sigma}) \mathrm{d} \boldsymbol{x} \right)^{\frac{1}{2H}} \mathrm{d} \boldsymbol{s} \right)^{2H} \\ &= C_{H}^{m} \left( \int_{[0,1]^{m}_{<}} (m!)^{1-\frac{1}{H}} \left( \int_{\mathbb{R}^{m}} \prod_{i=1}^{m} |\mathfrak{g}(s_{i+1} - s_{i}, x_{i+1} - x_{i})|^{2} \mathrm{d} \boldsymbol{x} \right)^{\frac{1}{2H}} \mathrm{d} \boldsymbol{s} \right)^{2H}, \end{split}$$

Noting that by (1.9), we have that for all t > 0,

$$\int_{\mathbb{R}} |\mathfrak{g}(t,x)|^2 \mathrm{d}x \le Ct^{-\frac{1}{\rho}},$$

where C is a positive constant depending only on  $\rho$ . Combining the last two estimates together with Lemma C.3, we get the desired inequality (2.46).

Finally, by (2.46) and Stirling's formula we get that (2.43) holds if  $\rho > 1$ .

**Remark 2.15.** The scaling limit of long-range random walk in i.i.d. random environment was considered in [8]. In this situation, the condition  $\rho > 1$  is also a necessary condition for (2.43). Indeed, now we have

$$\begin{aligned} \|\hat{\mathfrak{g}}_{m}(\cdot;t,x)\|_{\mathscr{H}^{\otimes m}}^{2} &= \int_{[0,1]^{m}} \int_{\mathbb{R}^{m}} |\hat{\mathfrak{g}}_{m}(s,x;t,x)|^{2} \mathrm{d}x \mathrm{d}s \\ &= \frac{1}{m!} \int_{[0,1]_{<}^{m}} \int_{\mathbb{R}^{m}} \prod_{i=1}^{m} |\mathfrak{g}(s_{i+1}-s_{i},x_{i+1}-x_{i})|^{2} \mathrm{d}x \mathrm{d}s \end{aligned}$$

By the scaling property (1.8), we have

$$\int_{\mathbb{R}} |\mathfrak{g}(s,x)|^2 \mathrm{d}x = As^{-\frac{1}{\rho}},$$

where  $A = \int_{\mathbb{R}} |\mathfrak{g}(1, x)|^2 dx$  is finite by (1.9). Thus, we get

$$\|\hat{\mathfrak{g}}_{m}(\cdot;t,x)\|_{\mathscr{H}^{\otimes m}}^{2} = \frac{A^{m}}{m!} \int_{[0,1]_{<}^{m}} \prod_{i=1}^{m} (s_{i+1}-s_{i})^{-\frac{1}{\rho}} \mathrm{d}s_{i}$$

which is finite only if  $\rho > 1$ . One can also check that (2.43) holds by Stirling's formula if  $\rho > 1$ .

**Remark 2.16.** From the proof of Proposition 2.14, we can see that the estimation (2.46) still holds if we replace  $\mathfrak{g}(t,x)$  by its upper bound  $C(t||x||^{-\rho-1}) \wedge t^{-1/\rho}$  (see (1.9)). This fact together with the upper bound for  $P_n(k)$  given in (1.7) will be used to estimate the moments of  $\mathbf{S}_m^{(N)}$  for the Skorohod case in Section 3.3.

**Remark 2.17.** Assume  $H \in (\frac{1}{2}, 1]$  and  $\rho \in (0, 2]$ . For (1.13), by the analysis in this section (or by Theorem 4.6 and Theorem 5.3 in [42]), we know that the condition for the existence of a Stratonovich solution is  $\theta = H - \frac{1}{2\rho} > \frac{1}{2}$  and the condition for the Itô-Skorohod case is  $\rho > 1$ . Note that for the Stratonovich case,  $\theta > \frac{1}{2}$  yields  $\rho > \frac{1}{2H-1} \ge 1$  and moreover  $\theta > \frac{1}{2}, \rho \le 2$  implies  $H \ge \frac{3}{4}$ . Thus, it requires more restrictive condition for the existence of a Stratonovich solution. This is because the  $L^2$ -norm of  $\mathbb{I}_m(\mathfrak{g}_m)$  is strictly bigger than  $\mathbf{I}_m(\mathfrak{g}_m)$  due to the extra trace terms appearing in  $\|\mathfrak{g}_m\|_{\mathbf{S}_m}^2$  (see (2.20)).

#### 3. On weak convergences

In this section, we aim to prove the main results Theorems 1.1 and 1.2. Recall that  $\theta = H - \frac{1}{2\rho}$ . For  $N \in \mathbb{N}$ , under the scaling  $\beta \to \hat{\beta}_N = \beta N^{-\theta}$ , the rescaled partition function  $Z_{\omega}^{(N)}$  given by (1.1) is

(3.1) 
$$Z_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0) = \mathbb{E}_S\left[\exp\left(\hat{\beta}_N\sum_{n=1}^N\omega\left(n, S_n^{(N+1,k)}\right)\right)\right],$$

where  $x_0 \in \mathbb{R}$  is fixed such that  $S_{N+1}^{(N+1,k)} = N^{1/\rho} x_0 = k \in \mathbb{Z}$ . To simply the notation, we use S to denote the backward random walk  $S^{(N+1,k)}$ . Taylor expansion yields

(3.2) 
$$Z_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0) = \sum_{m=0}^{\infty} \frac{1}{m!} \mathbb{S}_m^{(N)},$$

where  $\mathbb{S}_m^{(N)}$  is the *m*th moment of the Hamiltonian (with a factor  $\hat{\beta}_N$ ) given by

(3.3) 
$$\mathbb{S}_m^{(N)} = \mathbb{E}_S\left[\left(\hat{\beta}_N \sum_{n=1}^N \omega(n, S_n)\right)^m\right] = \hat{\beta}_N^m \mathbb{E}_S\left[\sum_{n \in [\![N]\!]^m} \omega(n_1, S_{n_1}) \cdots \omega(n_m, S_{n_m})\right].$$

For every  $\boldsymbol{n} = (n_1, \ldots, n_m) \in [\![N]\!]^m$ , the components can be arranged in an increasing order and the resulted sequence is denoted by  $\boldsymbol{n}^* = (n_1^*, \cdots, n_m^*)$  with  $n_1^* \leq n_2^* \leq \cdots \leq n_m^*$ . For each  $\boldsymbol{n} \in [\![N]\!]^m$ , there is a permutation  $\sigma$  of  $[\![N]\!]$  such that  $n_i^* = n_{\sigma_i}$  for  $i = 1, \ldots, m$ . Denote

(3.4) 
$$P_{\mathbf{n}^*} = P_{\mathbf{n}^*}(N^{1/\rho}x_0; k_1, \dots, k_m) \\ = P_{n_2^* - n_1^*}(k_{\sigma_2} - k_{\sigma_1}) \cdots P_{n_m^* - n_{m-1}^*}(k_{\sigma_m} - k_{\sigma_{m-1}})P_{(N+1) - n_m^*}(N^{1/\rho}x_0 - k_{\sigma_m}),$$

where we use the convention  $P_0(0) = 1$  and  $P_0(k) = 0$  for  $k \neq 0$ . We remark that  $P_{n^*}$  is symmetric in its *m* arguments. The *m*th moment in (3.3) can be written as

(3.5) 
$$\mathbb{S}_{m}^{(N)} = \hat{\beta}_{N}^{m} \sum_{n_{1},\dots,n_{m} \in [\![N]\!]} \sum_{k_{1},\dots,k_{m} \in \mathbb{Z}} \omega(n_{1},k_{1}) \cdots \omega(n_{m},k_{m}) P_{\boldsymbol{n}^{*}}(N^{1/\rho}x_{0};k_{1},\dots,k_{m}),$$

recalling that S is a backward random walk with  $S_{N+1} = N^{1/\rho} x_0$ .

A similar calculation can be done for the partition function  $\tilde{\mathbb{Z}}^{(N)}_{\omega}$  given in (1.16). Note that the exponential is actually a Wick exponential of  $\beta \sum_{i=1}^{N} \omega(i, S_i)$  conditional on the random walk S.

Hence, applying (2.15), we get the following resemblance of (3.2) and (3.3) respectively,

(3.6) 
$$\tilde{Z}_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0) = \sum_{m=0}^{\infty} \frac{1}{m!} \mathbf{S}_m^{(N)},$$

where

$$\mathbf{S}_{m}^{(N)} = \mathbb{E}_{S} \left[ \left( \hat{\beta}_{N} \sum_{n=1}^{N} \omega(n, S_{n}) \right)^{\diamond m} \right]$$

$$= \hat{\beta}_{N}^{m} \mathbb{E}_{S} \left[ \sum_{\boldsymbol{n} \in \llbracket N \rrbracket^{m}} : \omega(n_{1}, S_{n_{1}}) \cdots \omega(n_{m}, S_{n_{m}}) : \right]$$

$$= \beta^{m} N^{-m(\theta + \frac{1}{\rho})} \sum_{n_{1}, \dots, n_{m} \in \llbracket N \rrbracket} \sum_{k_{1}, \dots, k_{m} \in \mathbb{Z}} : \omega(n_{1}, k_{1}) \cdots \omega(n_{m}, k_{m}) :$$

$$\times \left( N^{m/\rho} P_{\boldsymbol{n}^{*}}(N^{1/\rho} x_{0}; k_{1}, \dots, k_{m}) \right).$$

Now, we introduce the following general U-statistics  $\mathbf{I}_m^{(N)}(f)$  which will be proven to converge weakly to a multiple Wiener integral. For  $f \in \mathcal{B}^{\otimes m}$  where we recall that  $\mathcal{B}$  is given in (2.3), we denote

(3.8)  

$$\mathbf{I}_{m}^{(N)}(f) \stackrel{\text{def}}{=} N^{-m(\theta+\frac{1}{\rho})} \sum_{\substack{n_{1},\dots,n_{m} \in \llbracket N \rrbracket}} \sum_{k_{1},\dots,k_{m} \in \mathbb{Z}} \left[ :\omega(n_{1},k_{1})\cdots\omega(n_{m},k_{m}): \times \mathcal{A}_{N}(f) \left(\frac{n_{1}}{N},\frac{k_{1}}{N^{1/\rho}},\cdots,\frac{n_{m}}{N},\frac{k_{m}}{N^{1/\rho}}\right) \right]$$

$$\stackrel{\text{def}}{=} N^{-m(\theta+\frac{1}{\rho})} \sum_{\boldsymbol{n} \in \llbracket N \rrbracket^{m}} \sum_{\boldsymbol{k} \in \mathbb{Z}^{m}} :\omega_{\llbracket m \rrbracket}: \mathcal{A}_{N}(f)(\boldsymbol{t},\boldsymbol{x}),$$

where  $:\omega_{[m]}:\stackrel{\text{def}}{=}:\omega_1\cdots\omega_m:$  is the physical Wick product of  $\omega_i\stackrel{\text{def}}{=}\omega(n_i,k_i), \mathcal{A}_N(f)$  is given in (2.5), and

(3.9) 
$$\boldsymbol{t} = (t_1, \dots, t_m) \stackrel{\text{def}}{=} \boldsymbol{n}/N = (n_1/N, \dots, n_m/N),$$
$$\boldsymbol{x} = (x_1, \dots, x_m) \stackrel{\text{def}}{=} \boldsymbol{k}/N^{\frac{1}{\rho}} = (k_1/N^{\frac{1}{\rho}}, \dots, k_m/N^{\frac{1}{\rho}}).$$

In order to write  $\mathbf{S}_m^{(N)}$  defined in (3.7) in the form of  $\mathbf{I}_m^{(N)}(f)$  given in (3.8), it suffices to extend the domain of the following function of (t, x)

$$N^{m/\rho}P_{Nt}(N^{1/\rho}x_0;N^{1/\rho}x_1,\ldots,N^{1/\rho}x_m) := N^{m/\rho}P_{n^*}(N^{1/\rho}x_0;k_1,\ldots,k_m)$$

to the whole  $[0,1]^m_<\times \mathbb{R}^m$  in a natural way, i.e., we define

(3.10) 
$$\tilde{P}_m(\boldsymbol{t}, \boldsymbol{x}) \stackrel{\text{def}}{=} N^{m/\rho} P_{\boldsymbol{n}^*}(N^{1/\rho} x_0; k_1, \dots, k_m), \text{ if } (\boldsymbol{t}, \boldsymbol{x}) \in \left[\frac{\boldsymbol{n}^*}{N}, \frac{\boldsymbol{n}^* + \overrightarrow{1}_m}{N}\right) \times \left[\frac{\boldsymbol{k}}{N^{1/\rho}}, \frac{\boldsymbol{k} + \overrightarrow{1}_m}{N^{1/\rho}}\right],$$
  
In this way, we can write  $\mathbf{S}^{(N)} = \beta^m \mathbf{I}^{(N)}(\tilde{\boldsymbol{p}})$ 

In this way, we can write  $\mathbf{S}_m^{(N)} = \beta^m \mathbf{I}_m^{(N)}(\tilde{P}_m).$ 

The rest of this Section is organized as follows. In Section 3.1, we prove the joint weak convergence for *U*-statistics  $\mathbf{I}_m^{(N)}(f)$  defined by (3.8). The joint weak convergence of *m*th moment  $\mathbb{S}_m^{(N)}$  (resp.  $\mathbf{S}_m^{(N)}$ ) appearing in the rescaled partition function  $Z_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0)$  (resp.  $(\tilde{Z}_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0))$ ) is established in Section 3.2. The  $L^p$ -bounds of  $Z_{\omega}^{(N)}(\hat{\beta}_N, k)$  and  $\tilde{Z}_{\omega}^{(N)}(\hat{\beta}_N, k)$  are obtained in Section 3.3. 3.1. Weak convergence of U-statistics. We first introduce the following lemma which provides a uniform (in N) bound for the second moment of  $\mathbf{I}_m^{(N)}(f)$  with  $f \in \mathcal{B}^{\otimes m}$ , which will be used to prove Propositions 3.2 and 3.3.

**Lemma 3.1.** Assume the disorder  $\omega$  is given as in Section 1.2. Then, there exists a constant C such that for all  $f \in \mathcal{B}^{\otimes m}$  and  $m, N \in \mathbb{N}$ ,

(3.11) 
$$\mathbb{E}\left[\left(\mathbf{I}_{m}^{(N)}(f)\right)^{2}\right] \leq C^{m} m! \|\hat{f}\|_{\mathcal{B}^{\otimes m}}^{2},$$

where  $\hat{f}$  is the symmetrization of f.

*Proof* For m = 1, we have

$$\mathbb{E}\left[\left(\mathbf{I}_{1}^{(N)}(f)\right)^{2}\right] = N^{-2(\theta+1/\rho)} \sum_{n,n' \in \llbracket N \rrbracket} \sum_{k,k' \in \mathbb{Z}} \mathbb{E}[\omega(n,k)\omega(n',k')]\mathcal{A}_{N}(f)\left(\frac{n}{N},\frac{k}{N^{1/\rho}}\right) \mathcal{A}_{N}(f)\left(\frac{n'}{N},\frac{k'}{N^{1/\rho}}\right) \\
= N^{-2(\theta+1/\rho)} \sum_{n,n' \in \llbracket N \rrbracket} \sum_{k \in \mathbb{Z}} \gamma(n-n')\mathcal{A}_{N}(f)\left(\frac{n}{N},\frac{k}{N^{1/\rho}}\right) \mathcal{A}_{N}(f)\left(\frac{n'}{N},\frac{k}{N^{1/\rho}}\right).$$

For  $t \in \mathbb{R}$ , let

(3.13) 
$$\gamma_N(t) \stackrel{\text{def}}{=} N^{2-2H} \gamma\left([|t|N]\right).$$

Then, by (1.11) and (1.12), we have for all  $t \in \mathbb{R}$ 

(3.14) 
$$\gamma_N(t) \lesssim |t|^{2H-2} \text{ and } \lim_{N \to \infty} \gamma_N(t) = |t|^{2H-2}$$

Then, using the notations  $t = n/N, t' = n'/N, x = k/N^{1/\rho}$ , we have

$$\mathbb{E}\left[\left(\mathbf{I}_{1}^{(N)}(f)\right)^{2}\right] = N^{-2-1/\rho} \sum_{n,n' \in \llbracket N \rrbracket} \sum_{k \in \mathbb{Z}} N^{2-2H} \gamma\left(n-n'\right) \mathcal{A}_{N}(f)\left(t,x\right) \mathcal{A}_{N}(f)\left(t',x\right)$$

$$\lesssim \int_{0}^{1} \int_{0}^{1} \int_{\mathbb{R}} |t-t'|^{2H-2} \left|\mathcal{A}_{N}(f)(t,x)\mathcal{A}_{N}(f)(t',x)\right| dx dt dt'$$

$$= \|\mathcal{A}_{N}(f)\|_{\mathcal{B}}^{2} \lesssim \|f\|_{\mathcal{B}}^{2},$$
(3.15)

where the last inequality follows from Lemma 2.2.

Recall that  $\theta = H - \frac{1}{2\rho}$  given in (1.15). For general m,

(3.16) 
$$\mathbb{E}\left[\left(\mathbf{I}_{m}^{(N)}(f)\right)^{2}\right] = N^{-2m(\theta+1/\rho)} \sum_{\boldsymbol{n},\boldsymbol{n}'} \sum_{\boldsymbol{k},\boldsymbol{k}'} \mathbb{E}\left[:\omega_{\llbracket m \rrbracket}::\omega'_{\llbracket m \rrbracket}:\right] \mathcal{A}_{N}(f)(\boldsymbol{t},\boldsymbol{x})\mathcal{A}_{N}(f)(\boldsymbol{t}',\boldsymbol{x}'),$$

where  $:\omega_{\llbracket m \rrbracket} :=: \prod_{i=1}^{m} \omega(n_i, k_i):$  and similarly for  $:\omega'_{\llbracket m \rrbracket}:$ .

Recalling that we have assumed that  $\omega$  is Gaussian, then by (A.8) we have

$$\mathbb{E}\left[:\omega_{\llbracket m\rrbracket}::\omega'_{\llbracket m\rrbracket}:\right] = \mathbb{E}\left[:\prod_{i=1}^{m}\omega(n_i,k_i)::\prod_{i=1}^{m}\omega(n'_i,k'_i):\right] = \sum_{\sigma\in\mathcal{P}_m}\prod_{i=1}^{m}\gamma(n_i-n'_{\sigma(i)})\delta_{k_ik'_{\sigma(i)}}$$

where the summation  $\sum_{\sigma \in \mathcal{P}_m}$  is taken over the set  $\mathcal{P}_m$  of all the permutations of [m]. Note that by the symmetry of the summation  $\sum_{n,n'} \sum_{k,k'}$ , we have  $\mathbf{I}_m^{(N)}(f) = \mathbf{I}_m^{(N)}(\hat{f})$ . Thus, by the symmetry

of  $\mathcal{A}_N(\hat{f})$ , we have

$$\mathbb{E}\left[\left(\mathbf{I}_{m}^{(N)}(\hat{f})\right)^{2}\right] = m! N^{-2m(\theta+1/\rho)} \sum_{\boldsymbol{n},\boldsymbol{n}'} \sum_{\boldsymbol{k}} \left(\prod_{i=1}^{m} \gamma(n_{i}-n_{i}')\right) \mathcal{A}_{N}(\hat{f})(\boldsymbol{t},\boldsymbol{x}) \mathcal{A}_{N}(\hat{f})(\boldsymbol{t}',\boldsymbol{x}).$$

Then, similarly to the case m = 1, by Lemma 2.2 we can get  $\mathbb{E}\left[\left(\mathbf{I}_{m}^{(N)}(f)\right)^{2}\right] \leq C^{m} \|\hat{f}\|_{\mathcal{B}^{\otimes m}}^{2}$ .  $\Box$ 

Now, we are ready to prove the weak convergence for  $\mathbf{I}_{m}^{(N)}(f)$  defined in (3.8). We first prove the weak convergence for m = 1.

**Proposition 3.2.** Let  $f : [0,1] \times \mathbb{R} \to \mathbb{R}$  be a function in  $\mathcal{B}$ . Then,  $\mathbf{I}_1^{(N)}(f)$  converges weakly to the Wiener integral  $\mathbf{I}_1(f)$  as  $N \to \infty$ . Moreover, for any  $k \in \mathbb{N}$  and  $f_1, \ldots, f_k \in \mathcal{B}$ , we have the joint convergence in distribution:

(3.17) 
$$\left(\mathbf{I}_{1}^{(N)}(f_{1}),\ldots,\mathbf{I}_{1}^{(N)}(f_{k})\right) \xrightarrow{d} \left(\mathbf{I}_{1}(f_{1}),\ldots,\mathbf{I}_{1}(f_{k})\right), \text{ as } N \to \infty.$$

*Proof* First we consider an indicator function  $f = \mathbf{1}_{[s,t] \times [y,z]}$  with  $0 \le s \le t \le 1$  and  $y \le z$ . In this case,

$$\mathbf{I}_1^{(N)}(f) = N^{-\theta - 1/\rho} \sum_{n \in [Ns, Nt] \cap \mathbb{N}} \sum_{k \in [N^{1/\rho}y, N^{1/\rho}z] \cap \mathbb{Z}} \omega(n, k).$$

Clearly it has mean zero and the covariance is

$$\mathbb{E}\left[\left(\mathbf{I}_{1}^{(N)}(f)\right)^{2}\right] = N^{-2-1/\rho} \sum_{n,n' \in [Ns,Nt] \cap \mathbb{N}} \sum_{k \in [N^{1/\rho}y,N^{1/\rho}z] \cap \mathbb{Z}} \gamma_{N}(r-r'),$$

where  $\gamma_N$  is given by (3.13) and r = n/N, r' = n'/N. By (3.14) and the dominated convergence theorem, we get

(3.18) 
$$\lim_{N \to \infty} \mathbb{E}\left[ \left( \mathbf{I}_{1}^{(N)}(f) \right)^{2} \right] = \int_{s}^{t} \int_{s}^{t} \int_{y}^{z} |r - r'|^{2H-2} \mathrm{d}x \mathrm{d}r \mathrm{d}r' = \|f\|_{\mathscr{H}}^{2}$$

where we recall that the inner product  $\langle \cdot, \cdot \rangle_{\mathscr{H}}$  is given in (2.2). Thus, we have  $\mathbf{I}_1^{(N)}(f) \xrightarrow{d} \mathbf{I}_1(f)$  as  $N \to \infty$ , noting that  $\mathbf{I}_1^{(N)}(f), N \in \mathbb{N}$  and  $\mathbf{I}_1(f)$  are Gaussian random variables and that  $\mathbf{I}_1(f)$  has zero mean and variance  $\|f\|_{\mathscr{H}}^2$ .

Similarly, one can show that (3.18) holds for each simple function f, i.e., f is a linear combination of indicator functions. For a general function  $f \in \mathcal{B}$ , there exists an approximation sequence of simple functions  $\{f^{(n)}\}_{n\in\mathbb{N}}$  in  $\mathcal{B}$ . On the one hand,  $\mathbf{I}_1^{(N)}(f^{(n)})$  converges weakly to  $\mathbf{I}_1(f^{(n)})$  as  $N \to \infty$  for each fixed n. On the other hand, by Lemma 3.1, we have  $\mathbf{I}_1^{(N)}(f^{(n)}) \to \mathbf{I}_1^{(N)}(f)$ in  $L^2$  uniformly in N as  $n \to \infty$ . Combining the above convergences together with the obvious  $L^2$ -convergence of  $\mathbf{I}_1(f^{(n)}) \to \mathbf{I}_1(f)$  as  $n \to \infty$ , by Lemma B.1 we obtain the weak convergence  $\mathbf{I}_1^{(N)}(f) \stackrel{d}{\longrightarrow} \mathbf{I}_1(f)$  as  $N \to \infty$  for  $f \in \mathcal{B}$ . The argument can be presented in a diagram:

$$\mathbf{I}_{1}^{(N)}(f^{(n)}) \xrightarrow{\text{ in } L^{2}, \text{ uniformly in } N}{n \to \infty} \mathbf{I}_{1}^{(N)}(f)$$

$$d \bigvee_{N \to \infty} d \bigvee_{N \to \infty} d \bigvee_{N \to \infty}$$

$$\mathbf{I}_{1}(f^{(n)}) \xrightarrow{\text{ in } L^{2}}{n \to \infty} \mathbf{I}_{1}(f))$$

Finally, the joint weak convergence (3.17) follows from the linearity of  $\mathbf{I}_1^{(N)}$  and the Cramér-Wold theorem (see Theorem B.1).

The main result in this section is presented below.

**Proposition 3.3.** For each  $m \in \mathbb{N}$  and  $f \in \mathcal{B}^{\otimes m}$ , we have

(3.19) 
$$\mathbf{I}_m^{(N)}(f) \xrightarrow{d} \mathbf{I}_m(f), \text{ as } N \to \infty,$$

where  $\mathbf{I}_m^{(N)}(f)$  is given in (3.8) and  $\mathbf{I}_m(f)$  is an *m*th multiple Wiener integral. Moreover, for any  $k \in \mathbb{N}$  and  $f_1, \ldots, f_k$  with  $f_i \in \mathcal{B}^{\otimes l_i}$  with  $l_i \in \mathbb{N}$ , we have the joint convergence in distribution:

(3.20) 
$$\left(\mathbf{I}_{l_1}^{(N)}(f_1),\ldots,\mathbf{I}_{l_k}^{(N)}(f_k)\right) \xrightarrow{d} \left(\mathbf{I}_{l_1}(f_1),\ldots,\mathbf{I}_{l_k}(f_k)\right), \text{ as } N \to \infty.$$

*Proof* We first consider  $\mathbf{I}_m^{(N)}(f)$  for  $f = h^{\otimes m}$  with  $h \in \mathcal{B}$ . For such f, we have

$$\mathbf{I}_{m}^{(N)}(f) = N^{-m(\theta+1/\rho)} \sum_{n_{1},\dots,n_{m} \in \llbracket N \rrbracket} \sum_{k_{1},\dots,k_{m} \in \mathbb{Z}} : \omega_{\llbracket m \rrbracket} : \mathcal{A}_{N}(h)^{\otimes m} =: \left(\mathbf{I}_{1}^{(N)}(h)\right)^{m} :,$$

where the term on the right-hand side is a physical Wick product (see Section A). Note that  $\mathbf{I}_{1}^{(N)}(h)$  is a centered Gaussian random variable with variance bounded by  $\|h\|_{\mathcal{B}}^{2}$  up to a multiplicative constant by Lemma 3.1. Therefore, for any  $q \in \mathbb{N}$ , by the Gaussianity of  $\omega$  and (3.15), we have

$$\mathbb{E}\left[\left(\mathbf{I}_{1}^{(N)}(h)\right)^{2q}\right] \lesssim (2q-1)!! \|h\|_{\mathcal{B}}^{2q} < \infty.$$

Then we can apply Lemma B.3 together with Proposition 3.2 and get that, as  $N \to \infty$ ,

$$\mathbf{I}_m^{(N)}(f) =: \left(\mathbf{I}_1^{(N)}(h)\right)^m : \longrightarrow : \left(\mathbf{I}_1(h)\right)^m := \mathbf{I}_m(h^{\otimes m}) = \mathbf{I}_m(f).$$

To prove the result, it suffices to consider symmetric functions in  $\mathcal{B}^{\otimes m}$ . For any symmetric function  $f \in \mathcal{B}^{\otimes m}$ , one can find a sequence  $\{f^{(n)}\}_{n\in\mathbb{N}}$  which are the linear combinations of the functions of the form  $h^{\otimes m}$  such that  $f^{(n)} \to f$  in  $\mathcal{B}^{\otimes m}$ . By the preceding argument, for each  $f^{(n)}$ , we have  $\mathbf{I}_m^{(N)}(f^{(n)}) \stackrel{d}{\longrightarrow} \mathbf{I}_m(f^{(n)})$  as  $N \to \infty$ . Then similarly as in the proof of Proposition 3.2, we have, recalling Lemma 3.1,

$$\mathbf{I}_{m}^{(N)}(f^{(n)}) \xrightarrow{\text{ in } L^{2}, \text{ uniformly in } N}{n \to \infty} \mathbf{I}_{m}^{(N)}(f)$$

$$d \bigvee_{N \to \infty} \mathbf{I}_{m}(f^{(n)}) \xrightarrow{\text{ in } L^{2}}{n \to \infty} \mathbf{I}_{m}(f))$$

This together with Lemma B.1 yields  $\mathbf{I}_m^{(N)}(f) \xrightarrow{d} \mathbf{I}_m(f)$  as  $N \to \infty$  for symmetric  $f \in \mathcal{B}^{\otimes m}$ .

Now we prove (3.20). Similar to the preceding step, we first consider the case  $f_i = h_i^{\otimes l_i}$  with  $h_i \in \mathcal{B}, i = 1, \ldots, k$ , then Proposition 3.2 and Lemma B.2 yield that for all  $(a_1, \ldots, a_k) \in \mathbb{R}^k$ ,  $\sum_{i=1}^k a_i \mathbf{I}_{l_i}^{(N)}(f_i) \xrightarrow{d} \sum_{i=1}^k a_i \mathbf{I}_{l_i}(f_i)$  as  $N \to \infty$ , which implies (3.20). The general case follows from a limiting argument together with Lemma B.1.

3.2. Weak convergence of rescaled partition functions. In this section, we prove Theorem 1.1 for the Stratonovich case and Theorem 1.2 for the Itô-Skorohod case. As explained in Section 1.3, we shall focus on the proof for the Stratonovich case which also encompasses the essentials for the Itô-Skorohod case.

Recall that the so-called *m*th moment  $\mathbb{S}_m^{(N)}$  appearing in the rescaled partition function  $Z_{\omega}^{(N)}(\hat{\beta}_N, N^{1/\rho}x_0)$  has the expression (3.5). We first prove the joint weak convergence of moments.

**Proposition 3.4.** Assume condition (1.15). For each  $m \in \mathbb{N}$ , we have

(3.21) 
$$\frac{1}{m!} \mathbb{S}_m^{(N)} \xrightarrow{d} \beta^m \mathbb{I}_m(\mathfrak{g}_m), \ as \ N \to \infty,$$

where  $\mathfrak{g}_m = \mathfrak{g}_m(t, x; 1, x_0)$  is given in (2.28) and  $\mathbb{I}_m(\mathfrak{g}_m)$  is an mth multiple Stratonovich integral. Moreover, for any  $k \in \mathbb{N}$  and  $l_1, \ldots, l_k \in \mathbb{N}$ , we have the joint convergence in distribution:

(3.22) 
$$\left(\frac{1}{l_1!}\mathbb{S}_{l_1}^{(N)},\ldots,\frac{1}{l_k!}\mathbb{S}_{l_k}^{(N)}\right) \xrightarrow{d} \left(\beta^{l_1}\mathbb{I}_{l_1}(\mathfrak{g}_{l_1}),\ldots,\beta^{l_k}\mathbb{I}_{l_k}(\mathfrak{g}_{l_k})\right), as N \to \infty.$$

*Proof* We first prove (3.21). Note that  $\operatorname{Tr}^{k} \hat{\mathfrak{g}}_{m} \in \mathcal{H}^{\otimes (m-2k)}$  (see (2.18) for the definition of  $\operatorname{Tr}^{k} \hat{\mathfrak{g}}_{m}$ ) for  $k = 0, 1, \ldots, [\frac{m}{2}]$  (see Remark 2.11). Thus, by Proposition 3.3, we have (we assume *m* is an odd integer and the analysis for the case of an even number *m* is the same),

(3.23) 
$$\begin{pmatrix} \mathbf{I}_{1}^{(N)}\left(\mathrm{Tr}^{\left[\frac{m}{2}\right]}\hat{\mathfrak{g}}_{m}\right), \mathbf{I}_{3}^{(N)}\left(\mathrm{Tr}^{\left[\frac{m}{2}\right]-1}\hat{\mathfrak{g}}_{m}\right), \dots, \mathbf{I}_{m}^{(N)}\left(\hat{\mathfrak{g}}_{m}\right) \\ \xrightarrow{d} \left(\mathbf{I}_{1}\left(\mathrm{Tr}^{\left[\frac{m}{2}\right]}\hat{\mathfrak{g}}_{m}\right), \mathbf{I}_{3}\left(\mathrm{Tr}^{\left[\frac{m}{2}\right]-1}\hat{\mathfrak{g}}_{m}\right), \dots, \mathbf{I}_{m}\left(\hat{\mathfrak{g}}_{m}\right) \end{pmatrix}, \text{ as } N \to \infty.$$

Recalling (3.5), by the symmetry of the summation, we can write  $\mathbb{S}_m^{(N)}$  as

$$\mathbb{S}_m^{(N)} = \hat{\beta}_N^m \sum_{n_1, \dots, n_m \in \llbracket N \rrbracket} \sum_{k_1, \dots, k_m \in \mathbb{Z}} \omega_{\llbracket m \rrbracket} P_{n^*},$$

where  $\hat{\beta}_N = \beta N^{-\theta}$ ,  $P_{n^*}$  is given in (3.4) and  $\omega_{[m]} = \prod_{i=1}^m \omega_i$  with  $\omega_i = \omega(n_i, k_i)$ . It follows from (A.5) and (A.4) that

$$\omega_{\llbracket m \rrbracket} = \sum_{j=0}^{\left[\frac{m}{2}\right]} \sum_{\substack{B \subset \llbracket m \rrbracket\\|B|=m-2j}} : \omega_B : \mathbb{E} \left[ \omega_{\llbracket m \rrbracket \setminus B} \right].$$

Therefore,

$$(3.24)$$

$$\mathbb{S}_{m}^{(N)} = \hat{\beta}_{N}^{m} \sum_{\substack{n_{1},...,n_{m} \ k_{1},...,k_{m}}} \sum_{\substack{|B|=m-2j}} \sum_{\substack{B \subset \llbracket m \rrbracket \\ |B|=m-2j}} :\omega_{B} : \mathbb{E}\left[\omega_{\llbracket m \rrbracket \setminus B}\right] \right) P_{\boldsymbol{n}^{*}}$$

$$= \sum_{j=0}^{\lfloor \frac{m}{2} \rfloor} \sum_{\substack{B \subset \llbracket m \rrbracket \\ |B|=m-2j}} \left( \hat{\beta}_{N}^{m} \sum_{\substack{n_{1},...,n_{m} \ k_{1},...,k_{m}}} :\omega_{B} : \mathbb{E}\left[\omega_{\llbracket m \rrbracket \setminus B}\right] P_{\boldsymbol{n}^{*}} \right).$$

For each  $j = 0, 1, \ldots, [\frac{m}{2}]$ , all the terms in the summation  $\sum_{\substack{B \subset \llbracket m \rrbracket \\ |B| = m - 2j}}$  are equal by the symmetry. We fix  $j \in \{0, 1, \ldots, [\frac{m}{2}]\}$ . Note that there are in total  $\binom{m}{m-2j}$  subsets in  $\llbracket m \rrbracket$  with cardinality m - 2j. Without loss of generality, we assume  $B = \{2j + 1, \ldots, m\}$ . Then by (A.4), we have

(3.25) 
$$\mathbb{E}[\omega_{\llbracket m \rrbracket \setminus B}] = \mathbb{E}[\omega_{\llbracket 2j \rrbracket}] = \sum_{V} \prod_{\{\ell_1, \ell_2\} \in V} \mathbb{E}[\omega_{\ell_1} \omega_{\ell_2}],$$

where the sum  $\sum_{V}$  is taken over all pair partitions of [2j]. By the symmetry again, the summations

$$\sum_{n_1,\dots,n_m} \sum_{k_1,\dots,k_m} : \omega_B : \prod_{\{\ell_1,\ell_2\} \in V} \mathbb{E}[\omega_{\ell_1}\omega_{\ell_2}] P_{\boldsymbol{n}^*}$$

coincide with each other for all the partitions V, and note that there are in total (2j - 1)!! pair partitions of [2j]. Thus, we have the following equation:

$$\mathbb{S}_{m}^{(N)} = \sum_{j=0}^{\left[\frac{m}{2}\right]} \binom{m}{m-2j} (2j-1)!! \left[ \hat{\beta}_{N}^{m} \sum_{n_{1},\dots,n_{m}} \sum_{k_{1},\dots,k_{m}} :\omega_{B} : \prod_{\ell=1}^{j} \mathbb{E}[\omega_{2\ell-1}\omega_{2\ell}]P_{\boldsymbol{n}^{*}} \right]$$

$$(3.26) \qquad = \sum_{j=0}^{\left[\frac{m}{2}\right]} \frac{m!}{j!(m-2j)!2^{j}} \left[ \hat{\beta}_{N}^{m} \sum_{n_{1},\dots,n_{m}} \sum_{k_{1},\dots,k_{m}} :\omega_{B} : \prod_{\ell=1}^{j} \mathbb{E}[\omega_{2\ell-1}\omega_{2\ell}]P_{\boldsymbol{n}^{*}} \right],$$

where we recall  $B = \{2j + 1, ..., m\}.$ 

Compare (3.26) with Hu-Meyer's formula (2.19) and also note (3.23). It then follows easily that in order to prove (3.21), it suffices to show, setting  $\beta = 1$  without loss of generality and hence  $\hat{\beta}_N = N^{-\theta}$ ,

$$Y_{N} \stackrel{\text{def}}{=} \frac{1}{m!} N^{-m\theta} \sum_{n_{1},...,n_{m}} \sum_{k_{1},...,k_{m}} :\omega_{B} : \prod_{\ell=1}^{j} \mathbb{E}[\omega_{2\ell-1}\omega_{2\ell}]P_{\boldsymbol{n}^{*}} - \mathbf{I}_{m-2j}^{(N)}(\mathrm{Tr}^{j}\hat{\boldsymbol{\mathfrak{g}}}_{m})$$

$$= \frac{1}{m!} N^{-(m-2j)(\theta+\frac{1}{\rho})} \sum_{n_{2j+1},...,n_{m}} \sum_{k_{2j+1},...,k_{m}} :\omega_{B} :$$

$$\times \left( N^{\frac{m}{\rho}} N^{-2j(\theta+\frac{1}{\rho})} \sum_{n_{i}:i \in [\![2j]\!]} \sum_{k_{i}:i \in [\![2j]\!]} \prod_{\ell=1}^{j} \mathbb{E}[\omega_{2\ell-1}\omega_{2\ell}]P_{\boldsymbol{n}^{*}} \right) - \mathbf{I}_{m-2j}^{(N)}(\mathrm{Tr}^{j}\hat{\boldsymbol{\mathfrak{g}}}_{m})$$

$$= \mathbf{I}_{m-2j}^{(N)} \left( N^{-2j(\theta+\frac{1}{\rho})} \sum_{n_{i}:i \in [\![2j]\!]} \sum_{k_{i}:i \in [\![2j]\!]} \prod_{\ell=1}^{j} \mathbb{E}[\omega_{2\ell-1}\omega_{2\ell}] \left( \tilde{P}_{m}/m! \right) \right) - \mathbf{I}_{m-2j}^{(N)}(\mathrm{Tr}^{j}\hat{\boldsymbol{\mathfrak{g}}}_{m})$$

$$\to 0 \text{ in } L^{2} \text{ as } N \to \infty,$$

for all  $j = 0, 1, \ldots, [\frac{m}{2}]$ , where we recall that  $\mathbf{I}_{m-2j}^{(N)}(f)$  is defined in (3.8) and  $\tilde{P}_m$  is given in (3.10). In the rest of the proof, we may abuse the notation  $N^{m/\rho}P_{n^*}$  for  $\tilde{P}_m(t, x)$  where it is appropriate.

We prove (3.27) for j = 0, and the general case can be proved in a similar spirit. When j = 0, we have

$$Y_N = \mathbf{I}_m^{(N)} \left( \tilde{P}_m / m! - \hat{\mathfrak{g}}_m \right)$$

Denoting  $\mathfrak{D}_N = \mathfrak{D}_N(\boldsymbol{n}, \boldsymbol{k}) = \tilde{P}_m/m! - \mathcal{A}_N(\hat{\mathfrak{g}}_m)(\boldsymbol{n}/N, \boldsymbol{k}/N^{1/\rho}), \mathfrak{D}'_N = \mathfrak{D}_N(\boldsymbol{n}', \boldsymbol{k}')$  and  $\tilde{\mathfrak{D}}'_N = \mathfrak{D}_N(\boldsymbol{n}', \boldsymbol{k})$ , we have

$$(3.28) \qquad \mathbb{E}[Y_N^2] = N^{-2m(\theta + \frac{1}{\rho})} \sum_{\boldsymbol{n}, \boldsymbol{n}' \in [\![N]\!]^m} \sum_{\boldsymbol{k}, \boldsymbol{k}' \in \mathbb{Z}^m} \mathbb{E}\left[:\omega_{[\![m]\!]}::\omega'_{[\![m]\!]}:\right] \mathfrak{D}_N \mathfrak{D}'_N$$
$$= m! N^{-2m(\theta + \frac{1}{\rho})} \sum_{\boldsymbol{n}, \boldsymbol{n}' \in [\![N]\!]^m} \sum_{\boldsymbol{k} \in \mathbb{Z}^m} \prod_{i=1}^m \gamma(n_i - n'_i) \mathfrak{D}_N \tilde{\mathfrak{D}}'_N$$
$$\lesssim m! \int_{[0,1]^{2m} \times \mathbb{R}^m} \prod_{i=1}^m |t_i - t'_i|^{2H-2} \mathfrak{D}_N \tilde{\mathfrak{D}}'_N \mathrm{d}t \mathrm{d}t' \mathrm{d}t$$

where the second equality follows from (A.8) and we use the notations that link the discrete summations and continuum integrals:  $t_i = n_i/N$ ,  $t'_i = n'_i/N$  and  $x_i = k_i/N^{1/\rho}$ . In light of Lemma 2.2, we have

(3.29) 
$$\mathbb{E}[Y_N^2] \lesssim m! \int_{[0,1]^{2m} \times \mathbb{R}^m} \prod_{i=1}^m |t_i - t_i'|^{2H-2} \left| \mathbf{D}_N \tilde{\mathbf{D}}_N' \right| \mathrm{d}t \mathrm{d}t' \mathrm{d}t,$$

where  $\mathbf{D}_N = \mathbf{D}_N(\boldsymbol{n}, \boldsymbol{k}) = \tilde{P}_m/m! - \hat{\mathfrak{g}}_m(\boldsymbol{t}, \boldsymbol{x})$  and  $\tilde{\mathbf{D}}'_N = \mathbf{D}_N(\boldsymbol{n}', \boldsymbol{k})$ .

Inspired by [8], we decompose  $I \stackrel{\text{def}}{=} [0,1]^{2m} \times \mathbb{R}^m$  as  $I \cap (I_1 \cup I_2 \cup I_3)$ , where

$$I_1 \stackrel{\text{def}}{=} \bigcap_{i=1}^m \left( \left\{ |t_i - t'_i| > \varepsilon \right\} \cap \{ |x_i| < M \} \right),$$
$$I_2 \stackrel{\text{def}}{=} \bigcup_{i=1}^m \{ |t_i - t'_i| \le \varepsilon \}, \text{ and } I_3 \stackrel{\text{def}}{=} \bigcup_{i=1}^m \{ |x_i| \ge M \}$$

for some fixed  $\varepsilon, M > 0$ . On  $I \cap I_1$ ,  $\prod_{i=1}^m |t_i - t'_i|^{2H-2} |\mathbf{D}_N \tilde{\mathbf{D}}'_N|$  is uniformly bounded and converges to 0 by the local limit theorem, and hence the integral converges to 0 by the dominated convergence theorem. Then, we can get  $\lim_{N\to\infty} \mathbb{E}[Y_N^2] = 0$ , if we can show the integral on  $I \cap I_2$  (resp.  $I \cap I_3$ ) can be arbitrarily small if we choose  $\varepsilon$  sufficiently small (resp. M sufficiently large).

By (1.5) and (1.8), we have

(3.30) 
$$\begin{pmatrix} N^{m/\rho} P_{\boldsymbol{n}^*} \end{pmatrix} \begin{pmatrix} N^{m/\rho} P_{(\boldsymbol{n}')^*} \end{pmatrix} \lesssim \begin{pmatrix} (t_2^* - t_1^*) \cdots (t_m^* - t_{m-1}^*)(1 - t_m^*) \end{pmatrix}^{-1/\rho} N^{m/\rho} P_{(\boldsymbol{n}')^*}, \\ \hat{\mathfrak{g}}_m(\boldsymbol{t}, \boldsymbol{x}; 1, x_0) \hat{\mathfrak{g}}_m(\boldsymbol{t}', \boldsymbol{x}; 1, x_0) \lesssim \begin{pmatrix} (t_2^* - t_1^*) \cdots (t_m^* - t_{m-1}^*)(1 - t_m^*) \end{pmatrix}^{-1/\rho} \hat{\mathfrak{g}}_m(\boldsymbol{t}', \boldsymbol{x}; 1, x_0)$$

and similarly for the cross terms  $N^{m/\rho}P_{\mathbf{n}^*}\hat{\mathfrak{g}}_m(t', x)$  and  $N^{m/\rho}P_{(\mathbf{n}')^*}\hat{\mathfrak{g}}_m(t, x)$ , where  $t_i^* = t_{\sigma(i)}$  for some permutation  $\sigma$  on [m] such that  $t_1^* \leq t_2^* \leq \cdots \leq t_m^*$ . These inequalities and the fact for all  $t \in [0, 1]^m$ ,

$$\int_{\mathbb{R}^m} N^{m/\rho} P_{\boldsymbol{n}^*} \mathrm{d}\boldsymbol{x} = \int_{\mathbb{R}^m} (m!) \hat{\mathfrak{g}}_m \mathrm{d}\boldsymbol{x} = 1,$$

yield that  $\int_{I} \prod_{i=1}^{m} |t_i - t'_i|^{2H-2} |\mathbf{D}_N \tilde{\mathbf{D}}'_N| dt dt' dx$  is bounded uniformly in N by (up to a multiplicative constant)

(3.31) 
$$C_m \stackrel{\text{def}}{=} \frac{1}{m!} \int_{[0,1]^{2m}} \prod_{i=1}^m |t_i - t'_i|^{2H-2} \left( (t_2^* - t_1^*) \cdots (t_m^* - t_{m-1}^*) (1 - t_m^*) \right)^{-1/\rho} \mathrm{d}t \mathrm{d}t',$$

which is finite due to the condition 2H - 2 > -1 and  $-1/\rho > -1$ . This implies that the integral on  $I \cap I_2$  can be arbitrarily small by choosing  $\varepsilon$  sufficiently small. Similarly, for the integral on  $I \cap I_3$ ,

$$\begin{split} &\int_{I \cap I_3} \prod_{i=1}^m |t_i - t_i'|^{2H-2} |\mathbf{D}_N \tilde{\mathbf{D}}_N'| \mathrm{d}t \mathrm{d}t' \mathrm{d}x \\ &\lesssim C_m \left( P\left\{ \sup_{0 \le t \le 1} |X_t| \ge M \Big| X_1 = x_0 \right\} + P\left\{ \max_{0 \le m \le N} |S_m| \ge N^{1/\rho} M \Big| S_N = N^{1/\rho} x_0 \right\} \right). \end{split}$$

Noting that

$$P\left\{\max_{0\leq m\leq N}|S_m|\geq N^{1/\rho}M\Big|S_N=N^{1/\rho}x_0\right\}\to P\left\{\sup_{0\leq t\leq 1}|X_t|\geq M\Big|X_1=x_0\right\} \text{ as } N\to\infty,$$

we can make the integral on  $I \cap I_3$  as small uniformly in N as we want by choosing M sufficiently large. This proves (3.27) for j = 0.

For general  $j = 0, 1, \ldots, [\frac{m}{2}]$ , using the same argument leading to (3.29), we have

(3.32) 
$$\mathbb{E}[Y_N^2] \lesssim (m-2j)! \int_{[0,1]^{2(m-2j)} \times \mathbb{R}^{m-2j}} \prod_{i=1}^{m-2j} |t_i - t_i'|^{2H-2} \left| \mathbf{D}_N \tilde{\mathbf{D}}_N' \right| \mathrm{d}t \mathrm{d}t' \mathrm{d}x,$$

where now

(3.33) 
$$\mathbf{D}_{N} = \mathbf{D}_{N}(\boldsymbol{n},\boldsymbol{k})$$
$$= N^{-2j\theta} N^{(m-2j)\frac{1}{\rho}} \sum_{n_{i}:i \in [[2j]]} \sum_{k_{i}:i \in [[2j]]} \prod_{\ell=1}^{j} \mathbb{E}[\omega_{2\ell-1}\omega_{2\ell}](P_{\boldsymbol{n}^{*}}/m!) - \mathrm{Tr}^{j}\hat{\mathfrak{g}}_{m}$$
$$\stackrel{\mathrm{def}}{=} \mathcal{Q}_{N} - \mathrm{Tr}^{j}\hat{\mathfrak{g}}_{m}$$

and  $\mathbf{D}'_N = \mathbf{D}_N(\mathbf{n}', \mathbf{k})$ . Then, we estimate the right-hand side of (3.32) in a similar way as for the case j = 0, i.e., we split  $[0, 1]^{2(m-2j)} \times \mathbb{R}^{m-2j}$  as the union of its restrictions on  $I_1, I_2$  and  $I_3$  and then analyse the integral restricted on  $I_i$ 's separately for i = 1, 2, 3. The analysis for the integral on  $I_1$  is the same as for j = 0. To argue that the integral on  $I_2$  and  $I_3$  can be arbitrarily small if we choose  $\varepsilon$  sufficiently small and M sufficiently large, it suffices to find a uniform (in N) upper bound for

(3.34) 
$$A \stackrel{\text{def}}{=} \int_{[0,1]^{2(m-2j)} \times \mathbb{R}^{m-2j}} \prod_{i=1}^{m-2j} |t_i - t'_i|^{2H-2} \operatorname{Tr}^j \hat{\mathfrak{g}}_m(\boldsymbol{t}, \boldsymbol{x}; 1, x_0) \operatorname{Tr}^j \hat{\mathfrak{g}}_m(\boldsymbol{t}', \boldsymbol{x}; 1, x_0) \mathrm{d}\boldsymbol{t} \mathrm{d}\boldsymbol{t}' \mathrm{d}\boldsymbol{x}$$

and

(3.35) 
$$B \stackrel{\text{def}}{=} \int_{[0,1]^{2(m-2j)} \times \mathbb{R}^{m-2j}} \prod_{i=1}^{m-2j} |t_i - t_i'|^{2H-2} \mathcal{Q}_N \tilde{\mathcal{Q}}_N' \mathrm{d}t \mathrm{d}t' \mathrm{d}t'$$

where  $Q_N$  is given in (3.33) and  $\tilde{Q}'_N = Q_N(n', k)$ . This is true, noting that Remark 2.11 yields

$$A = \frac{1}{(m-2j)!} \mathbb{E}\left[\left|\mathbf{I}_{m-2j}(\mathrm{Tr}^{j}\hat{\mathbf{g}}_{m})\right|^{2}\right] \leq \mathbb{E}\left[\left|\mathbb{I}_{m}(\mathbf{g}_{m})\right|^{2}\right] < \infty$$

and similarly, recalling that  $\mathbb{S}_m^{(N)}$  is given in (3.5), we have  $B \leq \mathbb{E}[|\mathbb{S}_m^{(N)}|^2]$  which has a uniform upper bound by (3.39).

In the above, we have assumed m is an odd integer. If m is even, the analysis is the same except for the case  $j = \frac{m}{2}$ . In this case, we have

$$Y_N = N^{-m\theta} \sum_{n_i: i \in \llbracket m \rrbracket} \sum_{k_i: i \in \llbracket m \rrbracket} \prod_{\ell=1}^{m/2} \mathbb{E}[\omega_{2\ell-1}\omega_{2\ell}](P_{\boldsymbol{n}^*}/m!) - \mathrm{Tr}^{m/2}\hat{\mathfrak{g}}_m,$$

which is deterministic and converges to 0 by the local limit theorem and a similar argument proving  $\lim_{N\to\infty} \mathbb{E}[Y_N^2] = 0$  for j = 0.

Finally, one can prove the weak convergence for linear combinations of  $(\mathbb{S}_{l_1}^{(N)}, \ldots, \mathbb{S}_{l_k}^{(N)})$  in a similar way, and hence (3.22) holds due to Theorem B.1.

By the proof of Proposition 3.4, in particular the part proving (3.27) for j = 0 under the condition  $H \in (1/2, 1]$  and  $\rho \in (1, 2]$ , we can get a parallel result for the Itô-Skorohod case which is stated below. Recall that  $\mathbf{S}_m^{(N)}$  is given in (3.7).

**Proposition 3.5.** Assume  $H \in (1/2, 1], \rho \in (1, 2]$ . For each  $m \in \mathbb{N}$ , we have

$$\frac{1}{m!} \mathbf{S}_m^{(N)} \xrightarrow{d} \beta^m \mathbf{I}_m(\mathfrak{g}_m), \ as \ N \to \infty,$$

where  $\mathfrak{g}_m = \mathfrak{g}_m(t, x; 1, x_0)$  is given in (2.28) and  $\mathbf{I}_m(\mathfrak{g}_m)$  is an mth multiple Wiener integral. Moreover, for any  $k \in \mathbb{N}$  and  $l_1, \ldots, l_k \in \mathbb{N}$ , we have the joint convergence in distribution:

$$\left(\frac{1}{l_1!}\mathbf{S}_{l_1}^{(N)},\ldots,\frac{1}{l_k!}\mathbf{S}_{l_k}^{(N)}\right) \stackrel{d}{\longrightarrow} \left(\beta^{l_1}\mathbf{I}_{l_1}(\mathfrak{g}_{l_1}),\ldots,\beta^{l_k}\mathbf{I}_{l_k}(\mathfrak{g}_{l_k})\right), as N \to \infty$$

Now we are ready to prove our main results.

*Proof of Theorem 1.1.* By Proposition 3.4, we have for all  $M \in \mathbb{N}$ ,

$$Z_{\omega}^{(N,M)} \stackrel{\text{def}}{=} \sum_{m=0}^{M} \frac{1}{m!} \mathbb{S}_{m}^{(N)} \stackrel{d}{\longrightarrow} \mathcal{Z}^{(M)} \stackrel{\text{def}}{=} \sum_{m=0}^{M} \beta^{m} \mathbb{I}_{m}(\mathfrak{g}_{m}(\cdot; 1, x_{0})), \text{ as } N \to \infty.$$

Recalling Proposition 2.12 which yields the  $L^1$ -convergence of  $\mathcal{Z}^{(M)}$  to  $\mathcal{Z}$  as  $M \to \infty$ , we only need to show that  $Z^{(N,M)}_{\omega}$  converges to  $Z^{(N)}_{\omega} = Z^{(N)}_{\omega}(\hat{\beta}_N, N^{1/\rho}x_0)$  given in (3.2) in probability uniformly in N as  $M \to \infty$  by Lemma B.1. This follows from (3.40) obtained in Section 3.3.

*Proof of Theorem 1.2.* The proof is the same as that of Theorem 1.1 except that Proposition 3.4 and (3.40) are replaced by Proposition 3.5 and (3.41), respectively.

3.3.  $L^p$ -bounds of rescaled partition functions. In this section, we study the  $L^p$ -bounds of  $Z^{(N)}_{\omega}(\hat{\beta}_N, k)$  and  $\tilde{Z}^{(N)}_{\omega}(\hat{\beta}_N, k)$ . We first deal with the Stratonovich case to find  $L^1$ -bound of  $Z^{(N)}_{\omega}(\hat{\beta}_N, k)$ , and then as a consequence we obtain the  $L^2$ -bound for the Itô-Skorohod case.

Recalling that  $\omega$  is Gaussian, we have

$$\mathbb{E}\left[\exp\left(N^{-\theta}\sum_{n=1}^{N}\omega(n,S_{n})\right)\right]$$

$$=\mathbb{E}\left[\exp\left(N^{-\theta}\sum_{n=1}^{N}\sum_{k\in\mathbb{Z}}\omega(n,k)\mathbf{1}_{[S_{n}=k]}\right)\right]$$

$$=\mathbb{E}\left[\exp\left(\frac{1}{2}N^{-2\theta}\sum_{n,n'=1}^{N}\sum_{k,k'\in\mathbb{Z}}\gamma(n-n')\delta_{kk'}\mathbf{1}_{[S_{n}=k]}\mathbf{1}_{[S_{n'}=k']}\right)\right]$$

$$=\mathbb{E}\left[\exp\left(\frac{1}{2}N^{-2\theta}\sum_{n,n'=1}^{N}\gamma(n-n')\mathbf{1}_{[S_{n}=S_{n'}]}\right)\right].$$

In this section, we shall prove a discretised version of Proposition 2.7, i.e., to show that the above exponential moment is uniformly bounded. As a consequence, we get a discretised version of Proposition 2.12 (see eq. (3.40)).

Noting that  $S_n \in \mathbb{Z}$  and using the identity

$$\mathbf{1}_{[S_n=S_{n'}]} = \frac{1}{2\pi} \int_{-\pi}^{\pi} \mathrm{e}^{i(S_n-S_{n'})\lambda} \mathrm{d}\lambda,$$

we have

$$\mathbb{E}\left[\exp\left(\frac{1}{2}N^{-2\theta}\sum_{n,n'=1}^{N}\gamma(n-n')\mathbf{1}_{[S_{n}=S_{n'}]}\right)\right]$$
  
=  $\sum_{m=0}^{\infty}\frac{1}{m!}(4\pi)^{-m}N^{-2m\theta}\sum_{n_{1},\dots,n_{m}=1}^{N}\sum_{n'_{1},\dots,n'_{m}=1}^{N}\int_{[-\pi,\pi]^{m}}\prod_{i=1}^{m}\gamma(n_{i}-n'_{i})\mathbb{E}\left[e^{i\sum_{i=1}^{m}(S_{n_{i}}-S_{n'_{i}})\lambda_{i}}\right]d\lambda$ 

Using changes of variables  $n_i = Nt_i$ ,  $n'_i = Nt'_i$  and  $\lambda_i = u_i/N^{1/\rho}$  for i = 1, ..., m, we have for each term on the right-hand side of the above equation,

(3.36) 
$$A_{m} \stackrel{\text{def}}{=} \frac{1}{m!} (4\pi)^{-m} N^{-2m} \sum_{\boldsymbol{n},\boldsymbol{n}'} \prod_{i=1}^{m} \left[ N^{2-2H} \gamma(N(t_{i} - t_{i}')) \right] \\ \times \int_{[-\pi N^{1/\rho}, \pi N^{1/\rho}]^{m}} \mathbb{E} \left[ e^{i \sum_{i=1}^{m} (S_{n_{i}} - S_{n_{i}'}) u_{i}/N^{1/\rho}} \right] d\boldsymbol{u}.$$

Recall that  $S_n = S_0 + \sum_{j=1}^n Y_j$  and that  $\psi(u) = \mathbb{E}[e^{iuY_j}]$  is the characteristic function of  $Y_j$ . The 1-lattice distribution of  $Y_j$  implies that, for any  $\varepsilon > 0$ , one can find a positive constant c such that (see [38, eq. (5.14)])

(3.37) 
$$|\psi(u/N^{1/\rho})| \le e^{-c|u|^{\rho-\varepsilon}/N}, \text{ for } 1 \le |u| \le \pi N^{1/\rho}.$$

Thus, for  $n \in [N]$ , we have

(3.38)  
$$\begin{aligned} \int_{-\pi N^{1/\rho}}^{\pi N^{1/\rho}} \left| \psi(u/N^{1/\rho}) \right|^n \mathrm{d}u &\leq 2 + \int_{\{1 \leq |u| \leq \pi N^{1/\rho}\}} \left| \psi(u/N^{1/\rho}) \right|^n \mathrm{d}u \\ &\leq 2 + \int_{\mathbb{R}} \mathrm{e}^{-c|u|^{\rho-\varepsilon}n/N} \mathrm{d}u \\ &\leq C(n/N)^{-\frac{1}{\rho-\varepsilon}}. \end{aligned}$$

Under the condition (1.15), we may choose  $\varepsilon > 0$  sufficiently small such that  $2H - \frac{1}{\rho - \varepsilon} > 1$ . To estimate  $A_m$  given in (3.36), we combine the estimate (3.38) with the argument used in the proof of Proposition 2.7 which leads to (2.36), and we can get for all  $N \in \mathbb{N}$ ,

$$A_{m} \leq \frac{(2m)!}{m!} C^{m} \int_{[0,1]_{<}^{2m}} \prod_{i=1}^{m} |t_{2i}^{*} - t_{2i-1}^{*}|^{2H-2-\frac{1}{\rho-\varepsilon}} \mathrm{d}\boldsymbol{s}$$
$$\leq \frac{(2m)! C^{m}}{m! \Gamma\left(m(2H-\frac{1}{\rho-\varepsilon})+1\right)}$$

Therefore, we have uniformly in N,

$$\sum_{m=0}^{\infty} A_m \le \sum_{m=0}^{\infty} \frac{(2m)!C^m}{m!\Gamma\left(m(2H - \frac{1}{\rho - \varepsilon}) + 1\right)} < \infty,$$

where the finiteness follows from Stirling's formula and  $2H - \frac{1}{\rho - \varepsilon} > 1$ . This implies

(3.39) 
$$\sup_{N \in \mathbb{N}} \mathbb{E}\left[\exp\left(N^{-\theta} \sum_{n=1}^{N} \omega(n, S_n)\right)\right] < \infty.$$

Finally, the above analysis also yields

(3.40) 
$$\lim_{M \to \infty} \sup_{N \in \mathbb{N}} \sum_{m=M+1}^{\infty} \mathbb{E} \left[ \frac{1}{m!} \left| N^{-\theta} \sum_{n=1}^{N} \omega(n, S_n) \right|^m \right] = 0,$$

which shall be used to prove the weak convergence of rescaled partition functions.

Now we consider the Itô-Skorohod case under the condition (1.18). Recall the partition function  $\tilde{Z}_{\omega}^{(N)}$  in (3.6). Noting that  $\mathbf{S}_{m}^{(N)}$  and  $\mathbf{S}_{n}^{(N)}$  given in (3.7) are orthogonal in  $L^{2}(\Omega)$  if  $m \neq n$ , we have

$$\mathbb{E}\left[\left|\sum_{m=0}^{\infty} \frac{1}{m!} \mathbf{S}_{m}^{(N)}\right|^{2}\right] = \sum_{m=0}^{\infty} \frac{1}{(m!)^{2}} \mathbb{E}\left[\left|\mathbf{S}_{m}^{(N)}\right|^{2}\right].$$

Applying (A.8), one can calculate  $\frac{1}{(m!)^2} \mathbb{E}\left[\left|\mathbf{S}_m^{(N)}\right|^2\right]$  to get the same upper bound uniformly in N (up to a multiplicative constant  $C^m$ ) as  $m! \|\hat{\mathfrak{g}}_m(\cdot;t,x)\|_{\mathscr{H}^{\otimes m}}^2$  (see (2.46) in Proposition 2.14 and see also Remark 2.16). Thus, assuming (1.18) we have

(3.41) 
$$\lim_{M \to \infty} \sup_{N \in \mathbb{N}} \mathbb{E} \left[ \left| \sum_{m=M+1}^{\infty} \frac{1}{m!} \mathbf{S}_m^{(N)} \right|^2 \right] = 0.$$

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### APPENDIX A. PHYSICAL WICK PRODUCT

In order to expand the partition function (1.1) in a proper way to obtain its weak convergence, we shall invoke the notion of *physical* Wick product for general random variables (in contrast to the *probabilistic* Wick product defined via Wiener chaos in Section 2.1). The physical Wick product (also known as Wick power, Wick polynomial or Wick renormalization) was introduced by Wick [45] in the study of quantum field theory. We collect some facts on physical Wick products in this subsection. The reader is referred to [2, 3, 16, 17, 31, 43] for more readable account.

Let  $\{X_i\}_{i\in\mathbb{N}}$  be a family of real random variables with finite moments of all orders. The physical Wick product  $:X_1\cdots X_n$ : is defined recursively as follows. For n = 0, we set ::= 1, and for  $n \ge 1$ ,

$$\frac{\partial}{\partial X_i} : X_1 \cdots X_n :=: X_1 \cdots \hat{X}_i \cdots X_n :, \quad \mathbb{E} \left[: X_1 \cdots X_n :\right] = 0,$$

where  $\hat{X}_i$  means the absence of  $X_i$  in the product. For example,

$$:X_1:=X_1 - \mathbb{E}[X_1];$$
  
$$:X_1X_2:=X_1X_2 - X_1\mathbb{E}[X_2] - X_2\mathbb{E}[X_1] + 2\mathbb{E}[X_1]\mathbb{E}[X_2] - \mathbb{E}[X_1X_2]$$

We remark that different indices may refer to the same random variable. In this situation, as an example, we also write  $:X^nY^m:\stackrel{\text{def}}{=}:X_1\cdots X_nY_1\cdots Y_m:$ , if  $X_1 = \cdots = X_n = X$  and  $Y_1 = \cdots = Y_m = Y$ .

If we assume  $\mathbb{E}\left[e^{\beta \sum_{i=1}^{n} |X_i|}\right] < \infty$  for some  $\beta > 0$ , the physical Wick product can be equivalently defined by

(A.1) 
$$: X_1 \cdots X_n : \stackrel{\text{def}}{=} \left. \frac{\partial^n}{\partial z_1 \cdots \partial z_n} G(z_1, \cdots, z_n; X_1, \cdots, X_n) \right|_{z=0},$$

where

$$G(z_1, \cdots, z_n; X_1, \cdots, X_n) \stackrel{\text{def}}{=} \frac{e^{\sum_{i=1}^n z_i X_i}}{\mathbb{E}\left[e^{\sum_{i=1}^n z_i X_i}\right]}$$

is called the generating function (or Wick exponential). If  $\{X_i\}_{i\in\mathbb{N}}$  is a centred Gaussian family, the generating function is simply  $G(z, \mathbf{X}) = e^{z \cdot \mathbf{X} - z \cdot Qz/2}$  where  $Q = (\mathbb{E}[X_i X_j])_{n \times n}$  is the covariance matrix of  $\mathbf{X} = (X_1, \dots, X_n)$ , and the resulting Wick products are related to Hermite polynomials (see (2.8) in Section 2.1). In particular, for a Gaussian random vector  $(X_1, \ldots, X_n)$ , the *physical* Wick product :  $X_1 \cdots X_n$ : coincides with *probabilistic* Wick product  $X_1 \diamond X_2 \diamond \cdots \diamond X_n$  defined in Section 2.1.

We collect some basic properties of physical Wick products. Clearly  $:X_1X_2...X_n$ : only involves the random variables  $X_1, ..., X_n$  and their joint moments up to order n, and it is symmetric and multilinear in  $(X_1, ..., X_n)$  (multilinear means linear in terms of each  $X_i, i = 1, ..., n$ ). If two groups of random variables  $\{X_1, X_2, ..., X_m\}$  and  $\{X_{m+1}, ..., X_n\}$  are independent of each other, then  $: X_1 \cdots X_n :=: X_1 \cdots X_m :: X_{m+1} \cdots X_n :$  (see [2, eq. (2.4)]). We remind that the physical Wick product is no longer associative, which is different from the ordinary product. For instance, assuming  $\mathbb{E}[X] = 0$ , we have  $: XXX := X^3 - 3X\mathbb{E}[X^2] - \mathbb{E}[X^3]$  which is different from  $:XY: |_{Y=:X^2:} = X^3 - X\mathbb{E}[X^2] - \mathbb{E}[X^3]$ . Thus a physical Wick product is a single term whose value is determined by the definition, and cannot be viewed as a composition of two (or several) physical Wick products. For instance, all the physical Wick products  $:X^3:::X^2X:$  and  $:XX^2:$  mean the same :XXX:, and in particular  $:XX^2:\neq:XY: |_{Y=:X^2:}$ .

For any finite index set  $A = \{i_1, \ldots, i_n\} \subset \mathbb{N}$ , we denote  $: X_A :\stackrel{\text{def}}{=} : X_{i_1} \cdots X_{i_n} :$  and similarly, we take the notation  $X_A \stackrel{\text{def}}{=} \prod_{i \in A} X_i$  for the ordinary product. We use  $X^A$  to denote the set  $\{X_i, i \in A\}$  of random variables.

We recall some facts about cumulants. Let  $\kappa(X^A)$  denote the joint cumulant of  $X^A = \{X_i, i \in A\}$ . Then

(A.2) 
$$\kappa(X^A) = \sum_{V} (|V| - 1)! (-1)^{|V| - 1} \prod_{i=1}^{|V|} \mathbb{E}[X_{V_i}]$$

and

(A.3) 
$$\mathbb{E}[X_A] = \sum_V \prod_{i=1}^{|V|} \kappa(X^{V_i}),$$

where the sum  $\sum_{V}$  is taken over all partitions  $V = \{V_1, \ldots, V_k\}, k \ge 1$  of A, and |V| = k is the partition number. For instance,

$$\kappa(X_1) = \mathbb{E}[X_1], \ \kappa(X_1, X_2) = \mathbb{E}[X_1 X_2] - \mathbb{E}[X_1]\mathbb{E}[X_2],$$

and

$$\kappa(X_1, X_2, X_3) = \mathbb{E}[X_1 X_2 X_3] - \mathbb{E}[X_1 X_2] \mathbb{E}[X_3] - \mathbb{E}[X_1 X_3] \mathbb{E}[X_2] - \mathbb{E}[X_2 X_3] \mathbb{E}[X_1] + 2\mathbb{E}[X_1] \mathbb{E}[X_2] \mathbb{E}[X_3].$$

If  $\{X_i\}_{i\in\mathbb{N}}$  is *Gaussian*, we have  $\kappa(X^A) = 0$  if  $|A| \ge 3$ . If we assume further  $\mathbb{E}[X_i] = 0$ , the formula (A.3) reduces to the Wick's theorem:

(A.4) 
$$\mathbb{E}[X_A] = \begin{cases} 0, & \text{if } |A| \text{ is odd,} \\ \sum_{V} \prod_{\{i,j\} \in V} \mathbb{E}[X_i X_j], & \text{if } |A| \text{ is even,} \end{cases}$$

where the summation  $\sum_{V}$  is taken over all *pair partitions*  $V = \{V_1, \ldots, V_{|A|/2}\}$  of A.

Ordinary products and physical Wick products are connected by the following formula (see [43, Prop. 1] or [3, Appendix B]),

(A.5) 
$$X_A = \sum_{B \subset A} : X_B : \sum_V \prod_{i=1}^{|V|} \kappa(X^{V_i}) = \sum_{B \subset A} : X_B : \mathbb{E}[X_{A \setminus B}],$$

and

(A.6) 
$$: X_A := \sum_{B \subset A} X_B \sum_{V} (-1)^{|V|} \prod_{i=1}^{|V|} \kappa(X^{V_i}),$$

where the sum  $\sum_{B \subset A}$  is taken over all subsets  $B \subset A$  including  $B = \emptyset$ , and the sum  $\sum_{V}$  is over all partitions  $V = \{V_1, \ldots, V_k\}, k \ge 1$  of the set  $A \setminus B$ . We use the convention  $X_{\emptyset} =: X_{\emptyset} := \kappa(X_{\emptyset}) = 1$ .

The following formula (see [3, Appendix B] or [18, Lemma 4.5]) will be used

(A.7) 
$$\mathbb{E}[:X_A::X_B:] = \sum_V \prod_{i=1}^{|V|} \kappa(X^{V_i}),$$

where the summation  $\sum_{V}$  is taken over all partitions  $V = \{V_1, \ldots, V_k\}, k \ge 1$  of  $A \cup B$  satisfying  $V_i \cap A \neq \emptyset \neq V_i \cap B$  for each  $V_i$ . In particular, if we assume  $\{X_i\}_{i \in \mathbb{N}}$  is a centered *Gaussian* family, equations (A.7) and (A.4) yield

(A.8) 
$$\mathbb{E}[:X_A::X_B:] = \begin{cases} 0, & \text{if } |A| \neq |B| \\ \sum_{V} \prod_{\{i,j\} \in V} \mathbb{E}[X_i X_j], & \text{if } |A| = |B|, \end{cases}$$

where the summation  $\sum_{V}$  is taken over all *pair partitions*  $V = \{V_1, \ldots, V_{|A|}\}$  of  $A \cup B$  such that  $V_k = \{i, j\}$  with  $i \in A, j \in B$  for  $k = 1, \ldots, |A|$ .

#### Appendix B. Some preliminaries on convergence of probability measures

The following result can be found in [7, Theorem 3.2, Chapter 1].

**Lemma B.1.** Consider random vectors  $Y_n^{(N)}, Y^{(N)}, Y_n$  and Y, such that  $Y_n^{(N)} \xrightarrow{d} Y_n$  as  $N \to \infty$ ,  $Y_n \xrightarrow{d} Y$  as  $n \to \infty$ , and  $Y_n^{(N)}$  converges in probability to  $Y^{(N)}$  uniformly in N as  $n \to \infty$ , then we have  $Y^{(N)} \xrightarrow{d} Y$  as  $N \to \infty$ . That is, assuming

$$\begin{array}{c|c} Y_n^{(N)} & \xrightarrow{\text{in probability, uniformly in } N} & Y^{(N)} \\ \hline & & & & \\ d & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & &$$

we have  $Y^{(N)} \xrightarrow{d} Y$  as  $N \to \infty$ .

Lemma B.2. Let  $k \ge 1$  be an integer. If  $(X_1^{(n)}, \ldots, X_k^{(n)}) \xrightarrow{d} (X_1, \ldots, X_k)$  as  $n \to \infty$ , we have  $f(X_1^{(n)}, \ldots, X_k^{(n)}) \xrightarrow{d} f(X_1, \ldots, X_k)$ 

for any continuous function f.

**Lemma B.3.** Let  $k \geq 1$  be an integer. Suppose  $(X_1^{(n)}, \ldots, X_k^{(n)}) \xrightarrow{d} (X_1, \ldots, X_k)$  as  $n \to \infty$ and assume that for any subset A of  $\{1, \ldots, k\}$ ,  $\{\prod_{i \in A} X_i^{(n)}\}_{n \in \mathbb{N}}$  is uniformly integrable. Then  $:X_1^{(n)} \cdots X_k^{(n)} : \xrightarrow{d} : X_1 \cdots X_k : as n \to \infty.$ 

*Proof* Lemma B.2 yields the weak convergence of the ordinary product  $\prod_{i \in A} X_i^{(n)} \xrightarrow{d} \prod_{i \in A} X_i$  for any  $A \subset \{1, \ldots, k\}$  as  $n \to \infty$ . By (A.6), it suffices to prove the convergence of the cumulants

$$\lim_{n \to \infty} \kappa(X_i^{(n)}, i \in A) = \kappa(X_i, i \in A).$$

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This follows from the Skorohod representation theorem, equation (A.2), and the assumption of uniform integrability for the products of  $X_1^{(n)}, \ldots, X_k^{(n)}$ .

**Theorem B.1.** [Cramér-Wold Theorem] As  $n \to \infty$ ,  $(X_1^{(n)}, \ldots, X_k^{(n)}) \xrightarrow{d} (X_1, \ldots, X_k)$  if and only if

$$\sum_{i=1}^{k} a_i X_i^{(n)} \xrightarrow{d} \sum_{i=1}^{k} a_i X_i, \text{ for all } (a_1, \dots, a_k) \in \mathbb{R}^k.$$

APPENDIX C. MISCELLANEOUS RESULTS

Let  $f : [0,1]^m \to \mathbb{R}$  be an integrable function. For a fixed  $N \in \mathbb{N}$ , let  $t_i = i/N$  for i = 0, 1, ..., Nand denote  $I_i = (t_{i-1}, t_i]$ . Define

(B.1) 
$$f_N(t_1,\ldots,t_m) \stackrel{\text{def}}{=} N^m \int_{I_{\ell_1} \times \cdots \times I_{\ell_m}} f(s_1,\ldots,s_m) \mathrm{d}\boldsymbol{s},$$

if  $(t_1, \ldots, t_m) \in I_{\ell_1} \times \cdots \times I_{\ell_m}$  for some  $\ell_i \in [\![N]\!]$ . That is,  $f_N$  is the conditional expectation of f with respect to the the  $\sigma$ -field generated by the rectangles of the form  $I_{\ell_1} \times \cdots \times I_{\ell_m}$ . Then we have the following inequality of Jensen type.

**Lemma C.1.** Assume  $H \in (1/2, 1]$  and suppose  $f, g : [0, 1]^m \to \mathbb{R}$  are integrable functions. Then, for each  $N \in \mathbb{N}$ ,

(B.2) 
$$\int_{[0,1]^{2m}} \prod_{i=1}^{m} |s_i - t_i|^{2H-2} |f_N(s_1, \dots, s_m)| |g_N(t_1, \dots, t_m)| \mathrm{d}\mathbf{s} \mathrm{d}\mathbf{t}$$
$$\leq C^m \int_{[0,1]^{2m}} \prod_{i=1}^{m} |s_i - t_i|^{2H-2} |f(s_1, \dots, s_m)| |g(t_1, \dots, t_m)| \mathrm{d}\mathbf{s} \mathrm{d}\mathbf{t},$$

for some constant C depending on H only.

*Proof* We only prove the case m = 1, and the other cases m > 1 follows from an induction argument together with Fubini's theorem. It suffices to prove (B.2) with m = 1 for nonnegative functions, i.e., for  $f, g \ge 0$ ,

(B.3) 
$$\int_{I_i \times I_j} |s-t|^{2H-2} f_N(s) g_N(t) \mathrm{d}s \mathrm{d}t \le C \int_{I_i \times I_j} |s-t|^{2H-2} f(s) g(t) \mathrm{d}s \mathrm{d}t, \ 1 \le i, j \le N,$$

where recalling  $I_i = (t_{i-1}, t_i]$ . We shall prove (B.3) for simple nonnegative functions. The general case can be proved by a limiting argument and thus is omitted.

For i = j, we assume that on  $I_i$ ,

$$f(s) = \sum_{\ell=1}^{k} a_{\ell} \mathbf{1}_{A_{\ell}}(s), \ g(t) = \sum_{\ell=1}^{k} b_{\ell} \mathbf{1}_{A_{\ell}}(t),$$

where  $A_{\ell} = (t_{i-1} + (\ell - 1)/Nk, t_{i-1} + \ell/Nk]$  with  $\ell = 1, \ldots, k$  form a uniform partition of the interval  $I_i$ , and  $a_{\ell}, b_{\ell}$  are nonnegative numbers. Then, denoting  $\bar{a} = \frac{1}{k} \sum_{\ell=1}^{k} a_{\ell}, \bar{b} = \frac{1}{k} \sum_{\ell=1}^{k} b_{\ell}$ ,

 $f_N(s) = \bar{a}, \ g_N(t) = \bar{b}, \ \text{on } I_i.$ 

The left-hand side of (B.3) is

(B.4) 
$$\bar{a}\bar{b}\int_{I_i \times I_i} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t = \sum_{\ell=1}^k \sum_{m=1}^k a_\ell b_m \left(\frac{1}{k^2} \int_{I_i \times I_i} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t\right),$$

and the right-hand side without the constant C is

(B.5) 
$$\sum_{\ell=1}^{k} \sum_{m=1}^{k} a_{\ell} b_m \left( \int_{A_{\ell} \times A_m} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t \right).$$

Then, (B.3) follows from (B.4), (B.5) and the following inequality

$$\frac{1}{k^2} \int_{I_i \times I_i} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t \le C \min_{\ell, m \in \llbracket k \rrbracket} \int_{A_\ell \times A_m} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t,$$

where C is a constant depending on H only. Indeed, by change of variables, the above inequality is equivalent to

$$\frac{1}{k^2} \int_0^1 \int_0^1 |s-t|^{2H-2} ds dt \le C \min_{\ell,m \in [\![k]\!]} \int_{(m-1)/k}^{m/k} \int_{(\ell-1)/k}^{\ell/k} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t,$$

which holds for  $C = \int_0^1 \int_0^1 |s-t|^{2H-2} ds dt$ , noting that for all  $\ell, m \in [\![k]\!]$ , we have  $|s-t| \le 1$  on  $[(\ell-1)/k, \ell/k] \times [(m-1)/k, m/k]$ , and hence  $\int_{(m-1)/k}^{m/k} \int_{(\ell-1)/k}^{\ell/k} |s-t|^{2H-2} ds dt \ge 1/k^2$ . This proves (B.3) for i = j.

Now we prove (B.3) for  $i \neq j$ . By symmetry, we only need to consider j > i. Let

$$f(s) = \sum_{\ell=1}^{k} a_{\ell} \mathbf{1}_{A_{\ell}}(s) \text{ on } I_i = (t_{i-1}, t_i],$$

and

$$g(t) = \sum_{\ell=1}^{k} b_{\ell} \mathbf{1}_{B_{\ell}}(t) \text{ on } I_{j} = (t_{j-1}, t_{j}],$$

where  $a_{\ell}, b_{\ell}$  are nonnegative numbers, and  $\{A_{\ell}, \ell = 1, \ldots, k\}$  and  $\{B_{\ell}, \ell = 1, \ldots, k\}$  are uniform partitions of  $I_i$  and  $I_j$ , respectively. Then,

$$f_N(s) = \bar{a}$$
 on  $I_i$  and  $g_N(t) = b$  on  $I_j$ ,

where  $\bar{a} = \frac{1}{k} \sum_{\ell=1}^{k} a_{\ell}$  and  $\bar{b} = \frac{1}{k} \sum_{\ell=1}^{k} b_{\ell}$ . The left-hand side of (B.3) is

(B.6) 
$$\bar{a}\bar{b}\int_{I_i \times I_j} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t = \sum_{\ell=1}^k \sum_{m=1}^k a_\ell b_m \left(\frac{1}{k^2} \int_{I_i \times I_j} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t\right),$$

and the right-hand side of (B.3) without C is

(B.7) 
$$\sum_{\ell=1}^{k} \sum_{m=1}^{k} a_{\ell} b_m \left( \int_{A_{\ell} \times B_m} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t \right).$$

To get (B.3), it suffices to show

(B.8) 
$$\frac{1}{k^2} \int_{I_i \times I_j} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t \le C \min_{\ell,m \in [\![k]\!]} \int_{A_\ell \times B_m} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t$$

for some constant C depending on H only. Note that by change of variables, (B.8) is equivalent to the following equality

(B.9) 
$$\frac{1}{k^2} \int_{j-i}^{j-i+1} \int_0^1 |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t \le C \min_{\ell,m \in \llbracket k \rrbracket} \int_{j-i+(m-1)/k}^{j-i+m/k} \int_{(\ell-1)/k}^{\ell/k} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t$$

We shall prove (B.9) for j = i + 1 and  $j \ge i + 2$  separately. If j = i + 1, (B.9) follows directly by choosing

$$C = 2^{2-2H} \int_{1}^{2} \int_{0}^{1} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t,$$

noting that

$$\int_{1+(m-1)/k}^{1+m/k} \int_{(\ell-1)/k}^{\ell/k} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t \ge 2^{2H-2}/k^2, \text{ for all } \ell, m \in [\![k]\!].$$

When  $n \stackrel{\text{def}}{=} j - i \ge 2$ , to prove (B.9), it suffices to prove that the biggest integral among the integrals over the regions  $\{(n + (m - 1)/k, n + m/k] \times ((\ell - 1)/k, \ell/k], \ell \in [\![k]\!]\}$  can be dominated by the smallest one, i.e.,

$$\int_{n}^{n+1/k} \int_{1-1/k}^{1} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t \le C \int_{n+1-1/k}^{n+1} \int_{0}^{1/k} |s-t|^{2H-2} \mathrm{d}s \mathrm{d}t$$

for some constant C only depending on H. This is true because there exists a finite constant depending on H only such that

$$|n-1|^{2H-2} \le 3^{2-2H}|n+1|^{2H-2}$$
 for all  $n \ge 2$ 

This proves (B.9), and hence completes the proof of (B.3) for  $i \neq j$ .

The following Hardy-Littlewood-Sobolev inequality is taken from [4, Lemma B.3] (see [33] for the one-dimensional version).

**Lemma C.2.** For  $H \in (\frac{1}{2}, 1]$ , the following inequality holds:

$$\int_{\mathbb{R}^m} \int_{\mathbb{R}^m} f(\boldsymbol{t}) f(\boldsymbol{s}) \prod_{i=1}^m |t_i - s_i|^{2H-2} \mathrm{d}\boldsymbol{t} \mathrm{d}\boldsymbol{s} \le C_H^m \left( \int_{\mathbb{R}^m} |f(\boldsymbol{t})|^{1/H} \mathrm{d}\boldsymbol{t} \right)^{2H},$$

where  $C_H > 0$  is a constant depending on H, and we denote  $\mathbf{t} = (t_1, \ldots, t_n)$  and  $\mathbf{s} = (s_1, \ldots, s_n)$ .

The following result can be calculated by a direct calculation.

**Lemma C.3.** Suppose  $\alpha_i < 1$  for i = 1, ..., m and let  $\alpha = \sum_{i=1}^m \alpha_i$ . Then

$$\int_{[0 < r_1 < \dots < r_m < r_{m+1} = t]} \prod_{i=1}^m (r_{i+1} - r_i)^{-\alpha_i} \mathrm{d}\boldsymbol{r} = \frac{\prod_{i=1}^m \Gamma(1 - \alpha_i)}{\Gamma(m - \alpha + 1)} t^{m - \alpha_i}$$

where  $\Gamma(x) = \int_0^\infty t^{x-1} e^{-t} dt$  is the Gamma function.

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