Quentin Berger & Hubert Lacoin
The continuum directed polymer in Lévy noise
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THE CONTINUUM DIRECTED POLYMER IN LÉVY NOISE

by Quentin Berger & Hubert Lacoin

Abstract. — We present in this paper the construction of a continuum directed polymer model in an environment given by space-time Lévy noise. One of the main objectives of this construction is to describe the scaling limit of a discrete directed polymer in a heavy-tail environment and for this reason we put special emphasis on the case of \( \alpha \)-stable noises with \( \alpha \in (1, 2) \). Our construction can be performed in arbitrary dimension, provided that the Lévy measure satisfies specific (and dimension dependent) conditions. We also discuss a few basic properties of the continuum polymer and the relation between this model and the stochastic heat equation with multiplicative Lévy noise.

Résumé (Polymère dirigé continu dans un bruit de Lévy). — Nous présentons dans cet article la construction d'un modèle de polymère dirigé continu dans un environnement donné par un bruit de Lévy en espace-temps. L'un des principaux objectifs de cette construction est de décrire la limite d'échelle d'un polymère dirigé discret dans un environnement à queue lourde et, pour cette raison, nous mettons plus particulièrement en avant le cas de bruits \( \alpha \)-stables avec \( \alpha \in (1, 2) \). Notre construction peut être réalisée en n'importe quelle dimension, pourvu que la mesure de Lévy satisfasse certaines conditions (qui dépendent de la dimension). Nous examinons aussi quelques propriétés de base du polymère continu, ainsi que la relation entre ce modèle et l'équation de la chaleur stochastique avec bruit multiplicatif de Lévy.

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1. Introduction

The aim of this paper is to build a continuum model which describes the scaling limit of directed polymers in $\mathbb{Z}^d$ with an environment which has infinite second moment: the continuum directed polymer in a space-time Lévy noise. Our construction can be thought as an extension to arbitrary noise and dimension of that presented in [2] of a continuum polymer in dimension 1 with Gaussian white noise. In a companion paper [9], we prove that the scaling limit of the directed polymer in $\mathbb{Z}^d$ with heavy tailed environment is indeed the continuum model constructed in the present paper.

Whereas the construction in [2] is directly based on the solution of the stochastic heat equation (SHE) with multiplicative noise, our approach here needs to be slightly different since the solution of SHE with a general Lévy noise (see [23] for recent developments) does not display sufficient regularity. Our continuum model is thus defined via a martingale approximation of the noise obtained by truncating the “small jumps” part of the noise. This construction is not specific to directed polymers and can possibly be applied to describe the scaling limit of a wide variety of disordered models with heavy tailed noise, including the disordered pinning model (see [17, 19] for the construction of the corresponding Gaussian scaling limits).

In order to motivate our construction, we provide a brief introduction to the directed polymer model, the notion of its scaling limit and review some literature on the subject.

1.1. Directed polymer in a random environment [the discrete model]

Let us consider $\eta = (\eta_{n,x})_{n \in \mathbb{N}, x \in \mathbb{Z}^d}$ a discrete $(1 + d)$-dimensional field of i.i.d. random variables, with law denoted by $\mathbb{P}$. We assume that

\[ P[\eta \geq -1] = 1 \quad \text{and} \quad E[\eta] = 0. \tag{1.1} \]

With some harmless abuse of notation, we let $\eta$ denote a generic random variable with the same law as $\eta_{n,x}$. We consider the following $(1 + d)$-dimensional (discrete) directed polymer model, in environment $(\eta_{n,x})_{n \in \mathbb{N}, x \in \mathbb{Z}^d}$. Let $S = (S_i)_{i \geq 0}$ be the simple symmetric random walk on $\mathbb{Z}^d$, with law denoted by $\mathbb{P}$. Given a parameter $\beta \in (0, 1)$ (which allows to tune the disorder’s intensity) we define the partition function $Z_{N,\beta}^\eta$ by

\[ Z_{N,\beta}^\eta := E \left[ \prod_{n=1}^N (1 + \beta \eta_{n,S_n}) \right], \tag{1.2} \]

and the associated polymer (Gibbs) measure $\mathbf{P}_{N,\beta}^\eta$ by

\[ \frac{d\mathbf{P}_{N,\beta}^\eta}{d\mathbf{P}}(S) := \frac{1}{Z_{N,\beta}^\eta} \prod_{n=1}^N (1 + \beta \eta_{n,S_n}). \tag{1.3} \]

The environment $\eta$ can be thought as a field of impurities, and under $\mathbf{P}_{N,\beta}^\eta$ the law of the random walk is modified so that it favors visits to (space-time) sites where $\eta$ assumes a larger value. Assumptions (1.1) are merely practical: they ensure that $1 + \beta \eta_{n,S_n}$ is always positive and imply that $E[Z_{N,\beta}^\eta] = 1$. 
The directed polymer model has a long history, dating back to [36], see [26] for an extensive review. In many directed polymer references (including [26]) the setup is slightly different and the Gibbs weights are rather written in an exponential form $\exp(\beta \sum_{n=1}^{N} \tilde{\eta}_{n,S_n})$ instead of $\prod_{n=1}^{N} (1 + \beta \eta_{n,S_n})$ used here. For most purposes the two formalisms are equivalent, but the latter turns out to be the adequate one for the specific problem we wish to study (we discuss this point later in the introduction, see Remark 1.1).

Localization transition. — A major point of focus in the directed polymer model has been the localization transition from a high temperature diffusive phase ($\beta$ small) to a low temperature localized phase ($\beta$ large). This phase transition can be studied via the free-energy $p(\beta) := -\lim_{N \to \infty} \frac{1}{N} \mathbb{E}[\log Z_{N,\beta}^{\eta}]$; we refer to [27, Prop. 2.5] for a proof of its existence. The free-energy is a non-negative, non-decreasing and continuous function of $\beta \in (0, 1)$ (see [29, Th. 3.2] for a proof, [48, Th. A.1] for its adaptation to the setup presented here). In particular there exists a critical value $\beta_c \in [0, 1]$ which is such that $p(\beta) = 0$ if and only if $\beta \leq \beta_c$.

This phase transition has been mostly studied in the case where the environment has a finite second moment $\mathbb{E}[\eta^2] < \infty$. In the exponential setup, this corresponds to having $\mathbb{E}[e^{2\beta \tilde{\eta}}] < \infty$ (the standard assumption considered in the literature is that $\eta$ has exponential moment of all orders see e.g. [1]) and under this assumption it has been show that $\beta_c > 0$ when $d \geq 3$, in [13, 37], while $\beta_c = 0$ when $d = 1$ [28] and $d = 2$ [39]. In particular this implies that there is no observable transition in dimension $d = 1$ and $d = 2$.

Intermediate disorder regime and scaling limit. — When $d \leq 2$, under a finite second moment assumption (assuming that $\beta > 0$ and $\text{Var}(\eta) > 0$), we have

$$\lim_{N \to \infty} Z_{N,\beta}^{\eta} = 0 \quad \text{and} \quad \lim_{\beta \to 0} Z_{N,\beta}^{\eta} = 1.$$ 

A legitimate question is therefore to know how to scale $\beta$ with $N$ (or $N$ with $\beta$) in order to observe a non-trivial random behavior for $Z_{N,\beta_N}^{\eta}$ and $P_{N,\beta}$ in the limit $N \to \infty$.

This problem has been the object of a large number of works [1, 17, 19] (see the review [18] and references therein). When $d = 1$, the correct scaling is to take $\beta$ proportional to $N^{-1/4}$ — note that in this case, $N$ is proportional to the correlation length of the system which is given by $|p(\beta)|^{-1} \approx \beta^{-4}$ see [3, 44]. The limit is formally obtained by replacing the random walk path and its environment by their scaling limit, which are respectively given by Brownian motion and space-time White Noise. In particular, the scaling limit of the partition function $\lim_{N \to \infty} Z_{N,\beta N^{-1/4}}^{\eta}$ is intimately related to the solution of the stochastic heat equation (SHE) with multiplicative noise [11].

The case of the dimension $d = 2$ presents additional difficulty as the SHE with multiplicative noise is ill-defined, so that the heuristic picture we had in dimension $d = 1$ cannot be valid. For a hierarchical version of the model the scaling limit of
the polymer measure when $\beta_N$ is sent to zero at the appropriate rate is identified in [24]. The original problem for the model on $\mathbb{Z}^2$ (and its continuum counterpart on $\mathbb{R}^2$) is still partially open but it has witnessed substantial progress in the recent years [20, 21, 35]. A decisive step toward the identification of the scaling limit, i.e., the convergence of the partition functions for the right value of $\beta_N$, has been made in a recent breakthrough paper [22]. More precisely it is shown that taking $\beta_N = \frac{\sqrt{\pi}}{\sqrt{\log N}}(1 + \vartheta/\log N)$ for some $\vartheta \in \mathbb{R}$, the rescaled field of point-to-point partition functions $N Z_{N,\beta_N}^\eta(\sqrt{|N x|}, \sqrt{|N y|})_{x \in \mathbb{R}^d, y \in \mathbb{R}^d}$ converges in distribution when $N \to \infty$ to a non-trivial limit, called the critical $2D$ stochastic heat flow. Note that in this case also, the choice for $\beta_N$ is such that the corresponding correlation length $|p(\beta_N)|^{-1}$ is proportional to $N^{1+o(1)}$, since $\log |p(\beta)| \sim -\pi/\beta^2$ as $\beta \downarrow 0$ (see [7]).

Heavy tailed disorder. — Our main motivation is to investigate intermediate disorder limits beyond the case $\mathbb{E}[\eta^2] < \infty$. Our interest lies in the case where $\eta$ is in the domain of attraction of an $\alpha$-stable law for $\alpha \in (1,2)$ and that (1.1) still holds (we can also consider the case $\alpha \in (0,1]$ if one drops the assumption that $\eta$ has zero average). To be more specific, let us assume the tail distribution has a pure power-law decay, i.e., that in the large $z$ limit we have

\begin{equation}
\mathbb{P}(\eta > z) = z^{-\alpha}(1 + o(1)).
\end{equation}

This kind of heavy tail environment has been studied in [48]. In this case, the existence of a non-trivial weak disorder phase depends on $\alpha$ and the dimension $d$. We have $\beta_c = 0$ if and only if $d \geq 2/(\alpha - 1)$, see [48, Th.1.1]. Moreover, when $d < 2/(\alpha - 1)$, the behavior of the free energy near criticality (that is, for $\beta$ small) is given by $p(\beta) = \beta^{(1+\alpha)(1)}$ with $\nu = 2\alpha/(2 - d(\alpha - 1))$.

One of our main goals is to identify the intermediate disorder scaling limit of this model under the assumption (1.4), when $\alpha < 1 + 2/d$, i.e., $\alpha \in (0,1]$ or $\alpha \in (1,2)$ and $d < 2/(\alpha - 1)$. We present in this paper the construction of the continuum measure that appears as the limit of $P_{N,\beta_N}^\eta$ in the intermediate disorder regime. The convergence of the discrete model to the continuum one, when $\beta_N$ goes to 0 at some adequate rate, is the object of a separate work [9], see Theorem A below.

**Remark 1.1.** — Let us stress that directed polymers in heavy-tail random environment are also considered in [4, 10, 30]: the main difference is that in these papers the Gibbs weights are written in the exponential form $\exp(\beta \sum_{n=1}^{N} \tilde{\eta}_{n,S_n})$. When the second moment of $\tilde{\eta}_{n,S_n}$ is infinite, such a model exhibits very strong localization properties: polymer trajectories remain in the neighborhood of a single favorite trajectory which visits the high energy sites (see [4, 10]). Also, the intermediate disorder regime is somehow trivial in this case. Indeed, in [10] the authors show that there is a specific scaling at which a sharp weak-to-strong disorder transition occurs. Under this scaling, there is a (random) threshold below which the partition function goes to 1 and above which it goes to $+\infty$ (see [10, Th.2.7-2.8] for a more precise statement). For a fixed value of $\beta$, the two setups, exponential ($e^{\beta \tilde{\eta}}$) and additive ($1 + \beta \eta$)
are equivalent, since one can be obtained from the other via a simple transformation. Of course such a correspondence between the two setups disappears when studying the limit $\beta \to 0$, with a fixed distribution for $\eta$ or $\tilde{\eta}$. In the heavy-tail environment case, the additive setup $(1 + \beta \eta)$ is the one which keeps a balance between the randomness of the random walk and the rewards of the environment and allows the existence of a scaling limit in which both are retained. This comes from the fact the field $\beta \eta_{n,x}$ converges after scaling (as a distribution) to a non trivial limit — this is never the case for $\exp(\beta \tilde{\eta}_{n,x})$, even after centering, because large values of $\tilde{\eta}$ create too wild fluctuations.

1.2. An informal definition of a continuum polymer with Lévy noise

Before stating our main result concerning the intermediate disorder regime in $\alpha$-stable environment, we need to provide a description of the scaling limit. The object we construct is formally obtained by considering a Feynman–Kac formula where the random walk and the environment are replaced by their respective scaling limits.

The scaling limit of our random walk is a Brownian motion with covariance matrix $\frac{1}{2}I_d$ where $I_d$ is the identity matrix in $\mathbb{R}^d$. To define the continuum polymer, we rather consider a standard $d$-dimensional Brownian motion $(B_t)_{t \in [0,T]}$ (for practical reason it is convenient to define $B$ only until a fixed finite time horizon $T$). We let $Q$ denote the associated distribution (we omit the dependence in $T$ to lighten notation) on the Wiener space

$$C_0([0,T]) := \{ \varphi : [0,T] \to \mathbb{R}^d : \varphi \text{ is continuous and } \varphi(0) = 0 \},$$

endowed with the topology of uniform convergence and the associated Borel $\sigma$-algebra.

When $\eta$ has a finite second moment, the scaling limit for the environment is given by a space-time Gaussian white noise. In that case a Brownian polymer model in dimension $d = 1$ can be (and has been) constructed based on the solution of the stochastic heat equation, see [2]. On the other hand, in the case where (1.4) is satisfied for some $\alpha \in (0,2)$, we have to consider a different object, namely the space-time $(1 + d)$-dimensional $\alpha$-stable noise with Lévy measure supported on $\mathbb{R}_+$. This is the multidimensional analog of the derivative of the $\alpha$-stable process with only positive jumps. This is a well studied object, see [31] and references therein, but we try to offer here a short and self-contained introduction for the sake of completeness. For simplicity, we focus our exposition on the case $\alpha \in (1,2)$, which displays the most interesting phenomenology. However we also treat below a much more general class of noise which includes the case $\alpha \in (0,1]$.

One-sided $\alpha$-stable noise in $\mathbb{R} \times \mathbb{R}^d$. — Given $\alpha \in (1,2)$, we start with a Poisson point process $\omega$ on $\mathbb{R} \times \mathbb{R}^d \times \mathbb{R}_+$ (time, space, and value of disorder) with intensity

$$dt \otimes dx \otimes \alpha v^{-(1+\alpha)}dv,$$

which is obtained as the scaling limit of the extremal process associated with the field $(\eta_{n,x})_{n \in \mathbb{N}, x \in \mathbb{Z}^d}$, under the assumption (1.4). As it shall draw no confusion the...
distribution of $\omega$ is also denoted by $\mathbb{P}$. Our $\alpha$-stable Lévy noise $\xi_\omega$ is the random distribution which is formally obtained by summing weighted Dirac masses $v \delta_{(t,x)}$ corresponding to all the points $(t,x,v) \in \omega$ and subtracting a non-random quantity so that the obtained distribution is centered in expectation. The delicate part is that, as in the definition of $\alpha$-stable processes, the counter term that has to be subtracted is infinite.

Let us thus explain how $\xi_\omega$ can be obtained using a limiting procedure. We consider $\omega$ as a set of points, and for any $a \in (0,1]$ we define $\omega(a) := \{(t, x, v) : v \geq a\}$ the truncated environment, i.e., removing atoms (jumps) of size less than $a$. We then let $\xi(a) \omega$ be the random measure on $\mathbb{R} \times \mathbb{R}^d$ defined by

$$\xi(a) \omega := \left( \sum_{(t, x, v) \in \omega} v \mathbf{1}_{\{v \geq a\}} \delta_{(t,x)} \right) - \frac{\alpha(a^{1-\alpha} - 1)}{\alpha - 1} L,$$

where $L$ denotes Lebesgue measure on $\mathbb{R} \times \mathbb{R}^d$ (note that our centering only compensates the jumps of intensity smaller than one, so that $\xi(a) \omega$ is not centered). We define $\xi_\omega$ as the distributional limit of $\xi(a) \omega$ when $a$ tends to zero.

For the sake of fixing ideas, let us specify a functional space in which this convergence holds. Given $s \in \mathbb{R}$, the Sobolev space $H^s(\mathbb{R}^{1+d})$ is defined as the closure of the space of smooth compactly supported function with respect to the norm

$$\|f\|_{H^s} := \left( \int_{\mathbb{R}^{1+d}} (1 + |z|^2)^s |\hat{f}(z)|^2 \, dz \right)^{1/2},$$

where $\hat{f}(z) = \int_{\mathbb{R}^{1+d}} f(x) e^{-iz \cdot x} \, dx$ is the Fourier transform of $f$. We also consider the local Sobolev space

$$H^s_{\text{loc}}(\mathbb{R}^{1+d}) := \{ f : f \psi \in H^s \text{ for every } C^\infty \text{ compactly supported } \psi \},$$

considered with the topology induced by the family of semi-norms $\|f \psi\|_{H^s}$ indexed by $\psi$. We then have the following (standard) result: When $\alpha \in (1,2)$, then $\xi(a)$ converges almost surely in $H^s_{\text{loc}}(\mathbb{R}^{1+d})$ with $s > (1 + d)/2$ towards a limit $\xi_\omega \in H^s_{\text{loc}}(\mathbb{R}^{1+d})$, see Proposition A.1. In particular, this means that $\xi_\omega$ can be integrated against any function in $H^s(\mathbb{R}^{1+d})$ which has compact support.

**Informal description of the scaling limit.** — In order to describe the candidate scaling limit of the model (1.3) we must make sense of a Feynman–Kac formula analogous to (1.2) in which the random walk $S$ is replaced by a Brownian motion $B$ and $\eta$ replaced by the $\alpha$-stable noise $\xi_\omega$. Similarly to (1.3) we wish to define a polymer model which is a modification of the Wiener Brownian measure $Q$ obtained via tilting by an energy functional. For $T > 0$ and $\beta > 0$ we would like to define $Q_{T,\beta}$ as

$$\frac{dQ_{T,\beta}}{dQ}(B) = \frac{1}{Z_{T,\beta}} e^{\beta H_L(B)},$$
where the energy functional is given by \( \xi \), integrated against the Brownian trajectory, in the following sense (\( \delta(s,y) \) denotes the Dirac mass at \((s,y)\))

\[
H_\omega(B) = \xi \left( \int_0^T \delta(t,B_t) \, dt \right).
\]

At this stage, we only consider this expression at a formal level, as it is quite clear that the fact that \( \xi \in H^{-s}_{\text{loc}} \) is not sufficient to provide a mathematical interpretation of this expression.

The exponential \( e^{\beta H_\omega(B)} \): is to be interpreted as an analogous of the time-ordered Wick exponential which is considered for the continuum directed polymer in white noise, see [2]. Informally, \( e^{\beta H_\omega(B)} \): is defined via the following expansion

\[
e^{\beta H_\omega(B)} := \sum_{k=0}^{\infty} \frac{\beta^k}{k!} \int_0^T \cdots \int_0^T \prod_{i=1}^{k} \xi \left( \delta(t_i,B_{t_i}) \right) dt_1 \cdots dt_k.
\]

While it is challenging to make sense of the above formula, things become simpler if one looks at the partition function \( Z_\omega^{\gamma}\beta \): because a formal integration with respect to Brownian trajectories makes the integrand more regular. Let us denote

\[
\rho_t(x) := \frac{1}{(2\pi t)^{d/2}} e^{-\|x\|^2/2t},
\]

the heat-kernel on \( \mathbb{R}^d \) associated with the Brownian motion (\( \|x\| \) stands for the Euclidean norm of \( x \)). For \( 0 < t_1 < \cdots < t_k \) and \( x_1, \ldots, x_k \in \mathbb{R}^d \), we also use the short-hand notation

\[
\varrho(t, x) := \prod_{i=1}^{k} \rho_{t_i-t_{i-1}}(x_i-x_{i-1}),
\]

with by convention \( t_0 = 0 \) and \( x_0 = 0 \) (in the following, if a different choice is made it will be duly notified). We will also use the notation \( dt \) and \( dx \) for Lebesgue measure on \( \mathbb{R}^k \) and \( (\mathbb{R}^d)^k \) respectively. With these notation, the expectation of (1.9) with respect to the Wiener measure can be formally defined by

\[
\mathbb{E}_\omega^{\gamma}\beta = 1 + \sum_{k=1}^{\infty} \frac{\beta^k}{k!} \int_0^T \cdots \int_0^T \int_{(\mathbb{R}^d)^k} \varrho(t, x) \prod_{i=1}^{k} \xi \left( dt_i, dx_i \right).
\]

In order to give a meaning to the above expression, we will approximate \( \xi \) by \( \xi^{(a)} \) and investigate the limiting behavior when \( a \) goes to zero. As it will be seen later, giving a meaning to \( \mathbb{E}_\omega^{\gamma}\beta \) is the most important step in order to give a rigorous interpretation to (1.8).

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2. Model and results

We can now introduce our results. We present in Sections 2.1–2.2 our construction of the continuum measure $Q_{T,\beta}^\omega$, thus defining the continuum directed polymer in Lévy noise. For pedagogical reasons, we first present in Section 2.1 the case of the $\alpha$-stable noise with $\alpha \in (1, 2)$, since it corresponds to the scaling limit of the model introduced in Section 1.1 above; we turn afterward in Section 2.2 to the case of a general heavy-tail noise. In Section 2.3, we present finer properties of the measure constructed and in Section 2.4 we discuss the relation between our model and the stochastic heat equation with multiplicative Lévy noise. Further comments on the results are made in Section 2.5.

2.1. The construction of the continuum polymer in Lévy $\alpha$-stable noise

Our main result is the construction of a measure on the Wiener space $C_0([0, T])$, corresponding to the definition (1.8). To ease the exposition, we single out the most important step of this construction which is the construction of the partition function, that is giving a mathematical interpretation for the formal integral (1.12). As mentioned above, we treat the case of an $\alpha$-stable noise first, before we turn to more general noises.

Recall the definition (1.6) of the truncated noise $\xi^{(a)}$. We define, for $a > 0$,

\begin{equation}
Z_{\omega,a}^{T,\beta} := 1 + \sum_{k=1}^{\infty} \beta^k \int_{0<t_1<\cdots<t_k<T} \int_{(\mathbb{R}^d)^k} g(t, x) \prod_{i=1}^k \xi^{(a)}(dt_i, dx_i).
\end{equation}

Since $\xi^{(a)}(dt_i, dx_i)$ is a locally finite signed measure, the only possible issue with the above definition is the integrability over $t_i$’s and $x_i$’s and summability over $k$. These conditions are not difficult to check and this done in Proposition 3.1. It is also not immediate from (2.1) that $Z_{\omega,a}^{T,\beta}$ is positive (which is a required property for being a partition function), but this is ensured by Lemma 3.3.

We prove that considering the limit of $Z_{\omega,a}^{T,\beta}$ when $a \downarrow 0$, we obtain a non-trivial (i.e., disordered) quantity, provided that $\alpha$ is smaller than a critical threshold. Let us define

\begin{equation}
\alpha_c = \alpha_c(d) = \begin{cases} 2 & \text{if } d = 1, 2, \\ 1 + 2/d & \text{if } d \geq 3. \end{cases}
\end{equation}

**Theorem 2.1.** — If $\alpha \in (1, \alpha_c)$ with $\alpha_c$ defined in (2.2), there exists an almost surely positive random variable $Z_{\omega}^{T,\beta}$ such that the following convergence

\[ \lim_{a \to 0} Z_{\omega,a}^{T,\beta} = Z_{\omega}^{T,\beta} \]

holds almost surely and in $\ell_1$. When $d \geq 3$ and $\alpha \in [\alpha_c, 2)$ then for all $\beta > 0$ we have

\[ \lim_{a \to 0} Z_{\omega,a}^{T,\beta} = 0 \]

almost surely.

**Remark 2.2.** — Note that the definitions (1.6) and (2.1) also make sense when $\alpha \geq 2$. In that case $\xi^{(a)}$ does not converge to a limiting distribution but this does a priori
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prevent $\mathcal{L}_{T,\beta}^\alpha$ from having a non-trivial limit. Proposition 2.12 below shows that we have in fact $\lim_{\alpha \downarrow 0} \mathcal{L}_{T,\beta}^\alpha = 0$ for every $\alpha \in (\alpha_c, \infty)$ in any dimension $d \geq 1$.

Let us now present the construction of the polymer measure described in (1.8). Recall that our objective is to define a probability on the Wiener space $C_0([0,T])$ which corresponds to the formal definition (1.8). We proceed in a similar manner as with the partition function: we first consider a measure on $C_0([0,T])$ built with the truncated noise $\xi^\alpha$. Let us follow the following families of functions on $C_0([0,T])$:

- $\mathcal{B} := \{ f : C_0([0,T]) \to \mathbb{R} : f$ measurable and bounded, $\}$,
- $\mathcal{C} := \{ f : C_0([0,T]) \to \mathbb{R} : f$ continuous and bounded, $\}$,
- $\mathcal{B}_b := \{ f \in \mathcal{B} : \text{Support}(f) \text{ is bounded}, \}$
- $\mathcal{C}_b := \{ f \in \mathcal{C} : \text{Support}(f) \text{ is bounded}, \}$

Recall that $C_0([0,T])$ is equipped with the topology of the supremum norm: hence we say that $f : C_0([0,T]) \to \mathbb{R}$ has bounded support if there exists $M > 0$ such that $f(\varphi) = 0$ for any $\varphi \in C_0([0,T])$ with $||\varphi||_\infty > M$.

Given a bounded Borel-measurable function $f \in \mathcal{B}$, we define

\begin{equation}
\mathcal{L}_{T,\beta}^\alpha(f) = \mathcal{Q}(f) + \sum_{k=1}^\infty \beta^k \int_{0<t_1<\cdots<t_k<T} \int_{(\mathbb{R}^d)^k} \varrho(t, x, f) \prod_{i=1}^k \xi^\alpha(\{d \tau_i, dx_i\})
\end{equation}

where we use the notation $\mathcal{Q}(f) := \mathcal{Q}(f((B_t)_{t \in [0,T]}))$, and where $\varrho(t, x, f)$ is defined by (recall (1.11))

$\varrho(t, x, f) = \varrho(t, x)\mathcal{Q}\left[ f \left( (B_t)_{t \in [0,T]} \right) \mid \forall i \in [1, k], B_{t_i} = x_i \right].$

With some abuse of notation, the conditional measure $\mathcal{Q}(\cdot \mid \forall i \in [1, k], B_{t_i} = x_i)$ denotes the distribution of the process obtained by concatenating independent Brownian bridges linking $(t_{i-1}, x_{i-1})$ to $(t_i, x_i)$ for $i \in [1, k]$. The fact that (2.3) is well-defined for $f \in \mathcal{B}_b$ is ensured by Proposition 3.1 below: the extension to non-negative $f \in \mathcal{B}$ is given in Lemma 3.3. Note that $f \mapsto \varrho(t, x, f)$ is linear and thus so is $\mathcal{L}_{T,\beta}^\alpha(\cdot)$.

From Lemma 3.3 below, $\mathcal{L}_{T,\beta}^\alpha(f) \geq 0$ when $f \geq 0$ and $\mathcal{L}_{T,\beta}^\alpha(1) = \mathcal{L}_{T,\beta}^\alpha > 0$. As a consequence, for any $a > 0$, we can define a probability measure $\mathcal{Q}_{T,\beta}^{\alpha,a}$ on $C_0([0,T])$ by setting

\begin{equation}
\mathcal{Q}_{T,\beta}^{\alpha,a}(A) := \frac{\mathcal{L}_{T,\beta}^\alpha(1_A)}{\mathcal{L}_{T,\beta}^\alpha},
\end{equation}

for any Borel set $A$. We also write $\mathcal{Q}_{T,\beta}^{\alpha,a}(f)$ for the expectation, with respect to $\mathcal{Q}_{T,\beta}^{\alpha,a}$, of a function $f : C_0([0,T]) \to \mathbb{R}$. In the same way as for the partition function, we define the measure $\mathcal{Q}_{T,\beta}$ as the limit of $\mathcal{Q}_{T,\beta}^{\alpha,a}$ when $a$ goes to zero: this requires $\alpha < \alpha_c$, and the convergence holds for the weak topology. Let $\mathcal{M}_T$ denote the space of probability measures on $C_0([0,T])$ equipped with the topology of weak convergence.
Theorem 2.3. — If $\alpha \in (1, \alpha_c)$, there exists a probability measure $Q_{T,\beta}$ on $C_0([0, T])$ such that the following convergence holds almost surely in $\mathcal{M}_T$

$$\lim_{a \to 0} Q_{T,\beta}^a = Q_{T,\beta}.$$

In other words, we have almost surely for every $f \in \mathcal{C}$

$$\lim_{a \to 0} Q_{T,\beta}^a(f) = Q_{T,\beta}(f).$$

Since $Z_{T,\beta}^a(\cdot)$ induces a positive measure on $C_0([0, T])$, the above statement turns out to be equivalent to the existence of a positive measure $Z_{T,\beta}^\omega$ such that for every $f \in \mathcal{C}$

$$\lim_{a \to 0} Z_{T,\beta}^a(f) = Z_{T,\beta}(f).$$

2.1.1. Scaling limit of the discrete model. — In order to justify the fact that $Q_{T,\beta}$ is the natural model for a continuum polymer based on $\alpha$-stable noise, let mention here the scaling limit result which we prove in [9], namely that the discrete polymer model defined in (1.3), when properly rescaled, converges to the continuum polymer in Lévy stable noise. We present the convergence with time horizon $T = 1$ (which yields no loss of generality by scaling) and set $Q_{T,\beta} := Q_{T,\beta}^\omega$.

Let $S_t^{(N)}$ be the linear interpolation of a random walk trajectory, rescaled diffusively:

$$S_t^{(N)} := \sqrt{\frac{d}{N}} \left((1 - ut)S_{\lfloor Nt \rfloor} + utS_{\lfloor Nt \rfloor + 1}\right), \quad \text{with} \quad ut = Nt - \lfloor Nt \rfloor.$$

We then have the following convergence result.

Theorem A (cf. [9]). — Assume that the distribution of the environment $\eta$ satisfies (1.4) for some $\alpha \in (1, \alpha_c)$, with $\alpha_c$ defined in (2.2). Setting

$$\beta_N := \frac{2(1-\alpha)/\alpha \beta(1-\alpha)/2 - d(2\alpha)(1+2/d-\alpha)}{N^{-d/2}},$$

then we have the following convergence in distribution in $\mathcal{M}_1$,

$$P_{N,\beta_N} \left( (S_t^{(N)})_{t \in [0,1]} \in \cdot \right) \xrightarrow{N \to \infty} Q_{\beta}.$$  

Remark 2.4. — The prefactor in $\beta_N$ comes from various factors, including the normalization of the Brownian motion and the periodicity of the random walk. The above theorem remains valid slightly beyond the assumption (1.4), one can allow for a slowly varying function in the tail distribution provided an appropriate correction in the scaling of $\beta_N$ is made. The analogous result is valid also for $\alpha \in (0, 1]$. We refer to [9] for details.

2.2. The case of a general noise. — We have focused until now on the case of an $\alpha$-stable noise with $\alpha \in (1, 2)$, both because our motivation is to describe the scaling limit for the discrete polymer model with heavy tailed environment and to make the exposition lighter. Our result can nonetheless be applied to a much larger variety of noise. Let us consider in this section a Poisson process $\omega$ on $\mathbb{R} \times \mathbb{R}^d \times \mathbb{R}_+$, with density

$$dt \otimes dx \otimes \lambda(dv),$$

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where \( \lambda \) is a measure on \((0, \infty)\) such that \( \lambda([a, \infty)) < \infty \) for every \( a > 0 \). One may keep in mind the case \( \lambda(dv) = \alpha v^{-(1+\alpha)} dv \) with \( \alpha \in (0, 2) \), referred to as \( \alpha \)-stable.

We define the truncated noise \( \xi^{(a)}_\omega \) for \( a > 0 \) similarly to (1.6) (recall that \( \mathcal{L} \) denotes the Lebesgue measure on \( \mathbb{R} \times \mathbb{R}^d \))

\[
\xi^{(a)}_\omega = \left( \sum_{(t,x,\upsilon) \in \omega} \upsilon 1_{\{\upsilon \geq a\}} \delta_{(t,x)} \right) - \kappa_a \mathcal{L},
\]

where

\[
\kappa_a = \int_{[a, 1)} \upsilon \lambda(d\upsilon),
\]

note that we have in particular \( \kappa_a = 0 \) for \( a \geq 1 \). The truncated noise \( \xi^{(a)}_\omega \) converges to a limit \( \xi_\omega \in H^{-s}_{loc}(\mathbb{R}^{1+d}) \) with \( s > (1 + d)/2 \), if and only if 

\[
\int_{[0,1)} \upsilon^2 \lambda(d\upsilon) < \infty
\]

(see Proposition A.1). Note also that if 

\[
\int_{[0,1)} \upsilon \lambda(d\upsilon) < +\infty,
\]

then \( \kappa_0 < \infty \) and the definition (2.5) directly makes sense with \( a = 0 \) so this approximation procedure is not required. We define, similarly to (2.3), for any \( f \in \mathcal{B} \)

\[
Z^{w,a}_{T,\beta}(f) := Q(f) + \sum_{k=1}^{\infty} \beta^k \int_{0<t_1<\cdots<t_k<T} \int_{\mathbb{R}^d} g(t,x,f) \prod_{i=1}^{k} \xi^{(a)}_\omega(dt_i, dx_i).
\]

The condition that \( f \) has a bounded support ensures that all the integrals are well-defined since the integration is only over a bounded space-time region; the summability is shown in Proposition 3.1 below. Let us stress that Lemma 3.3 below ensures that 

\[
Z^{w,a}_{T,\beta}(f) \geq 0
\]

when \( f \) is non-negative. Given an increasing sequence of positive functions \( f_n \in \mathcal{B} \) converging to 1, one sets

\[
Z^{w,a}_{T,\beta} := \lim_{n \to \infty} Z^{w,a}_{T,\beta}(f_n).
\]

Lemma 3.3 also ensures that the above definition does not depend on the choice of \( f_n \).

Note that the above definition makes it possible to have \( Z^{w,a}_{T,\beta} = \infty \), but this does not occur provided the following condition is satisfied:

\[
\int_{[1, \infty)} (\log \upsilon)^{d/2} \lambda(d\upsilon) < \infty.
\]

Proposition 2.5. — Under the assumption (2.9), we have \( Z^{w,a}_{T,\beta} \in (0, \infty) \) for any \( a \in (0, 1) \).

The condition (2.9) is in fact optimal, if it fails to hold then our partition function is degenerate.

Proposition 2.6. — If the measure \( \lambda \) does not satisfy (2.9) then for any \( a \in (0, 1] \) we have \( Z^{w,a}_{T,\beta} = \infty \) almost surely.
Remark 2.7. — The condition (2.9) arises from an entropy-energy comparison. We let \( X_{R,T} \) be the largest atom in the environment located within a distance \( R \) in the time interval \( [0, T] \), that is

\[
X_{R,T} := \max \{ v : (t, x, v) \in \omega, t \in [0, T], \frac{R}{2} \leq \|x\| \leq R \}.
\]

In the large \( R \) limit, the probability of visiting the corresponding atom under the Wiener measure is of order \( \exp(-cT R^2) \). For any fixed \( \delta, T > 0 \), the criterion (2.9) is equivalent (via a Borel–Cantelli argument) to having

\[
\sup_{R>0} \left( X_{R,T} e^{-\delta R^2} \right) < \infty.
\]

From this, it is not difficult to deduce that the contribution of trajectories that visit one large atom located far away remains finite. This provides a heuristic justification of the criterion (2.9).

Using the partition function (2.7), we can define a probability measure \( Q_{T,\beta} \) on \( C_0([0, T]) \) in the same way as in (2.4), i.e., setting for any Borel set \( A \subset C_0([0, 1]) \)

\[
Q_{T,\beta}(A) := \frac{\mathcal{Z}_{T,\beta}(1_A)}{\mathcal{Z}_{T,\beta}}.
\]

Note that \( A \mapsto \mathcal{Z}_{T,\beta}(1_A) \) defines a locally finite measure on \( C_0([0, T]) \) even when (2.9) is not satisfied, cf. Proposition 3.1. Our main result in this section is that the limit when \( a \) goes to 0 is non-degenerate if \( \lambda \) satisfies the following assumption:

\[
\int_{(0,1)} v^2 \lambda(dv) < \infty, \quad \text{if } d = 1,
\]

\[
\int_{(0,1)} v^p \lambda(dv) < \infty \quad \text{for some } p < 1 + 2/d, \quad \text{if } d \geq 2.
\]

Theorem 2.8. — Under the assumption (2.10), for any fixed \( f \in \mathcal{B} \), the limit

\[
\lim_{a \to 0} \mathcal{Z}_{T,\beta}^a(f) = \mathcal{Z}_{T,\beta}(f)
\]

exists almost surely and is finite. We have \( \mathcal{Z}_{T,\beta}(f) > 0 \) if \( f \) is non-negative and \( Q(f) > 0 \). Furthermore, if (2.9) also holds then (2.11) also holds for \( f \in \mathcal{B} \). In particular, in the case \( f \equiv 1 \) we have,

\[
\lim_{a \to 0} \mathcal{Z}_{T,\beta}^a = \mathcal{Z}_{T,\beta} \in (0, \infty).
\]

The convergence holds in \( L_1 \) if and only if \( \int_{[1, \infty)} v \lambda(dv) < \infty \).

Additionally, \( Q_{T,\beta}^a \) converges weakly when \( a \to 0 \): There exists a probability measure \( Q_{T,\beta}^a \) on \( C_0([0, T]) \) such that, almost surely, for every \( f \in \mathcal{C} \) we have

\[
\frac{\mathcal{Z}_{T,\beta}^a(f)}{\mathcal{Z}_{T,\beta}} = \lim_{a \to 0} Q_{T,\beta}^a(f) = Q_{T,\beta}(f).
\]

Remark 2.9. — Note that the conditions (2.9)–(2.10) are satisfied when \( \lambda(dv) = \alpha v^{-(1+\alpha)} dv \) for \( \alpha \in (0, \alpha_c) \). When \( \alpha \in [\alpha_c, \infty) \), Proposition 2.12 below establishes that the limit is degenerate.
Remark 2.10. — When (2.9) is not satisfied, it is not difficult to check from our proof that almost surely, the convergence (2.11) holds simultaneously for all $f \in C_b$, that is $\mathcal{Z}_{T,\beta}^\omega (\cdot \cap A)$ converges vaguely (as a measure), for any bounded set $A$.

Remark 2.11. — A statement similar to the weak convergence of $Q_{T,\beta}^\omega$ can also be made in the case when (2.9) does not hold. In that case there exists a locally finite measure $\mathcal{Z}_{T,\beta}^\omega$ on $C_0([0,T])$ (in the sense that for every $M$ it gives finite mass to the set $\{\varphi : \|\varphi\|_\infty < M\}$) which is such that almost surely, for every $f \in C_b$

$$\lim_{a \to 0} \mathcal{Z}_{T,\beta}^\omega (f) = \mathcal{Z}_{T,\beta}^\omega (f).$$

The condition (2.10), which prevents $\mathcal{Z}_{T,\beta}^\omega (f)$ from vanishing as $a$ tends to zero, is close to optimal. Let us introduce the following alternative and almost equivalent condition

$$\begin{cases}
\int_{(0,1)} v^2 \lambda (dv) < \infty, & \text{if } d = 1, \\
\int_{(0,1)} |v|^2 \log(v) |\lambda (dv) < \infty, & \text{if } d = 2, \\
\int_{(0,1)} v^{1+2/d} \lambda (dv) < \infty, & \text{if } d \geq 3.
\end{cases}$$

Then we prove that the limit is degenerate as soon as (2.14) is violated. In particular the following result ensures that one cannot define the continuum polymer model when $\int_{(0,1)} v^2 \lambda (dv) = \infty$, in which case the noise $\xi_{\omega}$ is itself not well-defined (see Remark A.2).

Proposition 2.12. — If the measure $\lambda$ does not satisfies (2.14) then for any $f \in B_b$ we almost surely have

$$\lim_{a \to 0} \mathcal{Z}_{T,\beta}^{\omega,a}(f) = 0.$$

If (2.9) also holds, we have $\lim_{a \to 0} \mathcal{Z}_{T,\beta}^{\omega,a}(f) = 0$.

Note that this result proves the last statement of Theorem 2.1.

Remark 2.13. — From a Borel–Cantelli consideration, we have (for any value of $d$)

$$\int_{(0,1)} v^{1+2/d} \lambda (dv) < \infty \iff \sup_{(t,x,v) \in \omega} v^\rho (t,x) < \infty \text{ a.s.}$$

The quantity $v^\rho (t,x)$ corresponds to the multiplicative weight gained by trajectories which visit the point $(t,x)$ multiplied by the entropic cost to visit it. On a heuristic level, if $v^\rho (t,x)$ is unbounded, it means that there are atoms with arbitrarily small amplitude $z$ which have a large impact on the value log $\mathcal{Z}_{T,\beta}^{\omega,a}$, preventing the convergence of $\mathcal{Z}_{T,\beta}^{\omega,a}$ to a non-zero value. In view of the conditions (2.10)–(2.14) in Theorem 2.8 and Proposition 2.12, this criterion is not far from being sharp when $d \geq 2$. When $d = 1$ the condition (2.15) is less restrictive than the one necessary for the convergence of $\mathcal{Z}_{T,\beta}^{\omega,a}$ (cf. Proposition A.1).
**Remark 2.14.** — The difference between the conditions (2.10) and (2.14) when $d \geq 2$ leaves a small family of Lévy measures for which the question whether $2^{\omega,d}_T$ converges to a positive limit or to zero remains open. We do not believe that either condition (2.10) or (2.14) are optimal. Although refinements of the proofs presented here could most likely yield slightly finer condition on both sides, finding the necessary and sufficient condition remains a challenging issue. Even if the condition is not optimal, the $|\log v|$ factor present in (2.14) is of importance in dimension $d = 2$, since it underlines that in any dimension $d \geq 2$, in contrast with the case $d = 1$, there are some (Poisson) noises for which the continuum polymer (and the noisy stochastic heat equation see Section 2.4 below) are not defined.

**Remark 2.15.** — After a first draft of this manuscript was published, in a collaborative effort with C. Chong [6], the conditions (2.10) and (2.14) have been improved. More precisely, it is shown in [6] that when $d \geq 2$ Theorem 2.8 holds under the condition

$$\int_{(0,1)} v^{1+2/d} |\log(v)| |\lambda(\mathrm{d}v)| < \infty,$$

This identifies the necessary and sufficient condition for convergence when $d = 2$. When $d \geq 3$ the condition (2.14) for Proposition 2.12 has been improved to

$$\int_{(0,e^{-\cdot})} v^{1+2/d} |\log(v)||\log(\log(v)))|^{-c} \lambda(\mathrm{d}v) < \infty,$$

with $c$ any constant greater than $5 + 4/d$, leaving only a very narrow gap between the necessary and the sufficient condition.

2.3. **Main properties of the continuum directed polymer in Lévy noise**

Let us assume throughout the rest of this section that Assumptions (2.9)–(2.10) are satisfied. We describe under these assumptions a few key properties of our polymer measure. First, we underline how $Q^{\omega,\beta}_{T,T}$ is in some aspects very similar to the Wiener measure and is in others very singular with respect to it. Then, we provide an explicit expression for the finite-dimensional marginal density of the measure, via point-to-point partition functions.

2.3.1. **Basic properties of the continuum polymer in Lévy noise.** — Let us define $\mathbb{P} \times Q^{\omega,\beta}_{T,T}$ the averaged polymer measure as follows

$$\mathbb{P} \times Q^{\omega,\beta}_{T,T}(A) := \mathbb{E}[Q^{\omega,\beta}_{T,T}(A)].$$

**Proposition 2.16.** — The averaged polymer measure $\mathbb{P} \times Q^{\omega,\beta}_{T,T}$ is absolutely continuous with respect to $Q$.

Proposition 2.16 yields an important information concerning $Q^{\omega,\beta}_{T,T}$ since it implies that $\mathbb{P}$-almost surely $Q^{\omega,\beta}_{T,T}$ inherits any given $Q$-almost sure property of the Brownian motion.

**Corollary 2.17.** — If $A$ is a Borel set of $C_0([0,T])$ such that $Q(A) = 1$, then $\mathbb{P}$-a.s. we have $Q^{\omega,\beta}_{T,T}(A) = 1$. As an example, for almost every $\omega$, a trajectory $(B_t)_{t \in [0,T]}$ has
\( Q_{T,\beta} \)-a.s. a modulus of continuity given by \( \sqrt{2/\log(1/h)} \): in other words,

\[
Q_{T,\beta} \left( \{ \varphi \in C_0([0,T]): \limsup_{h \downarrow 0} \sup_{0 \leq t < T-h} \frac{|\varphi(t+h) - \varphi(t)|}{\sqrt{2/\log(1/h)}} = 1 \} \right) = 1.
\]

This implies in particular that for any \( Q \) everywhere locally most strikingly when \( \int_{(0,1)} \nu \lambda(\mathrm{d}v) = \infty \). To illustrate this fact, given \( \varphi \in C_0([0,T]) \) let us consider \( \Delta(\varphi, \omega) := \{ t \in [0,T]: \exists \nu > 0, (t, \varphi(t), \nu) \in \omega \} \).

Let us set

\[
\mathcal{A}_\text{dense}(\omega) := \{ \varphi \in C_0([0,T]): \Delta(\varphi, \omega) \text{ is dense in } [0,T] \},
\]

\[
\mathcal{A}_\text{empty}(\omega) := \{ \varphi \in C_0([0,T]): \Delta(\varphi, \omega) = \emptyset \},
\]

\[
\mathcal{A}_\infty(\omega) := \{ \varphi \in C_0([0,T]): \# \Delta(\varphi, \omega) = \infty \}.
\]

**Proposition 2.18.** Under Assumptions (2.9)–(2.10) the following statements hold.

(i) We have almost surely \( Q(\mathcal{A}_\text{empty}) = 1 \).

(ii) If \( \int_{(0,1)} \nu \lambda(\mathrm{d}v) < \infty \) then \( Q_{T,\beta}(\mathcal{A}_\text{empty}) \in (0,1) \) and \( Q_{T,\beta}(\mathcal{A}_\infty) = 0 \) a.s.

(iii) If \( \int_{(0,1)} \nu \lambda(\mathrm{d}v) = \infty \) we have \( Q_{T,\beta}(\mathcal{A}_\text{dense}) = 1 \) a.s.

**Remark 2.19.** Note that when \( \int_{(0,1)} \nu \lambda(\mathrm{d}v) < \infty \), we have in fact

\[
Q_{T,\beta}^\nu(\cdot | \mathcal{A}_\text{empty}) = Q,
\]

so \( Q_{T,\beta}^\nu \) is in that case not singular with respect to the Wiener measure. The technique used for the proof of Proposition 2.18 can possibly be pushed a bit further to yield the following statement:

- When \( \int_{(0,1)} \nu \lambda(\mathrm{d}v) < \infty \) then the convergence of \( Q_{T,\beta}^\nu \) towards \( Q_{T,\beta}^\nu \) holds also for the total variation distance.

- When \( \int_{(0,1)} \nu \lambda(\mathrm{d}v) = \infty \) then \( \| Q_{T,\beta}^\nu - Q_{T,\beta} \|_{TV} = 1 \) for every \( a > 0 \).

### 2.3.2. Point-to-point partition functions and finite dimensional marginals.

The aim of this section is to give an explicit description of the finite-dimensional marginals of \( Q_{T,\beta}^\nu \). If we fix \( 0 < t_1 < \cdots < t_k \leq T \), then the distribution of \( (B_{t_1}, \ldots, B_{t_k}) \) under \( Q_{T,\beta}^\nu \) is absolutely continuous with respect to the Lebesgue measure and its density can be expressed using the so-called point-to-point partition functions. For any \( a > 0 \), define for all \( t > 0 \) and \( x \in \mathbb{R}^d \) the partition function from \((0,0)\) to \((t,x)\) as (recall the definition (1.11))

\[
\mathcal{Z}_{T,\beta}^\nu(t,x) := \rho_t(x) + \sum_{k=1}^\infty \beta^k \int_{0 \leq t_1 < \cdots < t_k \leq t} \int_{[\mathbb{R}^d]^k} \varphi(t,x) \rho_{t-t_k}(x-x_k) \prod_{i=1}^k \xi^{(a)}(\mathrm{d}t_i, \mathrm{d}x_i),
\]
if the integral is convergent (and set $Z_\omega(t, x) = \infty$ if not). The following proposition shows that the point-to-point partition function of our continuum model — defined as the limit of $Z_\omega(t, x)$ when $a$ tends to zero — is well-defined, positive and finite.

**Proposition 2.20.** — Suppose that (2.9) holds, then given $a \in (0, 1]$, $t > 0$ and $x \in \mathbb{R}^d$ we have almost surely

$$Z_\omega(t, x) \in (0, \infty).$$

If (2.10) also holds then given $t > 0$ and $x \in \mathbb{R}^d$, we have almost surely

$$Z_\omega(t, x) = \lim_{a \to 0} Z_\omega(t, x), \quad \text{with } Z_\omega(t, x) \in (0, \infty).$$

If $\int^\infty_1 v \lambda(dv) < \infty$ then the convergence holds in $L_1$.

For $(s, y) \in \mathbb{R} \times \mathbb{R}^d$, let us define the shifted environment

$$\theta_{(s, y)} \omega := \{(t - s, x - y, v) : (t, x, v) \in \omega\}.$$ 

Then, we can define for $(t_1, x_1), (t_2, x_2) \in \mathbb{R} \times \mathbb{R}^d$, $t_1 < t_2$ the partition function linking two arbitrary points:

$$Z_\omega([t_1, x_1], (t_2, x_2)) := Z_\omega^{(t_1, x_1)\omega a}(t_2 - t_1, x_2 - x_1).$$

Finally we set

$$Z_\omega([t_1, x_1], (t_2, x_2)) := \limsup_{a \to 0} Z_\omega^{(t_1, x_1)\omega a},$$

and we omit the first coordinate in the notation when it is equal to 0. We use a limsup instead of a limit in the definition only to make sure that $Z_\omega([t_1, x_1], (t_2, x_2))$ is defined simultaneously for all $(t_1, x_1)$ and $(t_2, x_2)$. Note that Proposition 2.20, together with translation invariance, shows that for any fixed $(t_1, x_1), (t_2, x_2)$ the limsup in (2.20) can almost surely be replaced by a limit (so the point-to-point partition function $Z_\omega([t_1, x_1], (t_2, x_2))$ is almost surely well-defined, positive and finite).

**Proposition 2.21.** — For any $0 < t_1 < \cdots < t_k = T$, the set

$$\{(x_1, \ldots, x_k) : \forall i \in [1, k], \ Z_\omega^{(t_{i-1}, x_{i-1})\omega a}(t_i, x_i) = \lim_{a \to 0} Z_\omega^{(t_{i-1}, x_{i-1})\omega a}(t_i, x_i)\}$$

has almost surely full Lebesgue measure. Furthermore, the convergence

$$\lim_{a \to 0} \prod_{i=1}^k Z_\omega^{(t_{i-1}, x_{i-1})\omega a}\delta(t_i, x_i) = \prod_{i=1}^k Z_\omega^{(t_{i-1}, x_{i-1})\omega a}\delta(t_i, x_i),$$

holds almost surely in $L_1((\mathbb{R}^d)^k)$. Additionally, for almost every $\omega$, the $k$-marginals measure $Q_t^{\omega}(B_{t_1}, \ldots, B_{t_k} \in \cdot)$ is absolutely continuous with respect to the Lebesgue measure and we have for any bounded measurable $g$ on $(\mathbb{R}^d)^k$

$$Q_t^{\omega}(g(B_{t_1}, \ldots, B_{t_k})) = \frac{1}{Z_\omega^{\omega a}(T)} \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^k Z_\omega^{\omega a}\delta(t_{i-1}, x_{i-1}), (t_i, x_i) dx.$$ 

**Remark 2.22.** — For $k = 1$ the above proposition states that the density of the distribution of $B_T$ under $Q_t^{\omega}$ is given by $Z_\omega(T, \cdot)/Z_\omega^{\omega a}(T).$
Remark 2.23. — Let us stress that in the above proposition, we fix $0 < t_1 < \cdots < t_k = T$ before considering a realization of $\omega$. This is an important point since there are exceptional times for which $Q_{T,\beta}^\omega(B_t \in \cdot)$ admits no density. In fact is not difficult to check that if $(t, x, \nu) \in \omega$ then $Q_{T,\beta}^\omega(B_t = x) > 0$.

2.4. Connection with the stochastic heat equation with multiplicative Lévy noise

In [2] the continuum directed polymer model with white noise is constructed directly from the solution of the stochastic heat equation (SHE) with multiplicative Gaussian white noise. It is not possible to proceed in this way with a general Lévy noise (simply because the solution is not regular enough) and our approach here is quite different. The continuum model constructed in Theorem 2.8 bears nonetheless a strong connection with the SHE with multiplicative Lévy noise. We discuss here this connection in some more detail and compare our results with the existing ones concerning the SHE with Lévy noise. Our formal definition $Z_{T,\beta}^\omega = Q_{T,\beta}^\omega[ : e^{\beta H_{\omega}^t(B) :}]$ (see (1.9)) corresponds to a Feynman–Kac formula associated with the following equation

\[ \partial_t u := \frac{1}{2d} \Delta u + \beta \xi_{\omega} \cdot u. \]

More precisely, the point-to-point partition function $Z_{T,\beta}^\omega(t, x)$ defined in (2.17) formally corresponds to the solution of (2.21) with $\delta_0$ initial condition. Starting from an arbitrary initial condition $u_0$ (a locally finite signed measure), the solution of (2.21) should take the form

\[ u(t, x) := \int_{\mathbb{R}^d} D_{T,\omega}^\beta([0, y), (t, x)] u_0(dy). \]

In the case $u_0(dy) = g_0(y)dy$ for some bounded and measurable function $g_0$, the fact that (2.22) is well-defined derives from Proposition 2.21 (in the case $k = 1$), combined with a time-reversal argument giving

\[ (\mathcal{D}_{T,\beta}^\omega([0, y), (t, x)])_{y \in \mathbb{R}^d} \overset{(d)}{=} (\mathcal{D}_{T,\beta}^\omega([0, x), (t, y)])_{y \in \mathbb{R}^d}, \]

that ensures that $\mathcal{D}_{T,\beta}^\omega([0, \cdot), (t, x)] \in L_1(\mathbb{R}^d)$ almost surely. For the general case where $u_0$ is a measure, we refer to Proposition 2.26 below for the well-posedness of (2.22).

The equation (2.21) has been extensively studied (often under a more general form, see e.g. [43, 45]). To our knowledge the most complete results concerning the existence of solutions have been given in [23]. More precisely, in [23] the existence of solutions in the integral form

\[ u(t, x) = \int_{\mathbb{R}^d} \rho_t(x - y) u_0(dy) + \beta \int_0^t \int_{\mathbb{R}^d} \rho_{t-s}(x - y) u(s, y) \xi_\omega(ds, dy), \]

called mild solutions, are obtained under the condition

\[ \int_{(0, 1]} v^p \lambda(dv) < \infty \quad \text{and} \quad \int_{[1, \infty)} v^q \lambda(dv) < \infty, \]
with $p \in (0,1+2/d)$ and $(2+2/d-p)^{-1} \leq q \leq p$ and for $u_0 = g_0(y)dy$ with $g_0$ bounded and measurable. Uniqueness has been established earlier [45] under the more stringent assumption $\int_{(0,\infty)} v^p \lambda(dv) < \infty$ for some $p \in [1,1+2/d)$, which for instance excludes $\alpha$-stable noises. Let us stress that the above is a very partial account of the results in [23] since the existence results deal with a more general class of equations and allows for a wider variety of noise (it allows for complex jumps and when $d = 1$ for a Gaussian white noise part as well as for space-time inhomogeneities).

While our assumptions (2.9)–(2.10) are less restrictive than (2.24), we cannot prove that (2.22) solves the equation (2.23) under these assumptions. However, we can show that the solution of the equation with a truncated noise converges almost surely when the truncation levels goes to zero and infinity respectively. Additionally, we keep quite a large freedom concerning the choice of the initial condition. Let us write this result in full detail for completeness. We are going to make the following assumption on the initial condition

$$\limsup_{r \to \infty} r^{-2} \log \left( \left| u_0 \right|([-r,r]^d) \right) < \frac{1}{2T}$$

where $|u_0|$ is the total variation of the measure $u_0$. This condition is present to ensure that (2.22) is well defined and almost surely finite on the interval $[0,T]$. For $b > a$, let us introduce $\xi^{(a,b)}_\omega$ the noise truncated at levels $a$ and $b$ (recall the notation (2.6) for $\kappa_a$)

$$\xi^{(a,b)}_\omega := \sum_{(t,x,v) \in \omega} v \mathbf{1}_{v \in (a,b)} \delta_{(t,x)} + (\kappa_b - \kappa_a) L.$$

Then, setting by convention the quantity to be $\infty$ when the integral is not well-defined, we set

$$w^{(a,b)}(t,x) := \int_{\mathbb{R}^d} \mathcal{Z}_{a}^{\omega,a}([0,y),(t,x)] u_0(dy),$$

where $\mathcal{Z}_{a}^{\omega,a}([0,y),(t,x)]$ is defined as in (2.17) with $\xi^{(a)}_\omega$ replaced by $\xi^{(a,b)}_\omega$. By [45, Th.1.2.1], if $u_0$ is absolutely continuous with bounded density w.r.t. to the Lebesgue measure, then $w^{(a,b)}$ is the unique solution (in some reasonable functional space) of (2.23) (with noise $\xi^{(a,b)}_\omega$). We first observe that $w^{(a,b)}$ converges when $b$ tends to infinity under very mild assumptions.

**Proposition 2.24.** — Assume that (2.9) holds, and that $u_0$ satisfies (2.25). Then for any given $t \in [0,T]$ and $x \in \mathbb{R}^d$, for any $a > 0$

$$w^a(t,x) := \int_{\mathbb{R}^d} \mathcal{Z}_{a}^{\omega,a}([0,y),(t,x)] u_0(dy)$$

is almost surely finite.

**Remark 2.25.** — When $\int_{[1,\infty)} v \lambda(dv) < \infty$, and for bounded initial conditions, applying Theorem 1.2.1 in [45] we get that $w^a(t,x)$ is the unique solution of (2.23). For noise with heavier tails, $w^a(t,x)$ should also be a solution of (2.23) and coincide with the solution considered in [23] whenever it is well-defined. Since this is not the main
focus of the paper we do not include a proof of this statement, which in any case
would only provide a minor extension on the class of noises considered [23] which
includes \( \int_{[1,\infty]} v^p \lambda(dv) < \infty \) for all \( p > 0 \). We do not have an argument establishing
uniqueness in that case.

Let us now present the result. It establishes the convergence of \( u^a \) when \( a \) tends
to 0. While the limit is the natural candidate to be a solution to (2.23) under less
restrictive assumptions than those considered in [23], we could not verify that \( u \) solves
the equation.

**Proposition 2.26.** — Assume that (2.9)–(2.10) are satisfied. Given \( u_0 \) a locally finite
signed measure on \( \mathbb{R}^d \) satisfying (2.25), then for \( (t,x) \in [0,T] \times \mathbb{R}^d \) the integral defining
\( u(t,x) \) in (2.22) is almost surely finite (and well-defined) and for every fixed \( t \in [0,T] \)
we have

\[
\lim_{a \to 0} u^a(t,x) = u(t,x),
\]

except on a set (of \( \mathbb{R}^d \)) of Lebesgue measure zero.

**Remark 2.27.** — When (2.24) is satisfied and the initial condition has a bounded
density w.r.t to the Lebesgue measure, it follows from results of [23] and [45] that \( u \)
is the solution of the (2.23) constructed in [23, Th. 3.1].

2.5. Further discussion on the results. — Let us now comment further on our
results, and explain how they compare with the literature, how they can be extended
and what interesting open questions remain to be solved.

2.5.1. Scaling properties in the case of \( \alpha \)-stable Lévy noise. — Let us come back further
on the case of the \( \alpha \)-stable noise, that is when \( \lambda(dv) = \alpha v^{-(1+\alpha)} dv \), with \( \alpha \in (0,2) \).
We have already seen that in that case Assumptions (2.9)–(2.10) are satisfied provided
that \( \alpha \in (0,\alpha_c) \), so Theorem 2.8 and more importantly (2.13) holds, so that \( Q^{\omega}_{T,\beta} \) is
well-defined. Now, notice that in the \( \alpha \)-stable case the Poisson point process \( \omega \) has
the following scaling property

\[
\omega = (d) \{ (rt,sx, (rs^d)^{1/\alpha} v) : (t,x,v) \in \omega \}
\]

for any \( r,s > 0 \). Using additionally the Brownian scaling, one can then check that the
continuum polymer in \( \alpha \)-stable Lévy environment satisfies the following scaling
property: if \( \alpha \in (0,\alpha_c) \), for all \( r > 0 \)

\[
Q^{\omega}_{T,\beta}(A) \overset{(d)}{=} Q^{\omega}_{rT, r^{-\zeta}\beta}(A_r), \quad \text{with} \quad \zeta = \frac{d}{2\alpha}(1 + 2/d - \alpha).
\]

where \( A_r := \{ \varphi_r : t \mapsto (1/\sqrt{r})\varphi(t/r), \varphi \in A \} \).

2.5.2. SHE with Lévy noise: advantages and disadvantages of our method

Let us now compare our Proposition 2.26 with the results of Chong [23]. First of all,
as we already stressed in Section 2.4, our Proposition 2.26 gives a weaker notion
of solution to the SHE (2.21) than existence of solutions in the integral form (2.23),

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as proved in [23]. Additionally, Chong’s results allows to deal with a larger class of integral equations
\[ Y(t, x) = \int_{\mathbb{R}^d} G(0, y; t, x)Y_0(y)dy + \int_0^t \int_{\mathbb{R}^d} G(s, y; t, x)\sigma(Y(s, y))M(ds, dy), \]
where: (i) \( M \) is a noise that can include a white noise part when \( d = 1 \) and a (signed) pure jump component; (ii) \( \sigma \) is a globally Lipschitz function; (iii) \( G(t, x; s, y) \) is measurable and dominated by a constant times the heat kernel \( \rho_{t-s}(x-y) \).

We have presented our results in the case where \( M = \xi_\omega \) (i.e., has no white noise component and only positive jumps), \( \sigma(Y) = Y \) and \( G(t, x; s, y) = \rho_{t-s}(x-y) \). Let us now present the advantages of our method, and in which directions it can be generalized.

(a) First and foremost, our method enables us to make sense of Feynman–Kac formulas containing a functional \( f \) of the Brownian motion, that is \( Z_\omega^{\beta, f} \) (see Theorem 2.8(2.11)). This is something absolutely required to be able to define the continuum model.

(b) Our tail assumptions (2.9)–(2.10) on the Lévy measure are less restrictive than those (2.24) which are used in [23]. In particular our method allows to treat the integrability issues at 0 and \( \infty \) separately. Note also that in view of Propositions 2.6 and 2.12, our assumptions (2.9)–(2.10) are close to being optimal.

(c) We are able to deal with more singular initial conditions than in [23]. For the application we have in mind, it is of the utmost importance to be able to deal with Dirac initial condition, which corresponds to the point-to-point partition function \( Z_\omega^{\beta, f}(t, x) \) and appears to be excluded in [23].

(d) We can easily adapt our proof to the case of an arbitrary kernel \( \rho_t \) (in particular, not only the ones dominated by Gaussians), but this would require to adapt the conditions (2.10)–(2.9) In particular, we could replace the Laplacian \( \Delta \) with more general operators. For instance, in dimension 1, we could replace the Brownian motion by a Lévy process, see the paragraph below for further discussion.

(e) While our method does not seem to allow to treat the case of general Lipschitz \( \sigma \) (for which we lose the existence of a Feynman–Kac representation of the solution), let us mention, that it should extend without much problem to the case where \( \sigma(u) = au + b \) with \( a, b > 0 \) (that is, considering a mixture of additive and multiplicative noise).

(f) To conclude, we stress that maybe the most problematic part would be to extend our results to a more general noise. In particular, our method does not allow to deal with general complex (or signed) noise: the issue essentially arises in the proof of Proposition 4.1, which shows that \( (Z_\beta^{\omega, a})_{a \in (0, 1]} \) is uniformly integrable (if \( \int_{[1, \infty)} v\lambda(dv) < \infty \)); all the other points extend quite easily. In view of our techniques (in particular Sections 4.3–4.4), this appears to be manageable in dimension \( d = 1 \), but it is possibly more problematic in dimension \( d \geq 2 \) (the truncation we use is based on a multi-body functional that needs to be adapted in the case of a complex or signed noise). Similarly, in accordance with the literature on directed polymer

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models, adding a white-noise component should be feasible in dimension $d = 1$, but it is likely that in dimension $d \geq 2$ it would make the limit degenerate (in analogy with the SHE with multiplicative white noise in dimension $d = 2$, see [21, 35]).

2.5.3. Applications of our method to other disordered systems. — Our method appears robust enough to be adapted to the setting of other models with heavy-tail disorder. In particular, in analogy with [19], one should be able to consider several (discrete) models, and construct their continuum counterpart with Lévy noise. This includes for instance:

(A) the $(1 + d)$-dimensional long-range directed polymer, see [25, 49] for the case of dimension $d = 1$, where the underlying random walk $(S_n)_{n \geq 0}$ is in the $\gamma$-stable domain of attraction, with $\gamma \in (0, 2)$;

(B) the disordered pinning model, see [33] for an overview (it has been studied in [40] in the case of a heavy-tail noise).

We could also consider other disordered models, such as the copolymer model (see [32, Ch. 6] for an overview and [14, 16] for the question of the scaling limit) or the random field Ising model (see [15, Ch. 7] for an overview and [19] for the question of the scaling limit). We however chose to focus on the two examples (A)–(B) above, which might provide a sufficient illustration on how general our construction is. In both cases (A) and (B) we only briefly present the models and discuss how the assumptions (2.9)–(2.10) have to be adapted to ensure the convergence of the partition function. In order to be fully understood, the discussion below requires to be familiarized with the proof of our main result. It can be thus be skipped during the first reading.

(A) The continuum $(1 + d)$-dimensional long-range directed polymer in Lévy noise

The idea is to replace in the definitions the Brownian motion $(B_t)_{t \geq 0}$ by a $d$-dimensional $\gamma$-stable process $(X_t)_{t \geq 0}$ with $\gamma \in (0, 2)$, that we suppose centered and isotropic for simplicity. More precisely, we can define, analogously to (2.1), the partition function

\begin{equation}
Z_{\beta, \omega, a}^{\text{long}} := 1 + \sum_{k=1}^{\infty} \beta^k \int_0^{<t_1 < \cdots < t_k < T} \int_{(\mathbb{R}^d)^k} \prod_{i=1}^k \rho_t^{(\gamma)}(x_i - x_{i-1}) \prod_{i=1}^k \xi_a(d\omega, dx),
\end{equation}

where $\rho_t^{(\gamma)}(x)$ is the transition kernel of our $\gamma$-stable process, and is defined by

$$
\rho_t^{(\gamma)}(x) := \frac{1}{(2\pi)^d} \int_{\mathbb{R}^d} e^{-\|z\|^\gamma} \cos(x \cdot z) \, dz.
$$

While $\rho_t^{(\gamma)}$ does not admit a closed expression, its asymptotic properties are well known (dating back to [38], see also [46, Ch. 2]). It is a bounded radial function and has the following asymptotic behavior

$$
\rho_1^{(\gamma)}(x) \sim c_{d, \gamma} \|x\|^{-(d+\gamma)}, \quad \text{as } \|x\| \to \infty.
$$

The scaling relation $\rho_t^{(\gamma)}(x) = t^{-d/\gamma} \rho_1(t^{-1/\gamma} x)$ also implies that $\|\rho_t^{(\gamma)}\|_{\infty} = c_{d, \gamma} t^{-d/\gamma}$. 

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Now let us discuss under which condition on the Lévy measure $\lambda$ the partition function in (2.28) remains finite. Note first that if \( \int_{[1, \infty)} \nu \lambda(\nu) < \infty \), then we have 
\[
\mathbb{E}[Z_{\omega, a, \text{long}}] < \infty
\]
from Lemma 3.1 (or rather its straightforward adaptation to this case) and a discussion is necessary only for the integrability of heavier-tailed noises. In analogy with (2.9), we want to make sure that the weight of Poisson points with large intensity is compensated by the cost of making a long jump to visit them, which by (2.28) is of order \( \|x\|^{-(d+\gamma)} \). Hence we need a condition that ensures that
\[
\sup\{ \nu(1 + \|x\|)^{-(d+\gamma)} : (t, x, \nu) \in \omega, t \in [0, T] \} < \infty.
\]
We should require in fact a bit more than (2.29) but not much more (we opt not to stretch the discussion any further) and we believe a condition that ensures that 
\[
Z_{\omega, a, \text{long}} < \infty
\]
and thus replaces (2.9) in this case is
\[
(2.30) \quad \int_{[1, \infty)} \nu^q \lambda(d\nu) < \infty \quad \text{for some } q > \frac{d}{d+\gamma}.
\]
On the other hand, the condition (2.10) prevents the possible accumulation of small weights that would make the limiting partition function degenerate. It is intimately related to the local limit behavior of \( \rho^\gamma_1(x) \) at small times, more precisely to \( \int_{\mathbb{R}} \rho^\gamma_1(x) dx \) which by scaling is equal to \( t^{-d/\gamma} \int_{\mathbb{R}} \rho^\gamma_1(x)^2 dx \). In analogy with (2.10), a (near-optimal) condition that ensures that \( Z_{\omega, a, \text{long}} \) is non-degenerate should therefore be
\[
(2.31) \quad \int_{(0,1)} \nu^p \lambda(d\nu) < \infty, \quad \text{for some } p < \min(1 + \gamma/d, 2).
\]
We therefore conjecture that if (2.30)--(2.31) hold, then the partition function \( Z_{\omega, a, \text{long}} \) defined in (2.28) converges a.s. to a non-degenerate limit and that one can construct a continuum measure corresponding to the \((1+d)\)-dimensional long-range directed polymer in Lévy noise. Let us stress that in the case of an \( \alpha \)-stable noise (i.e., \( \lambda(\nu) = \alpha \nu^{-(1+\alpha)} d\nu \)), the conditions (2.30)--(2.31) translate into the condition \( 1/\gamma < \alpha < \min(1 + d/\gamma, 2) \). Additionally, in analogy with Theorem A, the continuum long-range directed polymer model in \( \alpha \)-stable noise should appear as the scaling limit of the long-range directed polymer model, defined as in (1.2) with a random walk \( (S_n)_{n \geq 0} \) in the domain of attraction of a \( \gamma \)-stable law and heavy tailed disorder satisfying (1.4).

(B) The continuum disordered pinning model in Lévy noise. — The disordered pinning model describes a renewal process \( \tau = \{ \tau_0 = 0, \tau_1, \tau_2, \ldots \} \) on \( \mathbb{N} \) (representing contact points) interacting with an inhomogeneous defect line. In the case of a heavy tailed environment \( (\eta_x)_{x \in \mathbb{N}} \), it is convenient to write the partition function of the model as follows, see [40]:
\[
(2.32) \quad Z_{N,\beta,h}^\eta := \mathbb{E} \left[ \prod_{n=1}^N e^{h \textbf{1}_{\{n \in \tau\}} (1 + \beta \eta_n \textbf{1}_{\{n \in \tau\}})} \right],
\]
where $h$ in an additional (homogeneous) pinning parameter. A standard (and natural) assumption in the literature is that $P(\tau_1 = n) = (1 + o(1))cn^{-1(1+\gamma)}$ as $n$ goes to infinity, for some $\gamma > 0$. Under this assumption, if $\gamma \in (0, 1)$, then the set of contact points $\tau \cap [0, N]$, properly scaled, converges to what is called the regenerative set of index $\gamma$. This leads us to make the following definition for the truncated partition function of the continuum disordered pinning model: for $\beta > 0$ and $h \in \mathbb{R}$,

$$Z_{\beta,h}^{\omega,a} := 1 + \sum_{k=1}^{\infty} \beta^k \int_{0 < t_1 < \cdots < t_k < 1} \prod_{i=1}^{k} u_\gamma(t_i - t_{i-1}) \prod_{i=1}^{k} (\xi_\omega(t_i) + hZ_\beta)(dt_i),$$

where $u_\gamma(t) := c_\gamma t^{-(1-\gamma)}$ is the transition kernel of the regenerative set of index $\gamma$.

Here, no condition analogue to (2.9) is needed to keep $Z_{\beta,h}^{\omega,a}$ a.s. finite, since there is no spatial dimension. On the other hand, in analogy with (2.10), in view of the form $u_\gamma(t) = c_\gamma t^{-(1-\gamma)}$ and since there is no spatial dimension, a (near-optimal) condition that ensures that $\lim_{a \to 0} Z_{\beta,h}^{\omega,a}$ is non-degenerate should therefore be

$$\int_{(0,1)} v^p \lambda(dv) < \infty, \text{ for some } p < \min(1/(1-\gamma), 2).$$

Hence, we conjecture that if (2.33) holds, the partition function $Z_{\beta,h}^{\omega,a}$ defined in (2.28) converges a.s. to a non-degenerate limit, and that one can construct a continuum measure corresponding to the disordered pinning model in Lévy noise. In the case of an $\alpha$-stable noise (i.e., $\lambda(dv) = \alpha v^{-(1+\alpha)}dv$), the condition (2.33) translates into $\alpha < \min(1/(1-\gamma), 2)$, which corresponds to the disorder relevance condition found in [40] (where the roles of $\gamma$ and $\alpha$ are exchanged). Additionally, in analogy with Theorem A, the continuum pinning model in $\alpha$-stable noise should then appear as the scaling limit of the disordered pinning model defined above in (2.32) and heavy tailed disorder satisfying (1.4).

2.5.4. Other open questions. — To conclude this section, we present a brief list of interesting open questions.

(a) A first question that we already raised is that of considering a more general noise. We leave as an open problem the issue of adding a Gaussian white-noise component to $\xi_\omega$. We believe that in dimension $d = 1$ the partition function converges to a non-degenerate limit even when a Gaussian component is added to the noise and that consequently on can define a continuum polymer in that case.

(b) Another natural question is that of the $L_p$ convergence in Theorem 2.8. It is natural to expect some $L_p$ convergence to hold, but this appears to be technically challenging. We leave as an open problem to show that, if $\int_{(0,1)} v^p \lambda(dv) < \infty$ for some $p < \min(1 + 2/d, 2)$ and $\int_{[1,\infty)} v^q \lambda(dv) < \infty$ for some $q \geq 1$, then $Z_{\beta,h}^{\omega,a}$ converges to $Z_{\beta}^{\omega,a}$ in $L_{\min(p,q)}$.

(c) To conclude, let us mention an important and challenging question. In the case of an $\alpha$-stable noise, we have treated the case $\alpha < \alpha_c = \min(1 + 2/d, 2)$; in particular, $\alpha_c < 2$ in dimension $d \geq 3$. It would then be an interesting question to investigate the case $\alpha = \alpha_c$, called marginal (in particular in the case where $\alpha_c < 2$, since marginal behavior may depend on $\alpha_c$). In analogy with other marginally relevant disordered
systems (see [20] in the context of scaling limits), one would then expect that one should find a non-trivial limit in distribution by allowing $\beta$ to depend on $a$ in such a way that $\lim_{a \to 0} \beta a = 0$ (recall that from Proposition 2.12 we have $\lim_{a \to 0} Z_{\omega,a}^T = 0$ in that case). It seems likely (again considering the analogy with [20]) that the right scaling for $\beta_a$ should be of the form $\beta_a = c(\log a)^{-\nu} (1 + o(1))$ where $c$ and $\nu$ are positive constant.

**Remark 2.28.** — Since the first version of this paper appeared as a draft, progress has been made in [6] on several fronts including a few of the open questions mentioned above. We refer to the introduction of [6] for a full account.

2.6. **Organization of the rest of the paper.** — Let us briefly present how the rest of the paper is organized and outline the ideas of the proofs of Theorems 2.1 and 2.8.

– In Section 3 we present preliminary results concerning the partition function with truncated noise $Z_{\omega,a}^T,\beta$ that are needed in the rest of the paper. We prove in particular its well-posedness (Proposition 3.1), its positivity (Lemma 3.3, which provides an important alternative representation for the partition function), and a martingale property (under suitable integrability condition, see Lemma 3.5). We also give an enlightening representation of the size-biased law of the environment (i.e., its law biased by the partition function, see the definition (3.11)) and we recall Mecke’s multivariate equation for Poisson point processes, which is used throughout the paper.

– In Section 4, we prove our main result, that is, Theorem 2.8 (Theorems 2.1 and 2.3 being only particular cases). The proof needs to be decomposed in several steps, a detailed account of which is given in Section 4.1. Most of the proofs of this section can be adapted to control the point-to-point partition function, and thus we prove along the way Proposition 2.20 and Proposition 2.24.

– In Section 5, we study the cases where the limiting partition function degenerates either to zero or infinity, that is we prove Proposition 2.6 and Proposition 2.12.

– In Section 6, we prove the various properties of the continuum directed polymer in Lévy noise that are gathered in Section 2.3, that is Proposition 2.16, Proposition 2.18 and Proposition 2.21.

– In Section 7, we prove our statement concerning the convergence of the solution of the SHE with truncated noise, Proposition 2.26.

– Finally, we collect in the appendix several technical results that are used along the paper.

**Notational warning.** — For simplicity we assume in the rest of the paper that $T = 1$, and we drop the dependence in $T$ in all notations.

3. **Preliminaries: some properties of $Z_{\beta}^\omega,a$**

We let $|\xi^{(\omega)}|$ denote the total variation associated with the locally finite signed measure $\xi^{(\omega)}$ defined in (2.5), and we let

$$X^k := \{ t = (t_1, \ldots, t_k) \in \mathbb{R}^k : 0 < t_1 < \cdots < t_k < 1 \},$$

denote the open simplex.
3.1. Well-posedness. — Our first task is to check that our definitions \( \mathcal{Z}_\beta^{\omega,a} \) in (2.1) and \( \mathcal{Z}_\beta^{\omega,a}(f) \) in (2.7) are well posed. This is given by the following result.

**Proposition 3.1.** — For any choice of \( \lambda \) satisfying \( \lambda([a,\infty)) < +\infty \) for every \( a > 0 \), for any \( f \in \mathcal{B}_h \), the function \( \varrho(t, x, f) \) defined on \( \mathcal{X}^k \times (\mathbb{R}^d)^k \) is almost surely integrable with respect to the product measure \( |\xi^{(a)}| \otimes k \). Moreover we almost surely have, for any \( \beta > 0 \),

\[
(3.1) \quad \sum_{k=0}^{\infty} \beta^k \int_{\mathcal{X}^k \times (\mathbb{R}^d)^k} \varrho(t, x, f) \prod_{i=1}^{k} |\xi^{(a)}(i)| (dt_i, dx_i) < \infty.
\]

If \( \mu := \int_{[1,\infty)} \nu \lambda(dv) < \infty \), then \( \varrho(t, x) \) is integrable with respect to the product measure \( |\xi^{(a)}| \otimes k \) and

\[
(3.2) \quad \sum_{k=0}^{\infty} \beta^k \int_{\mathcal{X}^k \times (\mathbb{R}^d)^k} \varrho(t, x) \prod_{i=1}^{k} |\xi^{(a)}(i)| (dt_i, dx_i) < \infty.
\]

Furthermore, we have for all \( f \in \mathcal{B}_h \)

\[
(3.3) \quad \forall a \in (0,1], \quad \mathbb{E}[\mathcal{Z}_\beta^{\omega,a}(f)] = e^{\beta \mu} Q(f).
\]

**Proof.** — Let us start with the case \( \int_{[1,\infty)} \nu \lambda(dv) < \infty \). It is sufficient to check that the expectation of the l.h.s. in (3.2) is finite. Now using the definition (2.5) for \( \xi^{(a)} \) we have on \( \mathcal{X}^k \times (\mathbb{R}^d)^k \),

\[
\mathbb{E} \left[ \bigotimes_{i=1}^{k} |\xi^{(a)}(i)| (dt_i, dx_i) \right] = \left( \int_{[a,\infty)} \nu \lambda(dv) + \kappa_a \right)^k \prod_{i=1}^{k} dt_i, dx_i.
\]

Letting \( C_a := \int_{[a,\infty)} \nu \lambda(dv) + \kappa_a < \infty \), we therefore get that

\[
\mathbb{E} \left[ \int_{\mathcal{X}^k \times (\mathbb{R}^d)^k} \varrho(t, x) \prod_{i=1}^{k} |\xi^{(a)}(i)| (dt_i, dx_i) \right] = C_a \int_{\mathcal{X}^k \times (\mathbb{R}^d)^k} \varrho(t, x) \prod_{i=1}^{k} dt_i, dx_i = \frac{C_a^k}{k!}.
\]

This implies both the convergence of the integral and the summability in \( k \).

The fact that \( \mathbb{E}[\mathcal{Z}_\beta^{\omega,a}(f)] = e^{\beta \mu} Q(f) \) directly follows from the definition (2.6)–(2.7) and Fubini, using that \( \prod_{i=1}^{k} |\xi^{(a)}(i)| (dt_i, dx_i) \) has mean \( (\mu, \mathcal{Z}) \otimes k \).

Now let us prove (3.1) when \( \int_{[1,\infty)} \nu \lambda(dv) = \infty \). For this we first consider a truncated version of the noise to place ourselves back in the integrable case, and then let the truncation threshold go to infinity. This procedure is going to be used repeatedly in the paper. For \( b > a \), recall the definition (2.26) of \( \xi^{(a,b)}_\omega \). Using the assumption \( f \in \mathcal{B}_h \), we let \( M > 0 \) be such that \( f(\varphi) = 0 \) if \( \|\varphi\|_\infty \geq M \). Then \( \varrho(t, x, f) = 0 \) if \( \max_{i=1}^{k} |x_i|_\infty \geq M \). Therefore, since \( \lambda([a,\infty)) < \infty \), there exists \( b_0(M, \omega) \) such that for every \( b > b_0 \), the restriction of \( \xi^{(a,b)}_\omega \) on \( [0,1] \times [-M,M]^d \) coincides with that of \( \xi^{(a,b)}_\omega \). Hence it is sufficient to show that (3.1) holds for \( \xi^{(a,b)}_\omega \) for every \( b > 1 \), which we can do by repeating the proof of (3.2).
Remark 3.2. — Notice that we have the analogous result for the point-to-point partition function $X_\omega^\omega(t, x)$. For any $x \in \mathbb{R}^d$, the function $g(t, x)\rho_{t-t_x}(x-x_k)$ is almost surely integrable on $\mathbb{R}^k \times [-M, M]^k$ for any $M > 0$ with respect to the product measure $|\xi^{(a)}|$, and it is integrable on $\mathbb{R}^k \times (\mathbb{R}^d)^k$ if $\int_{[1, \infty[} v \lambda(\mathrm{d}v) < \infty$. We also have

$$\forall a \in (0, 1], \quad \mathbb{E}[X_\omega^\omega(t, x)'] = e^{\beta t_t} \rho_t(x).$$

3.2. The partition function $X_\omega^\omega$ as a sum. — Let us now present an alternative expression for $X_\omega^\omega$, from which it will be clear that $X_\omega^\omega$ is positive. Indeed, it is not obvious that the chaos decomposition (2.7) is non-negative, due to the term $-\kappa a \mathcal{L}$ in the definition (2.5) of $\xi^{(a)}$. The idea of Lemma 3.3 below is to integrate out this “Lebesgue part” of $\xi^{(a)}$. This operation has the effect of giving rise to a prefactor $e^{-\kappa a}$. We let $\mathcal{P}_{[0, t]}(\omega)$ denote the set of finite collections of points in $\omega$ whose time coordinates belong to the interval $[0, t]$. When $t = 1$ we simply write $\mathcal{P}(\omega)$. We define similarly $\mathcal{P}_{[0, t]}(\omega(a))$ and $\mathcal{P}(\omega(a))$ the sets of finite collections of points in $\omega(a) = \{(t, x, v) \in \omega : v \geq a\}$ (as defined above (1.6)). For $\sigma \in \mathcal{P}(\omega)$ we let $|\sigma|$ denote its cardinality and we use the notation $(t_i, x_i, u_i)_{i=1}^{|\sigma|}$ to denote the points in $\sigma$ ordered in increasing time. Given $a > 0$ we define the following weight function $w_{a, \beta}(\sigma)$ on $\mathcal{P}(\omega)$

$$w_{a, \beta}(\sigma) := e^{-\beta \kappa a \beta |\sigma|} g(t, x) \prod_{i=1}^{|\sigma|} u_i \mathbb{1}_{\{u_i > a\}},$$

with $\kappa a = \int_{[a, 1[} v \lambda(\mathrm{d}v)$ as defined in (2.6). Let us stress here that $w_{a, \beta}$ puts a positive weight only on elements of $\mathcal{P}(\omega(a))$. By convention, we say that the empty set belongs to $\mathcal{P}(\omega)$ and we set $w_{a, \beta}(\emptyset) := e^{-\beta \kappa a}$. Similarly, for any $f \in \mathcal{B}$, we define

$$w_{a, \beta}(\sigma, f) := e^{-\beta \kappa a \beta |\sigma|} g(t, x, f) \prod_{i=1}^{|\sigma|} u_i \mathbb{1}_{\{u_i > a\}},$$

$$w_{a, \beta}(\emptyset, f) := e^{-\beta \kappa a} Q(f).$$

The following lemma provides an alternative formulation of the partition function which is convenient to assert positivity. Its proof is straightforward (one only needs to integrate out the Lebesgue part of $\xi^{(a)}$) and is presented in Appendix B.1 for completeness.

Lemma 3.3. — Given $a > 0$, $\beta > 0$ and $f \in \mathcal{B}$ we have

$$X_\omega^\omega(f) = \sum_{\sigma \in \mathcal{P}(\omega)} w_{a, \beta}(\sigma, f).$$

In particular $X_\omega^\omega(f) > 0$ if $f \geq 0$ and $Q(f) > 0$. Also, by monotone convergence, recalling the definition (2.8), we have

$$X_\omega^\omega = \sum_{\sigma \in \mathcal{P}(\omega)} w_{a, \beta}(\sigma).$$
Let us stress that the representations (3.7) and (3.8) are valid without any assumption on the intensity measure \( \lambda \), since all the terms in the sums are positive, but it may be the case that both sides of (3.8) are infinite.

**Remark 3.4.** — Similarly, for the point-to-point partition function, the following identity holds

\[
\mathcal{Z}^{\omega_a}(t, x) = \sum_{\sigma \in \mathcal{P}[a, \sigma](\omega)} w_{a, \sigma}(\sigma, (t, x)),
\]

with \( w_{a, \sigma}(\sigma, (t, x)) := e^{-\beta \kappa_a g'(t, x)} \rho_{t-x} (x-x_k) \prod_{i=1}^{\vert \sigma \vert} u_{\{a, \sigma_i\}} \), as long as the r.h.s. is finite. Note that by Fubini’s theorem this gives that, for any \( w_x \),

\[
\mathcal{Z}^{\omega_a}(t, x) = \mathcal{Z}^{\omega_a}(x, t) = \int_{\mathbb{R}^d} \mathcal{Z}^{\omega_a}(t, x) dx = \mathcal{Z}^{\omega_a}(t, x).
\]

### 3.3. Martingale property

**Remark 3.5.** — In the case \( \int_{[1, \infty)} v \lambda(dv) < \infty \), the convergence of \( \mathcal{Z}^{\omega_a}(f) \) as \( a \downarrow 0 \) is an immediate consequence of the following observation.

**Lemma 3.5.** — Let \( \mathcal{F} = (\mathcal{F}_a)_{a \in (0,1]} \) be the filtration where \( \mathcal{F}_a \) is the \( \sigma \)-field generated by \( \omega \). If the measure \( \lambda \) satisfies \( \mu := \int_{[1, \infty)} v \lambda(v)dv < \infty \), then the following processes are (time-reversed) martingales for the filtration \( \mathcal{F} \):

- \( \mathcal{Z}^{\omega_a}(f)_{a \in (0,1]} \) for any \( f \in \mathcal{B} \), and in particular \( \mathcal{Z}^{\omega_a}(c)_{a \in (0,1]} \);
- \( \mathcal{Z}^{\omega_a}(t, x)_{a \in (0,1]} \) for any \( (t, x) \in \mathbb{R}_+ \times \mathbb{R}^d \).

The mean of these martingales are \( \mathbb{E} [\mathcal{Z}^{\omega_a}(f)] = e^{by} Q(f) \) and \( \mathbb{E} [\mathcal{Z}^{\omega_a}(t, x)] = e^{\beta t} \rho_t(x) \).

Moreover, if \( g \) is a bounded measurable function of \( \varphi \) and \( \omega \) and \( g(\varphi, \omega) \) is \( \mathcal{F}_{\alpha_0} \)-measurable for every \( \varphi \), then \( \mathcal{Z}^{\omega_a}(g(\cdot, \omega))_{a \in (0,\alpha_0]} \) is a (time-reversed) martingale.

**Proof.** — Using the expression (2.7) (or (2.17) for the point-to-point partition function), the result follows from the fact that the sequence of measures

\[
\left( \prod_{i=1}^{k} \xi^{(a)}_{\omega}(dt_i, dx_i) \right)_{a \in (0,1]}
\]

is a martingale. Indeed for \( b < a \leq 1 \) we have

\[
\mathbb{E} \left[ \prod_{i=1}^{k} \xi^{(b)}(dt_i, dx_i) - \prod_{i=1}^{k} \xi^{(a)}(dt_i, dx_i) \bigg| \mathcal{F}_a \right] = 0,
\]

where in the last equality we used that by construction \( \xi^{(b)} - \xi^{(a)} \) is of zero average, independent of \( \mathcal{F}_a \) and conditionally independent of \( \prod_{j=1}^{k} \xi^{(b)}(dt_j, dx_j) \).

The proof for a random function follows the same line, using that \( g(t, x, g(\cdot, \omega)) \) is \( \mathcal{F}_{\alpha_0} \)-measurable for all \( t \) and \( x \).
Since $Z_{\omega,a}^{\beta} \geq 0$, this directly implies in the case $\int_{[1,\infty)} v \lambda(v) dv < \infty$ that $\lim_{a \downarrow 0} Z_{\omega,a}^{\beta}$ exists almost surely. We will show that if additionally assumption (2.10) holds, the martingale is uniformly integrable.

**Remark 3.6.** — In the case $\int_{[1,\infty)} v \lambda(v) dv = \infty$, we will consider the truncated partition function $Z_{\beta}^{[a,b)}(f)$ defined as in (2.3) but with $\xi_{\omega}^{(a)}$ replaced by the truncated noise $\xi_{\omega}^{[a,b)}$ defined in (2.26) — this corresponds to considering the intensity measure $\lambda^{(0,b)}(dv) := 1_{\{v < b\}} \lambda(dv)$. Then, for any $b > 0$, Lemma 3.5 shows that $(Z_{\beta}^{[a,b)}(f))_{a \in (0,b]}$ is a martingale.

### 3.4. A representation for the size-biased measure

When $\mu := \int_{[1,\infty)} v \lambda(v) dv < \infty$, since

$$Z_{\beta}^{\omega,a} := e^{-\beta \mu} Z_{\omega,a}^{\beta},$$

is non-negative and of average one, one can define an alternative measure $\tilde{\mathbb{P}}_{\beta}^{a}$ for the environment, defined by

$$\tilde{\mathbb{P}}_{\beta}^{a}(\omega \in A) := \mathbb{E}\left[Z_{\omega,a}^{\beta} 1_{\omega \in A}\right].$$

(3.11)

The measure $\tilde{\mathbb{P}}_{\beta}^{a}$ is often referred to as the size-biased measure — the probability of an event is biased by the “size” of the partition function. Convenient representations of the size-biased measure have been given for directed polymers [12, Lem. 1] and similar models such as branching random walks (see [47, Ch. 4] and references therein) or the disordered pinning model [40, §5.2]. The size-biased measure for all these models is obtained by tilting the distribution of the environment along a randomly chosen trajectory. The result we present below is a strict analog in a continuous setup.

We let $\mathbb{P}_{a}^{\alpha}$ be the distribution of a Poisson point process $\omega_{a}^{\alpha}$ on $[0,1] \times \mathbb{R}_{+}$ whose intensity is $dt \otimes \beta v \alpha v^{-\alpha}1_{\{v \geq a\}} \lambda(dv)$, (that is $dt \otimes \beta \alpha v^{-\alpha}1_{\{v \geq a\}} \lambda(dv)$ in the $\alpha$-stable case) and we recall that $\mathbb{Q}$ is the distribution of a standard Brownian motion. We then introduce the random set of points $\hat{\omega}(\omega, \omega_{a}^{\alpha}, B)$ in $\mathbb{R} \times \mathbb{R}^{d} \times \mathbb{R}_{+}$ defined by

$$\hat{\omega} := \omega \cup \{(t, B_s, v) : (t, v) \in \omega_{a}^{\alpha}\}.$$  

Then, the distribution of $\omega$ under the measure $\tilde{\mathbb{P}}_{\beta}^{a}$ can be described as follows.

**Lemma 3.7.** — Suppose that $\mu := \int_{[1,\infty)} v \lambda(v) dv < \infty$. Then with the notation defined above, for any measurable bounded function $g$ we have

$$\tilde{\mathbb{P}}_{\beta}^{a}[g(\omega)] = \mathbb{P} \otimes \mathbb{P}_{a}^{\alpha} \otimes \mathbb{Q}[g(\hat{\omega}(\omega, \omega_{a}^{\alpha}, B))].$$

(3.12)

In other words, the distribution of $\omega$ under $\tilde{\mathbb{P}}_{\beta}^{a}$ is obtained by adding to the original point process an independent Poisson process of intensity $dt \otimes \beta v \alpha v^{-\alpha}1_{\{v \geq a\}} \lambda(dv)$ drawn on the trajectory a Brownian motion.

The proof, though elementary, requires some cumbersome computation. We present it in Appendix B.2 for completeness.
3.5. An important tool: Mecke’s multivariate equation. — Let us recall here a classical formula for Poisson point processes which we will repeatedly use in our computations. It is a particular case of Mecke’s multivariate equation (see e.g. [41, Th. 4.4]).

Proposition 3.8. — Given \( \lambda \) a sigma-finite measure on a measurable space \((X, \mathcal{F})\), and \( \omega \) a Poisson point process with intensity \( \lambda \), then for any \( k \in \mathbb{N} \) and any measurable function \( g : X^k \to \mathbb{R}_+ \) such that \( g(x_1, \ldots, x_k) = 0 \) as soon as \( x_i = x_j \) for some \( i \neq j \) then

\[
E \left[ \sum_{(x_1, \ldots, x_k) \in \omega^k} g(x_1, \ldots, x_k) \right] = \int_{X^k} g(x_1, \ldots, x_k) \lambda^\otimes k(dx_1, \ldots, dx_k).
\]

Of course we are going to apply this formula for the Poisson process \( \omega \). In our applications we mostly deal with sums running on subsets of \( \omega \) whose cardinality is not fixed, see the expression (3.8) for the partition function above. Hence, in practice, the formula we will use is rather

\[
E \left[ \sum_{k \geq 1} \sum_{(x_1, \ldots, x_k) \in \omega^k} g_k(x_1, \ldots, x_k) \right] = \sum_{k \geq 1} \int_{X^k} g_k(x_1, \ldots, x_k) \lambda^\otimes k(dx_1, \ldots, dx_k),
\]

where \( g_k \) is a sequence of positive functions on \( X^k \).

4. Convergence of the partition function and of the measure: proof of Theorem 2.8

4.1. Organization of the section. — We decompose the proof of the theorem in several steps. We provide the details of this decomposition before going to the core of the proof.

First step. — Our first and main task is to prove the convergence of the partition function under the additional assumption \( \int_{[1, \infty)} v \lambda(dv) < \infty \).

Proposition 4.1. — If the measure \( \lambda \) satisfies \( \int_{[1, \infty)} v \lambda(dv) < \infty \) and (2.10), then the martingale \( (\mathcal{F}^\omega_\beta)_{a \in (0,1]} \) is uniformly integrable. As a consequence there exists \( \mathcal{F}^\omega_\beta \) such that the following convergence holds almost surely and in \( L_1 \),

\[
\lim_{a \to 0} \mathcal{F}^\omega_\beta = \mathcal{F}^\omega_\beta.
\]

Since from Lemma 3.5 we know that \( (\mathcal{F}^\omega_\beta)_{a \in (0,1]} \) is a positive martingale, it is sufficient to show that \( (\mathcal{F}^\omega_\beta)_{a \in (0,1]} \) is uniformly integrable. Our strategy consists in considering a sequence of approximation \( (\mathcal{F}^\omega_{\beta,q})_{q \geq 1} \) of \( \mathcal{F}^\omega_\beta \), obtained by somehow restricting the partition function to “not-too-large” weights. We choose our restriction so that two key properties are satisfied:

(A) For large \( q \)’s, \( \mathcal{F}^\omega_{\beta,q} \) is a good approximation of \( \mathcal{F}^\omega_\beta \) in \( L_1 \), uniformly in \( a \).

(B) For any \( q \), \( (\mathcal{F}^\omega_{\beta,q})_{a \in (0,1]} \) is bounded in \( L_2 \).

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We refer to Section 4.2 for a more detailed description of this strategy, which is then implemented in Section 4.3 in dimension \( d = 1 \) and in Section 4.4 in dimension \( d \geq 2 \), where the restriction strategy is more subtle.

Notice that from Lemma 3.3 (in particular (3.8)), we have that \( Z_{\omega,a}^\beta(f) \leq \|f\|_\infty Z_{\omega,a}^\beta \) for any \( f \in \mathcal{B} \). Hence an immediate consequence of Proposition 4.1 is the following.

**Corollary 4.2.** — If the measure \( \lambda \) satisfies \( \int_{[1,\infty]} v\lambda(dv) < \infty \) and (2.10), then for every \( f \in \mathcal{B} \) the martingale \( (Z_{\omega,a}^\beta(f))_{a \in (0,1]} \) is uniformly integrable, and the following convergence holds almost surely and in \( L_1 \)

\[
\lim_{a \to 0} Z_{\omega,a}^\beta(f) = Z_\omega^\beta(f).
\]

**Second step.** — Our second task is to remove the assumption \( \int_{[1,\infty]} v\lambda(dv) < \infty \) from Proposition 4.1 (and Corollary 4.2), i.e., to prove Proposition 2.5. This is done in Section 4.5. Along the way we also prove Proposition 2.24 and the first part of Proposition 2.20, that is (2.18). We then use Proposition 2.5 to prove the following lemma (which corresponds to (2.11)–(2.12)), in Section 4.6.

**Lemma 4.3.** — Under the assumption (2.10), for every \( f \in \mathcal{B} \) the following convergence holds almost surely and the limit is finite

\[
\lim_{a \to 0} Z_{\omega,a}^\beta(f) = Z_\omega^\beta(f).
\]

Furthermore if (2.9) also holds then the statement is valid for \( f \in \mathcal{B} \). In particular, we have

\[
\lim_{a \to 0} Z_{\omega,a}^\beta = Z_\omega^\beta < +\infty.
\]

**Third step.** — Our third task, which is crucial for the convergence of \( Q_\omega^\beta \), is to ensure that the limiting partition function is positive (let us record the statement as a proposition). This is done in Section 4.7.

**Proposition 4.4.** — If \( \lambda \) satisfies (2.10), then for any non-negative \( f \in \mathcal{B} \) with \( Q(f) > 0 \), we have almost surely

\[
Z_\omega^\beta(f) > 0.
\]

As a consequence, if \( \lambda \) satisfies (2.9)–(2.10) then we have almost surely

\[
(4.1) \quad Z_\omega^\beta \in (0, \infty).
\]

**Fourth step.** — Finally we complete the proof of Theorem 2.8 by proving the convergence of \( Q_\omega^\beta \). Note that Lemma 4.3 and Proposition 4.4 imply for any given \( f \in \mathcal{B} \) the almost sure convergence of \( Q_\omega^\beta(f) \). Hence we only need to prove tightness.

**Proposition 4.5.** — If \( \lambda \) satisfies (2.9)–(2.10), then for almost every \( \omega \), the family of measures \( (Q_\omega^\beta)_{a \in (0,1]} \) is tight with respect to the topology of weak convergence on \( \mathcal{M}_\mathcal{T} \), the set of probability measures on \( C_0([0,1]) \).
The reader can then check that combining all the statements above yields the complete proof of Theorem 2.8.

Let us finally comment on how the proof of the second part of Proposition 2.20 (the convergence (2.19) to a positive limit) is completed. We simply need to show that Proposition 4.1 and Proposition 4.4 (that is Equation (4.1)) remain valid for the point-to-point partition function. Since the proofs are nearly identical we will point at the end of the various proofs which modifications are required, when there are any.

4.2. A uniform integrability criterion. — As outlined above, our proof of uniform integrability is going to rely on a second moment computations. This requires to overcome some subtleties since the second moment of $Z_{\omega,a}^{\omega,a}$ might be infinite for every $a > 0$ (this is for instance the case when $d \geq 2$). We follow an approach similar to the one used in [5] for the proof of the convergence of Gaussian Multiplicative Chaos. We look for a family of restrictions of the partition functions which is bounded in $L^2$ but does not produce any loss of mass at infinity. Let us summarize our approach in the form of a proposition.

**Proposition 4.6.** — Consider $(X_a)_{a \in (0,1]}$ a collection of positive random variables. Assume that there exists $X^{(q)}_a$ a sequence of approximation of $X_a$, indexed by $q \geq 1$, which satisfies:

$$\lim_{q \to \infty} \sup_{a \in (0,1]} \mathbb{E}[|X^{(q)}_a - X_a|] = 0;$$

$$\sup_{a \in (0,1]} \mathbb{E}[(X^{(q)}_a)^2] < \infty \quad \text{for every } q \geq 1.$$

Then $(X_a)_{a \in (0,1]}$ is uniformly integrable.

**Proof.** — We may write, for any $M > 0$ and $a > 0$,

$$\mathbb{E}[|X_a| 1_{(|X_a| > M)}] \leq \mathbb{E}[|X^{(q)}_a - X_a|] + \mathbb{E}[(X^{(q)}_a)^2]^{1/2} \mathbb{P}(|X_a| > M)^{1/2},$$

where we have used Cauchy–Schwarz inequality for the second term. Applying Markov’s inequality and taking the supremum over $a \in (0,1]$, we therefore get

$$\sup_{a \in (0,1]} \mathbb{E}[|X_a| 1_{(|X_a| > M)}] \leq \sup_{a \in (0,1]} \mathbb{E}[|X^{(q)}_a - X_a|] + M^{-1/2} \sup_{a \in (0,1]} \mathbb{E}[(X^{(q)}_a)^2]^{1/2} \sup_{a \in (0,1]} \mathbb{E}[|X_a|]^{1/2}.$$

The first term can be made smaller than $\epsilon/2$ by choosing $q$ sufficiently large. Then once $q$ is fixed, we can make the second term smaller than $\epsilon/2$ by choosing $M$ large (our assumptions imply that $(X_a)_{a \in (0,1]}$ is bounded in $L_1$). □

Our idea is now to apply Proposition 4.6 to variables $X^{(q)}_a$ which are obtained by considering the sum of the weights $w_{a,\beta}(\sigma)$ on a strict subset of $P(\omega)$ (recall the representation of $Z_{\omega,a}^{\omega,a}$ in Lemma 3.3).
4.3. Proof of Proposition 4.1 in dimension $d = 1$. — The case $d = 1$ gives us the occasion to apply Proposition 4.6 with a relatively simple setup. In this case, the only thing that prevents $Z_{\omega,a}^\omega$ from being bounded in $L^2$ are the large values of $u_i$. The modified partition function obtained by ignoring these points in the Poisson point process $\omega$ turns out to be bounded in $L^2$. The idea is thus to apply Proposition 4.6 for partition functions with truncated environment, taking $X_{\beta}^{(q)} = Z_{\beta}^{\omega,[a,q]}$ (recall the definition after (2.27)). We then prove the following.

**Proposition 4.7.** — Suppose that $\lambda$ satisfies $\mu := \int_{[1,\infty)} v\lambda(dv) < \infty$ and that (2.10) holds. For every $d \geq 1$, we have

$$
\lim_{q \to \infty} \sup_{a \in (0,1]} E \left[ \left| Z_{\omega,a}^\omega - Z_{\beta}^{\omega,[a,q]} \right| \right] = 0.
$$

Moreover, when $d = 1$, we additionally have that for every $q \geq 1$

$$
\sup_{a \in (0,1]} E \left[ \left( Z_{\beta}^{\omega,[a,q]} \right)^2 \right] < \infty.
$$

These statements imply that both requirement of Proposition 4.6 are satisfied and therefore that $(Z_{\omega,a}^\omega)_{a \in (0,1]}$ is uniformly integrable and converges in $L^1$. As can be checked from the proof, (4.5) is false when $d \geq 2$ and in that case we will need a more subtle restriction for the set of trajectories (developed in the next subsection). While the latter restriction also covers the $d = 1$ case, the proof presented in this section is considerably simpler, and may prepare the reader for the more involved proof in dimension $d \geq 2$. Additionally note that (4.4) is valid when $d \geq 2$; it will be used in Section 4.4.

**Proof of Proposition 4.7.** — To compute the expectation in (4.4) we use Proposition 3.8: recalling the definitions (3.5) of $w_{a,\beta}(\sigma)$ and (2.6) of $\kappa_{a}$, we obtain that

$$
E \left[ Z_{\omega,a}^\omega - Z_{\beta}^{\omega,[a,q]} \right] = E \left[ \sum_{\sigma \in \mathcal{P}(\omega)} w_{a,\beta}(\sigma) \mathbf{1}_{\{\exists i \in [1,|\sigma|], u_i \geq q\}} \right]
$$

$$
= e^{-\beta \kappa_a} \sum_{k \geq 1} \beta^k \int_{X^k \times (\mathbb{R}^d)^k \times (0,\infty)^k} \prod_{i=1}^k u_i \mathbf{1}_{\{u_i \geq q\}} \rho(t,x) dt dx \prod_{i=1}^k u_i \lambda(du_i).
$$

Note that the integral in $x$ and $t$ readily simplifies since we have (recall that $\rho_t(x)$ is a probability density)

$$
\int_{X^k} \rho(t,x) dx = 1 \quad \text{and} \quad \int_{X^k} dt = \frac{1}{k!}.
$$

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Hence, setting \( \mu := \int_{[1, \infty)} v \lambda(dv) \) and \( \mu_q := \int_{[1, q)} v \lambda(dv) \), the r.h.s. of (4.6) is equal to

\[
e^{-\beta \kappa_a} \sum_{k \geq 1} \frac{\beta^k}{k!} \int_{[a, \infty)^k} \prod_{i=1}^k u_i \lambda(du_i) = e^{-\beta \kappa_a} \sum_{k \geq 1} \frac{\beta^k}{k!} \left[ \int_{[a, \infty)^k} \prod_{i=1}^k u_i \lambda(du_i) - \int_{[a, q)^k} \prod_{i=1}^k u_i \lambda(du_i) \right]
\]

(4.8)

\[
e^{-\beta \kappa_a} \sum_{k \geq 1} \frac{\beta^k}{k!} (\kappa_a + \mu)^k - (\kappa_a + \mu_q)^k = e^\beta \mu - e^\beta \mu_q.
\]

Hence we have \( \mathbb{E}[\mathcal{L}_{\beta}^{\omega_a} - \mathcal{L}_{\beta}^{\omega_a[a, q]}] = e^\beta \mu - e^\beta \mu_q \), which does not depend on \( a \), and converges to 0 as \( q \to \infty \) (notice that this is true even for \( d \geq 2 \)).

To check (4.5), set \( \mathcal{A}_q := \{(t_i, x_i, u_i)_{i=1}^m : \forall i \in [1, k], u_i \leq q \} \). By Lemma 3.3 we have

\[
\mathbb{E}\left[\mathcal{L}_{\beta}^{\omega_a[a, q]} \right] = \mathbb{E}\left[ \sum_{\sigma_1, \sigma_2 \in \mathcal{P}(\omega)} w_{a, \beta}(\sigma_1) w_{a, \beta}(\sigma_2) \mathbf{1}_{\mathcal{A}_q}(\sigma_1) \mathbf{1}_{\mathcal{A}_q}(\sigma_2) \right].
\]

In order to facilitate the use of Mecke’s multivariate equation (Proposition 3.8) we set \( \varsigma = \sigma_1 \cap \sigma_2 \) and \( \varsigma_i = \sigma_i \setminus \varsigma \) for \( i = 1, 2 \).

By removing the constraint that \( u \leq q \) on \( \varsigma_i \), \( i = 1, 2 \), we obtain

(4.9) \[
\mathbb{E}\left[\mathcal{L}_{\beta}^{\omega_a[a, q]} \right] \leq \mathbb{E}\left[ \sum_{\varsigma_1, \varsigma_2, \varsigma \in \mathcal{P}(\omega) \text{ disjoints}} w(\varsigma_1 \cup \varsigma) w(\varsigma_2 \cup \varsigma) \mathbf{1}_{\mathcal{A}_q}(\varsigma) \right].
\]

Now we can apply Proposition 3.8. To do so, we split the sum according to the cardinality of \( \varsigma = \{(t_i, x_i, u_i)_{i=1}^m \} \), and also according to the number of points in \( \varsigma_1 \) and \( \varsigma_2 \) in each of the intervals \( (t_{i-1}, t_i), i \in [1, m+1] \) \( (t_0 = 0, t_{m+1} = 1) \). After factorizing we obtain that the r.h.s. in (4.9) is equal to

(4.10) \[
\sum_{m \geq 0} \int_{0 < t_1 < \cdots < t_m < 1} \int_{(\mathbb{R}^d)^m} \beta^m \prod_{i=1}^m u_i^2 z_{\beta, a}(t_{i-1}, x_{i-1}, (t_i, x_i))^2 \left( \int_{\mathbb{R}^d} z_{\beta, a}(t_m, x_m, (1, x)) dx \right) \prod_{i=1}^m dt_i dx_i \lambda(du_i),
\]

where \( z_{\beta, a}(t, x, (t', x')) \) is the expected value of the point-to-point partition function for the polymer in the environment \( \omega^{(a)} \). With the convention \( s_0 = t, s_{t+1} = t' \) and \( y_0 = x, y_{t+1} = x' \), it is given by

(4.11) \[
 z_{\beta, a}(t, x, (t', x')) := e^{-\beta \kappa_a (t' - t)} \left[ \rho_{t'-t}(x' - x) + \sum_{l=1}^{t} \int_{t < s_1 < \cdots < s_l < t'} \prod_{j=1}^{l} \rho_{s_j - s_{j-1}}(y_j - y_{j-1}) \prod_{j=1}^{l} dx_j dy_j v_j \lambda(dv_j) \right].
\]

To see that (4.10) holds, observe that expanding all the products we obtain a sum over the indices \( m \) (standing for the number of points in \( \varsigma \) and \( \varsigma_1, \varsigma_2 \) (we need...
two indices to expand $z^2_{\beta,a}$ which stands for the number of points of $z_1$ and $z_2$ in the time interval $(t_{i-1}, t_i)$ (the term $\rho_{t' \rightarrow t}(x' - x)$ in (4.11) corresponding to $\ell = 0$). The expression of $z_{\beta,a}(t, (t', x'))$ simplifies after integration over all intermediate variables

$$z_{\beta,a}(t, (t', x')) = e^{-\beta_s(t' - t) \rho_{t' \rightarrow t}(x' - x)} \sum_{t = 0}^{\infty} \frac{[\beta_s + \mu](t' - t)}{t!} = e^{\beta_s(t' - t) \rho_{t' \rightarrow t}(x' - x)}.$$ 

Reinjecting this into (4.10) and performing the integral over $u \in (0, q)$ instead of $[a, q]$ — this yields an upper bound which is uniform in $a$ —, we obtain that

$$E_\beta \left[ \left( \mathcal{Z}^{(a), q}_{\beta,q} \right)^2 \right] \leq \sum_{m \geq 0} \left( 2^{1/2} \beta e^{2\beta \mu} V_q \right)^m \int_{\mathbb{R}^m} \int_{\mathbb{R}^m} \frac{dt_i}{\sqrt{t_i - t_{i-1}}} \int_{\mathbb{R}^m} \int_{\mathbb{R}^m} \frac{dx_i}{\sqrt{x_i - x_{i-1}}}.$$

To conclude, notice that (with the convention $l_{m+1} = 1$)

$$a_m = \int_{\mathbb{R}^m} \prod_{i=1}^{m} \frac{dt_i}{\sqrt{t_i - t_{i-1}}} \leq a_m' := \int_{\mathbb{R}^m} \prod_{i=1}^{m+1} \frac{dt_i}{\sqrt{t_i - t_{i-1}}} = \frac{\Gamma((m+1)/2)}{\Gamma((m+1)/2)}.$$

where the last identity is a standard calculation, see e.g. [9, Lem.A.3]. Hence $(a_m)_{m \geq 0}$ decays super-exponentially, so the r.h.s. of (4.14) is finite for every value of $q$. □

**Remark 4.8.** — In the case of the point-to-point partition function $\mathcal{Z}^{(q)}_{\beta}(t, x)$, one uses $X^{(q)} = \mathcal{Z}^{\omega, [a,b]}_{\beta}(t, x)$ instead of $\mathcal{Z}^{\omega, [a,b]}_{\beta}$ and the representation (3.9) instead of (3.8) to compute the first and second moment. The proof is carried out in an identical manner as above, replacing $\rho(t, x)$ with $\rho(t, x) \rho_{t' \rightarrow t}(x - x_k)$ whose integral on $(\mathbb{R}^d)^k$ is $\rho(t, x)$. The main difference is in (4.10) where the variable $x$ in $z_{\beta,a}(l_{m+1}, x_m, (1, x))$ is no longer integrated, which leads to having an extra $\rho_{t' \rightarrow t}(x - x_m)^2$ in the last integral in (4.12). An easy induction on $m$ yields that

$$\int_{(\mathbb{R}^d)^m} \rho(t, x)^2 \rho_{t' \rightarrow t}(x - x_m)^2 dx = \frac{e^{-\|x\|^2/t}}{\pi^{(m+2)/2} \prod_{i=1}^{m+1} \sqrt{t_i - t_{i-1}}}.$$

This leads us to having $a_m'$ instead of $a_m$ in the series (4.14) (with an extra $t^m$ if $t \neq 1$) and does not change the conclusion.

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4.4. Proof of Proposition 4.1 in dimension $d \geq 2$. — The proof of the previous section cannot apply to higher dimension. From a purely technical point of view the reason for this is that the r.h.s. of (4.13) is not integrable in $t$ when $d \geq 2$, making the r.h.s. of (4.14) infinite.

To circumvent this problem we need to refine our selection of trajectories. As the divergence in (4.13) comes from small values of $t$, we want to add a restriction that forbids favorable sites to have an abnormal high concentration in a small time frame. Our selection scheme presents some formal analogy and was inspired by the multibody techniques used in [8, 34] in the very different context of disordered pinning models at marginality.

Fine tuning our parameters, under the assumption (2.10) we find a restriction of the trajectories based on this idea which allows to apply Proposition 4.6.

Remark 4.9. — An alternative proof of the uniform integrability of $Z^{\omega,a}_{\beta}$ for $d \geq 2$ was brought to our attention by C. Chong. Once (4.4) has been proved, it is sufficient to show that

$$\sup_{a \in (0,1]} E \left[ (Z^{\omega,a,q}_{\beta})^p \right] < \infty,$$

for some $p > 1$ (the conclusion of Proposition 4.6 remains valid, using Hölder’s inequality instead of Cauchy–Schwarz). This last bound can be extracted from [45, Th.1.3.1] after observing that $Z^{\omega,a,q}_{\beta}$ is the solution at time 1 and coordinate 0 of the stochastic heat equation with initial condition $u_0 \equiv 1$. To extend this argument to the point-to-point partition function, some more care is required since in this case one has to consider the solution of the SHE with Dirac initial condition (not treated in [45]) but the argument should in principle also work.

However, our argument presents a few advantages. Firstly, it does not rely on any tool of stochastic integration and only marginally on the properties of the heat-kernel: it is therefore easily adaptable to the context of other disordered systems presented in Section 2.5. Also, our proof of Theorem A in [9] relies on a similar strategy and we believe that the proof in the continuum setup (which is much simpler than that in the discrete one) could be of use for potential readers of [9].

4.4.1. A finer restriction of the set of trajectories. — Let us now consider the restriction of the partition function to “good trajectories” $\sigma$. Thanks to assumption (2.10), we can fix some $p \in (1,1+2/d)$ with $p < 2$ such that $\int_{[0,1]} v^p \lambda(dv) < \infty$. We then fix for the rest of this section a parameter $\gamma > 0$ which satisfies

$$\frac{d-2}{2(p-1)} < \gamma < \frac{1}{p-1} \quad \text{(i.e., $\gamma(p-1) < 1$ and $\frac{d}{2} - \gamma(2-p) < 1$).}$$

The assumption $p \in (1,1+2/d)$ entails that $\gamma = d/2$ is always a valid choice, but we prefer to write the two separate conditions we have on $\gamma$ to make the requirements more transparent. Then, for any $q \geq 1$, we define $R_q$ as

$$(4.16) \quad R_q := \left\{ \sigma \in \mathcal{P}(\omega) : \forall \sigma' \subset \sigma, \prod_{j=1}^{n(\sigma)} u_j < q^{n(\sigma)} \prod_{j=1}^{n(\sigma')} (t_{j} - (t_{j-1} - 1)^{\gamma}) \right\}.$$
where $t'$ and $u'$ is used to denote the coordinates of time ordered points in $\sigma'$ (with $t'_0 = 0$ by convention). We set in this section.

$$\tilde{\nu}_{\beta,q}^{\omega,a} := \sum_{\sigma \in \mathcal{R}_q} w_{a,\beta}(\sigma) 1_{\mathcal{R}_q}(\sigma).$$

Note that $\sigma \in \mathcal{R}_q$ implies in particular that $u_i \leq q$ for $i \in [1,|\sigma|]$ and hence $\tilde{\nu}_{\beta,q}^{\omega,a} \leq \tilde{\nu}_{\beta}^{\omega,[0,q]}$. Now, Theorem 2.1 is a consequence of Propositions 4.10 and 4.11 below (which allow to control respectively the first and second moment of $\tilde{\nu}_{\beta,q}^{\omega,a}$) and of Proposition 4.6.

**Proposition 4.10.** — Assuming that $\gamma < 1/(p-1)$ we have

$$\lim_{q \to \infty} \sup_{a \in (0,1]} E \left[ |\tilde{\nu}_{\beta,q}^{\omega,a} - \tilde{\nu}_{\beta}^{\omega,a} | \right] = 0.$$

**Proposition 4.11.** — Assuming that $\gamma > (d-2)/(2(p-1))$, for every $q \geq 1$ we have

$$\sup_{a \in (0,1]} E \left( \tilde{\nu}_{\beta,q}^{\omega,a} \right)^2 < \infty.$$

The proof of Propositions 4.10 and 4.11 are technically more involved than that of Proposition 4.7. This is in particular because the restriction $\mathcal{R}_q$ produces an integral that does not factorize as well as the one obtained when only $\mathcal{A}_q$ is considered. We first need to introduce some technical bounds on some type of multivariate integrals which appear in our first and second moment computations respectively.

**4.4.2. Technical preliminaries: an upper bound on multivariate integrals.** — The following upper bounds are the key ingredients in the proof of Proposition 4.10 and 4.11 as they allow to control the multivariate integrals produced by the application of Mecke’s multivariate equation (Proposition 3.8).

**Lemma 4.12.** — Assume that $\int_{(0,1)} v^p \lambda(dv) < \infty$ for some $p \in (1,2)$ and also that $\int_{[1,\infty)} v \lambda(dv) < \infty$. Then for any $q \geq 1$ there is a constant $c_q$, satisfying $\lim_{q \to \infty} c_q = 0$, such that for every $m \geq 1$ and every $h \in (0,1)$, we have

$$\int_{(0,q)^m} 1_{\{ \prod_{i=1}^m u_i \geq h \}q^m} \prod_{i=1}^m u_i \lambda(du_i) \leq \frac{(2^p c_q)^m h^{1-p}}{\prod_{i=1}^m (m-1)!} \left[ \log(2^m/h) \right]^{m-1}.$$

In particular, for any $0 < \epsilon < 1$, we have

$$\int_{(0,q)^m} 1_{\{ \prod_{i=1}^m u_i \geq h \}q^m} \prod_{i=1}^m u_i \lambda(du_i) \leq \frac{(2^p c_q)^m h^{1-p-\epsilon}}{\prod_{i=1}^m (m-1)!} h^{1-p-\epsilon}.$$

**Lemma 4.13.** — Assume that $\int_{(0,1)} v^p \lambda(dv) < \infty$ for some $p \in (1,2)$. Then for any $q \geq 1$ there is a constant $C_q$ such that for every $m \geq 1$ and every $h \in (0,1)$, we have

$$\int_{(0,q)^m} 1_{\{ \prod_{i=1}^m u_i \leq h \}q^m} \prod_{i=1}^m u_i^2 \lambda(du_i) \leq h^{2-p}(C_q)^m \sum_{\ell=1}^m \frac{(2^p-\ell)^{m-\ell}}{(m-\ell)!} \left[ \log(1/h) \right]^{m-\ell}.$$
In particular, for any $0 < \varepsilon < 2 - p$, we have
\begin{equation}
\int_{(0,q)^m} 1_{\{\prod_{i=1}^m u_i \leq q^m\}} \prod_{i=1}^m u_i^2 \lambda(du_i) \leq \frac{(C_q/\varepsilon)^m}{2 - p - \varepsilon} h^{2-p-\varepsilon}.
\end{equation}

For both lemmas, the idea is to compare the integrals with some integrals with respect to the Lebesgue measure: we postpone to the Appendix the proof of the two following claims (presented as Propositions C.1 and C.2 respectively).

**Claim 1.** — Under the assumptions of Lemma 4.12, there exists a constant $c_q$ satisfying $\lim_{q \to \infty} c_q q^{1-p} = 0$ such that
\begin{equation}
\int_{(0,q)^m} 1_{\{\prod_{i=1}^m u_i \leq q^m\}} \prod_{i=1}^m u_i \lambda(du_i) \leq (c_q)^m \int_{(0,q)^m} 1_{\{\prod_{i=1}^m u_i \leq q^m\}} \prod_{i=1}^m u_i^{-p} du_i.
\end{equation}

**Claim 2.** — Under the assumptions of Lemma 4.13, there exists a constant $C_q$ such that
\begin{equation}
\int_{(0,q)^m} 1_{\{\prod_{i=1}^m u_i \leq q^m\}} \prod_{i=1}^m u_i^2 \lambda(du_i) \leq (C_q)^m \int_{(0,q)^m} 1_{\{\prod_{i=1}^m u_i \leq q^m\}} \prod_{i=1}^m u_i^{-p} du_i.
\end{equation}

**Proof of Lemma 4.12.** — Thanks to (4.21), we only have to prove that for any $h \in (0,1)$
\begin{equation}
\int_{(0,q)^m} 1_{\{\prod_{i=1}^m u_i \leq q^m\}} \prod_{i=1}^m u_i^{-p} du_i \leq \frac{h^{1-p} q^{(1-p)m}}{p - 1} \frac{(\log(2m/h))^{m-1}}{(m-1)!}
\end{equation}
and then set $c_q := c_q q^{1-p}$ (which satisfies $\lim_{q \to \infty} c_q = 0$). First of all, notice that by a change of variable $v = u/2q$, it is sufficient to prove (4.23) only in the case $q = 1/2$. We set, for all $h \in (0,1)$,
\begin{equation}
p_k(h) := (p - 1)h^{p-1} \int_{(0,1)^k} 1_{\{\prod_{i=1}^k u_i \geq h\}} \prod_{i=1}^k u_i^{-p} du_i,
\end{equation}
so to obtain (4.23) with $q = 1/2$ we need to show that for $k \geq 1$
\begin{equation}
p_k(h) \leq \frac{1}{(k-1)!} (\log(1/h))^{k-1} \quad \forall h \in (0,1)
\end{equation}
and apply it to $2^{-m}h$ instead of $h$. By a direct calculation, we have $p_1(h) = 1 - h^{p-1} \leq 1$ for all $h \in (0,1)$, which gives the result for $k = 1$. Then we can proceed by induction. Integrating with respect to the value of $u_1$ and using the change of variable $v = \log(u_1/h)$, we obtain
\begin{equation}
p_{k+1}(h) = h^{p-1} \int_h^1 u_1^{-p} (h/u_1)^{1-p} p_k(h/u_1) du_1 = \int_0^{\log(1/h)} p_k(e^{-v}) dv.
\end{equation}
From this and $p_1(h) \leq 1$ we easily obtain (4.24) by induction.

To obtain (4.18) from (4.17), we just use that $e^{ct} \geq \frac{1}{(m-1)!} (ct)^m$ for any $t > 0$ and any $m \geq 1$, by a Taylor expansion: applying this to $t = \log(2^m/h)$, we get the bound (4.18). \hfill \Box
Proof of Lemma 4.13. — Similarly to Lemma 4.12, thanks to (4.22) we only have to prove

\[
\int_{(0,q)^m} 1 \{ \prod_{j=1}^m u_{i,j} \leq h \} \prod_{i=1}^m u_i^{1-p} du_i \leq h^{2-p} q^{(2-p)m} \sum_{\ell=1}^m (2-p)^{-\ell} \log(1/h)^{m-\ell}
\]

and then set \( C_q \coloneqq q^{2-p} \). Again, by a change of variable it is sufficient to prove (4.25) only in the case \( q = 1 \). We set, for all \( h \in (0,1) \),

\[
\tilde{p}_k(h) := h^{p-2} \int_{(0,1)^k} 1 \{ \prod_{j=1}^k u_{j} \leq h \} \prod_{i=1}^k u_i^{1-p} du_i,
\]

and a direct calculation gives that \( \tilde{p}_1(h) = (2-p)^{-1} \), which yields the result for \( m = 1 \). For the induction step, decomposing according to whether \( u_1 \leq h \) or \( u_1 > h \), we get that

\[
\tilde{p}_{k+1}(h) = h^{p-2} \left[ \int_0^h u_1^{1-p} du_1 \left( \int_0^1 v^{1-p} dv \right) + \int_h^1 u_1^{1-p} (h/u_1)^{2-p} \tilde{p}_k(h/u_1) du_1 \right]
\]

\[
= \left( \frac{1}{2-p} \right)^{k+1} + \int_0^{\log(1/h)} \tilde{p}_k(e^{-v}) dv,
\]

where we also used a change of variable \( v = \log(u_1/h) \) for the last identity. From this we easily obtain by induction that

\[
\tilde{p}_k(h) \leq \sum_{\ell=1}^k \frac{(2-p)^{-\ell}}{(k-\ell)!} \log(1/h)^{k-\ell},
\]

which is the desired result. Now, to obtain (4.20) from (4.19), we use the inequality \( e^{xt} \geq (xt)^m/(m-\ell)! \) with \( t = \log(1/h) \) to get that for any \( 0 < \varepsilon < 2 - p \)

\[
\sum_{\ell=1}^m \frac{(2-p)^{-\ell}}{(m-\ell)!} \log(1/h)^{m-\ell} \leq \sum_{\ell=1}^m \left( \frac{\varepsilon}{2-p} \right)^{\ell} \varepsilon^{-m} h^{-\varepsilon} \leq \frac{\varepsilon}{2-p - \varepsilon} \varepsilon^{-m} h^{-\varepsilon}.
\]

This concludes the proof of (4.20).

\[\Box\]

4.4.3. Proof of Proposition 4.10. — Note that as we have already proved (4.4) and since \( \mathcal{Z}_q^\omega, [a,q] \geq \mathcal{Z}_q^\omega, [a,q] \), it is sufficient to prove that

\[
\lim_{q \to \infty} \sup_{q \in (0,1]} \mathbb{E} \left[ \mathcal{Z}_q^\omega, [a,q] - \mathcal{Z}_q^\omega, [a,q] \right] = 0.
\]

Decomposing over the cardinality of \( \sigma \), using Proposition 3.8 as in (4.6) and integrating over the space variable (recall (4.7)) we obtain

\[
\mathbb{E} \left[ \mathcal{Z}_q^\omega, [a,q] - \mathcal{Z}_q^\omega, [a,q] \right] = e^{-\beta n} \sum_{k=0}^k \beta^k \int_{X^k \times [a,q]^k} 1_{\mathcal{Y}_q^\omega}(t,u) \prod_{i=1}^k u_i \lambda(du_i) dt_i,
\]

where we recall that \( \mathcal{Y}_q^\omega \) has been defined in (4.16). Here, with some abuse of notation, we identified \( \mathcal{Y}_q^\omega \) and its image by the projection \((t,x,u) \mapsto (t,u)\); note that \( \mathcal{Y}_q \) does
not involve the space variable. To estimate the above integral, we use a union bound for \(1_{(\mathscr{G}_t,p)}(t,u)\). When the value of \(k\) is fixed we have

\[
1_{(\mathscr{G}_t,p)}(t,u) \leq \sum_{m=1}^{k} \sum_{1 \leq i_1 < \cdots < i_m \leq k} 1_{\{\prod_{i=1}^{m} u_i > q^m \prod_{i=1}^{m} (t_i - t_{i-1})^\gamma\}}.
\]

With this done, we can perform first the integral with respect to \(u_j\) and \(t_j\) for \(j \notin \{i_1, \ldots, i_m\}\): summing over the number of points \(k_i\) that can be fitted between two points \(i \in (t_{i+1}, t_{i-1})\), we obtain after factorization that the r.h.s. in (4.26) is smaller than

\[
\sum_{k_1=0}^{\infty} \beta^{k_1} \int_{\mathbb{R}^m} \prod_{i=1}^{k_1} u_i^\lambda(du_i) dt_i,
\]

where we used the convention \(t_0 = 0\) and \(t_{m+1} = 1\). Now we can compute explicitly each term of the product in the second line above (as in (4.8)). Replacing \(q\) by \(\infty\) in the domain of integration, which yields an upper bound and makes the computation simpler, we get, recalling that \(\int_{[0,\infty)} u^\lambda(du) = \kappa_a + \mu\),

\[
\sum_{k_1=0}^{\infty} \beta^{k_1} \int_{\mathbb{R}^m} \prod_{i=1}^{k_1} u_i^\lambda(du_i) dt_i = e^{\beta(\kappa_a + \mu)}(t_{i-1} - t_i).
\]

The product of these terms gives a factor \(e^{\beta(\kappa_a + \mu)}\) and we therefore get the inequality

\[
(4.27) \quad \mathbb{E} \left[ \mathcal{X}_t^\omega_{\beta,a} - \mathcal{X}_t^\omega_{\beta,a} \right] \leq e^{\beta \mu} \sum_{m \geq 1} \int_{\mathbb{R}^m} \prod_{i=1}^{m} u_i^\lambda(du_i) dt_i,
\]

The r.h.s. in (4.27) can be bounded above using Lemma 4.12. More specifically we use (4.18) to bound the integral over \(u_i\), setting \(a\) to 0 to obtain an upper bound that does not depend on \(a\). We fix \(\epsilon\) small enough so that \(\gamma(p + \epsilon - 1) < 1\) (recall that we have \(\gamma(p - 1) < 1\) by assumption) and by (4.18) we obtain

\[
(4.28) \quad \int_{[0,\infty)^m} \prod_{i=1}^{m} u_i^\lambda(du_i) \leq \left(\frac{c_q(\epsilon)^m}{p - 1}\right) \prod_{i=1}^{m} (t_i - t_{i-1})^{\gamma(1-p-\epsilon)}.
\]

We therefore obtain that (4.27) is smaller than

\[
(4.29) \quad \frac{e^{\beta \mu}}{p - 1} \sum_{m \geq 1} (\beta c_q/\epsilon)^m \int_{\mathbb{R}^m} \prod_{i=1}^{m} (t_i - t_{i-1})^{\gamma(1-p-\epsilon)} dt_i
\]

\[
= \frac{e^{\beta \mu}}{p - 1} \sum_{m \geq 1} (\beta c_q/\epsilon)^m \frac{\Gamma(1 - \gamma(p + \epsilon - 1))^m}{\Gamma(m(1 - \gamma(p + \epsilon - 1)) + 1)}.
\]
where in the last equality we used that \( \gamma(p + \varepsilon - 1) < 1 \). The sum in the r.h.s. of (4.29) is finite for any value of \( q \geq 1 \), and can be made arbitrarily small by choosing \( q \) large (with \( \varepsilon \) and \( \gamma \) fixed), since Lemma 4.12 ensures that the constant \( c_q \) goes to 0 as \( q \to \infty \).

4.4.4. Proof of Proposition 4.11. — We have

\[
E \left[ (\mathcal{F}^\omega, a) \right]^2 = E \left[ \sum_{\sigma_1, \sigma_2 \in \mathcal{B}_q(\omega)} w_{a, \beta}(\sigma_1)w_{a, \beta}(\sigma_2)1_{\mathcal{B}_q}(\sigma_1)1_{\mathcal{B}_q}(\sigma_2) \right].
\]

We use again the notation \( \varsigma = \sigma_1 \cap \sigma_2 \) and \( \varsigma = \sigma_1 \setminus \varsigma \). Let us relax the condition \( \sigma_1, \sigma_2 \in \mathcal{B}_q \) to obtain something which is easier to handle in the computation. Formally the divergence of the second moment of the (unrestricted) partition is obtained when integrating the contribution of the environment at the points in the replica intersection \( \sigma_1 \cap \sigma_2 \), hence we should not be losing too much if we restrict our constraint to \( \varsigma \). With this in mind, we set

\[
\mathcal{B}_q = \left\{ \varsigma = (t_i, x_i, u_i)_{i=1}^{|\varsigma|} : \max_{1 \leq i \leq |\varsigma|} u_i < q, \prod_{i=1}^{|\varsigma|} u_i \leq q^{|\varsigma|} \prod_{i=1}^{|\varsigma|} (t_i - t_{i-1})^\gamma \right\},
\]

and observe that

\[
E \left[ (\mathcal{F}^\omega, a)^2 \right] \leq E \left[ \sum_{\varsigma_1, \varsigma_2 \leq \mathcal{B}_q(\omega) \text{ disjoints}} w_{a, \beta}(\varsigma_1 \cup \varsigma)w_{a, \beta}(\varsigma_2 \cup \varsigma)1_{\mathcal{B}_q}(\varsigma) \right].
\]

Using again Proposition 3.8 as in (4.10) and using (4.11) to integrate over all the variables associated with \( \xi_1 \) and \( \xi_2 \) we obtain

\[
E \left[ \sum_{\varsigma_1, \varsigma_2 \leq \mathcal{B}_q(\omega) \text{ disjoints}} w_{a, \beta}(\varsigma_1 \cup \varsigma)w_{a, \beta}(\varsigma_2 \cup \varsigma)1_{\mathcal{B}_q}(\varsigma) \right] = \sum_{m \geq 0} \left( \beta \varepsilon 2^p \mu \right)^m \int_{\mathbb{R}^m} \left| \prod_{i=1}^m u_i \right| \frac{m}{2 \pi} \lambda(du_i) dt_i dx_i.
\]

Now integrating over \( x \) and using (4.13), we obtain that the r.h.s. above is equal to

\[
\sum_{m \geq 0} \left( \beta \varepsilon 2^p \mu \right)^m \int_{\mathbb{R}^m} \left| \prod_{i=1}^m u_i \right| \lambda(du_i) dt_i \prod_{i=1}^m (t_i - t_{i-1})^\gamma u_i^2 \lambda(du_i).
\]

To estimate the integral over \( u_1, \ldots, u_m \), we use (4.20) in Lemma 4.13. We integrate over \( (0, q) \) to get an upper bound which is uniform in \( \alpha \). We fix \( \varepsilon \) such that \( \frac{d}{2} - (2 - p - \varepsilon)\gamma < 1 \) (recall that \( \frac{d}{2} - (2 - p)\gamma < 1 \) by assumption) and by (4.20) we obtain

\[
\int_{[0,q]^m} \left| \prod_{i=1}^m u_i \right| \frac{m}{2 \pi} \lambda(du_i) \leq \frac{(C_q/\varepsilon)^m}{2 - p - \varepsilon} \prod_{i=1}^m (t_i - t_{i-1})^{\gamma(2-p-\varepsilon)}.
\]
Reinjected in (4.31), this yields

\[ E\left[ \left( \frac{X_{\omega,a}}{Z_{\beta,q}} \right)^2 \right] \leq \sum_{m \geq 0} \frac{\left( 2 - d - d/2 - \beta \varepsilon \right)^{m}}{2 - p - \varepsilon} \int_{\mathbb{X}^{m}} \prod_{i=1}^{m} \frac{dt}{(t_i - t_{i-1})^{\frac{d}{2}}(2 - p - \varepsilon)^2 - (2 - p - \varepsilon)\gamma}. \]

To conclude we just need to show that the above sum is finite. To check this, we simply observe that, thanks to the fact that \( \frac{d}{2} - \gamma (2 - p - \varepsilon) < 1 \), the integral in \( t \) is equal to

\[ \frac{\Gamma(1 - (2 - p - \varepsilon)\gamma + d/2)^m}{\Gamma(m[1 - (2 - p - \varepsilon)\gamma + d/2] + 1)}, \]

and that the corresponding series in \( m \) has an infinite radius of convergence. \( \square \)

**Remark 4.14.** — For the proof of Propositions 4.10–4.11 in the case of the point-to-point partition function \( Z_{\omega,a}(t,x) \), we need to slightly change the definition of \( B_q \) to take care of the end point, setting

\[ B_q := \left\{ \sigma \in \mathcal{P}(\omega) : \exists \sigma' \subset \sigma, \prod_{j=1}^{\sigma'} u_j' < q^{\sigma'} \prod_{j=1}^{\sigma'} (t_j' - t_{j-1}'), \right\}, \]

with the convention that \( t_{\sigma'}^{\sigma'} = t \). For the proof of Proposition 4.10, there is an additional term \( \rho_t(x) \) in (4.26), coming from the integration of \( g(t,x) \) over the space variable, but the main difference comes in (4.28) when applying Lemma 4.12.

The computation in (4.29) is different (we have the integral of \( \prod_{i=1}^{m+1} (t_i - t_{i-1})^{\gamma(1-p)} \) over \( \mathbb{X}^{m} \) after scaling by \( t \) if \( t \neq 1 \)) but the conclusion is identical. For the proof of Proposition 4.11, there is an additional term \( \rho_{t-m}(x-x_m) \) in (4.30). We proceed as in (4.15) when integrating on \( x_1, \ldots, x_m \) and this yields only an extra multiplicative term \( C e^{-\varepsilon t^2}/(t-m)^{-d/2} \). Then, the integral in (4.32) is different (we have the integral of \( \prod_{i=1}^{m+1} (t_i - t_{i-1})^{-\frac{d}{2}} + (p-2+\varepsilon) \) over \( \mathbb{X}^{m} \)) but this does not change the conclusion.

### 4.5. Finiteness of partition functions

We are going to prove here simultaneously Proposition 2.5, the first part of Proposition 2.20, (2.18) and Proposition 2.24. Note that the fact that \( Z_{\omega,a} \) and other partition functions are positive is a direct consequence of the rewriting given in (3.8). It remains to prove that under assumption (2.9) they are almost surely finite. This is the following statement.

**Proposition 4.15.** — If (2.9) is satisfied, then for every \( u_0 \) satisfying (2.25) (with \( T = 1 \)) we have for any \( t \in [0,1] \), almost surely

\[ \int_{\mathbb{R}^d} Z_{\omega,a}(t,x) u_0(dx) < \infty. \]

In particular the cases \( u_0 = \delta_{\varepsilon} \) and \( u_0(dx) = dx \) respectively give

\[ Z_{\omega,a}(t,x) < \infty \quad \text{and} \quad Z_{\omega,a} < \infty. \]

**Remark 4.16.** — Proposition 2.5 and (2.18) are direct consequences of (4.33). Proposition 2.24 also follows by observing that by time reversal and translation invariance we have the following identity in distribution

\[ \int_{\mathbb{R}^d} Z_{\omega,a}(t,y)(t,x) u_0(dy) = \int_{\mathbb{R}^d} Z_{\omega,a}(t,y-x) u_0(dy) \]

and thus we just need to apply the result to \( u_0 \) translated by \( x \).
Proof of Proposition 4.15. — For \( \sigma = (t_i, x_i, u_i)_{i=1}^{|\sigma|} \in \mathcal{P}(\omega) \) (with \( t_k \leq 1 \)), we define
\[
H(\sigma) := \sum_{i=1}^{|\sigma|} \frac{\|x_i - x_{i-1}\|^2}{t_i - t_{i-1}} \quad \text{and} \quad G(\sigma) := \sum_{i=1}^{|\sigma|} \log(u_i).
\]
The quantity \( H(\sigma) \) and \( G(\sigma) \) corresponds roughly to the cost and gains at the exponential level to visit all the points in \( \sigma \). We refer to \( H(\sigma) \) as the entropy of the path.

Our statement is an almost direct consequence of the following lemma.

Lemma 4.17. — If (2.9) holds, for any fixed \( \varepsilon > 0 \), we have almost surely
\[
\mathcal{F}(\omega) := \sup_{\sigma \in \mathcal{P}(\omega)} \left\{ G(\sigma) - \frac{\varepsilon}{2} H(\sigma) \right\} < \infty.
\]

We postpone the proof of this lemma and first deduce the proposition from it. Using the representation (3.8) of the partition function (recall that \( \omega^{(a)} \) denote the set of points in the environment with jump size larger than \( a \)) and applying Lemma 4.17, we can write for \( t \leq 1 \)
\[
e^{\beta a} \mathcal{Z}_\omega^{(a)}(t, x) = \sum_{\sigma \in \mathcal{P}_{[0,1]}(\omega^{(a)})} e^{\beta |\sigma|} e^{G(\sigma)} \rho(t, \sigma_0) \prod_{i=1}^{|\sigma|} e^{-\frac{\varepsilon}{2} |\sigma|} \rho_{t-t_k} (x - x_k)
\]
so that setting \( \vartheta = \vartheta(\varepsilon) := (1 - \varepsilon)^{-1} \) and assuming that \( \varepsilon < 1/2 \)
\[
e^{-\vartheta} \mathcal{Z}_\omega^{(a)}(t, x) \leq e^{-\beta a} \sum_{\sigma \in \mathcal{P}_{[0,1]}(\omega^{(a)})} (2^{d/2} \beta)^{|\sigma|} \rho(t, \sigma_0) \vartheta^{d/2} \rho_{t-t_k} (x - x_k).
\]

By Mecke’s formulæ (Proposition 3.8) we conclude that
\[
(4.34) \quad \mathbb{E} \left[ e^{-\mathcal{F}(\omega)} \int_{\mathbb{R}^d} \mathcal{Z}_\omega^{(a)}(t, x) u_0(\,dx) \right] \leq 2^{d/2} e^{\beta \mathbb{E}[2^{d/2} \lambda([a, \infty)) - \kappa_a]} \int_{\mathbb{R}^d} \rho_{\vartheta t}(x) u_0(\,dx).
\]

Using assumption (2.25) on \( u_0 \) and fixing \( \varepsilon \) sufficiently small (i.e., \( \vartheta \) close enough to 1) so that \( \rho_{\vartheta t}(x) \) is integrable w.r.t. \( [u_0] \) (recall \( t \leq T = 1 \), we get that (4.34) is finite. This proves that \( \int_{\mathbb{R}^d} \mathcal{Z}_\omega^{(a)}(t, x) u_0(\,dx) < \infty \) almost surely thanks to Lemma 4.17. \( \square \)

Proof of Lemma 4.17. — First of all, notice that if \( \sigma = (t_i, x_i, u_i)_{i=1}^{\omega} \) has a point with \( u_i < 1 \), then by removing this point from \( \sigma \) we obtain a set \( \sigma' \) with (strictly) smaller entropy \( H(\sigma') < H(\sigma) \) and (strictly) higher energy \( G(\sigma') > G(\sigma) \). In the supremum, we therefore can restrict ourselves to points \( (t, x, v) \in \omega \) with \( v \geq 1 \). Let us now separate points according to the size of their jump. For each \( k \geq 1 \) we define
\[
\omega_k := \{(t, x, v) \in \omega : t \in [0, 1] \text{ and } \log v \in [e^{k-1}, e^k)\},
\]
and we let \(\pi(\omega_k)\) be its projection on the first two coordinates. Note that the \(\pi(\omega_k)\)'s are independent Poisson processes on \([0, 1] \times \mathbb{R}^d\) with respective intensity \(\lambda_k dt \otimes dx\), where

\[
\lambda_k := \lambda\left(\exp(e^{k-1}), \exp(e^k)\right).
\]

One can then easily see that our assumption (2.9) is equivalent to having

\[
(4.35) \quad \sum_{k \geq 1} \lambda_k e^{dk/2} < \infty.
\]

Our proof is based on the following statement, proved below.

Lemma 4.18. — Fix \(\theta \in (1, 1 + 1/d)\) and let \(K > 0\) be arbitrary. Let \(A_{k, n}\) be the event that there exists a path of \(n\) points in \(\omega_k\) whose entropy is smaller than \(Kn^\theta e^k\), i.e.,

\[
A_{k, n} = \bigcup_{\sigma \in \mathcal{P}(\omega_k), |\sigma| = n} \{H(\sigma) \leq Kn^\theta e^k\}.
\]

Then, assuming (2.9), we have \(\sum_{k, n} \mathbb{P}(A_{k, n}) < \infty\).

Given \(\sigma\), we let \(n_k(\sigma)\) denote the number of points that the path displays in \(\omega_k\). We have

\[
G(\sigma) = \sum_{k = 1}^\infty e^k n_k(\sigma).
\]

We let \(k_0(\omega), n_0(\omega)\) be such that \(A_{k_0, n_0}\) holds for every \(k \geq k_0\) (for every \(n\)), and for every \(n \geq n_0\) (for every \(k\)). For every \(k\), we have

\[
n_k(\sigma) \leq \frac{K^{-1}H(\sigma)e^{-k}}{\theta} + n_0 \mathbf{1}_{(k \leq k_0)},
\]

where we have used that \(H(\sigma) \geq H(\sigma_k)\), with \(\sigma_k \in \mathcal{P}(\omega_k)\) the set obtained by removing all points in \(\sigma\) which are not in \(\omega_k\). In particular, we have \(n_k(\sigma) = 0\) for \(k \geq 1 + \max(k_0, \log(K^{-1}H(\sigma)))\). This yields

\[
G(\sigma) \leq \sum_{k = 1}^{k_0} e^k n_0 + \left(\frac{K^{-1}H(\sigma)}{\theta}\right) \sum_{k = 0}^{\log(K^{-1}H(\sigma))} e^{k(\theta - 1)/\theta}.
\]

Since \(K\) is arbitrary, fixing \(K \geq (2/\varepsilon)(1 - e^{(1-\theta)/\theta})^{-1}\) yields that \(G(\sigma) \leq C(\omega) + (\varepsilon/2)H(\sigma)\) almost surely, which concludes the proof.

Proof of Lemma 4.18. — Let us start with the case \(n = 1\). Using that \(H(\sigma) \geq \|x\|^2\) if \(\sigma = (t, x, v)\) is reduced to one point (recall \(t \leq 1\), we get that

\[
A_{k, 1} \subset \bigcup_{(t, x, v) \in \omega_k} \{\|x\|^2 \leq Ke^k\};
\]

in other words, if \(A_{k, 1}\) is satisfied then there is a point in \(\omega_k\) within a distance \(\sqrt{Ke^k/2}\) from the origin. The probability of \(A_{k, 1}\) is therefore smaller than a constant times \(K^{d/2}e^{kd/2}\lambda_k\) and this is summable over \(k\) thanks to (4.35). The case \(n = 2\) can be treated similarly.
When $n \geq 3$, let $\sigma = (t_i, x_i, u_i)_{i=1}^n$ be a path of points in $\omega_k$ satisfying our event, i.e., $H(\sigma) \leq Kn^\theta e^k$. We make the following claims.

**Claim 1.** The path cannot venture too far:

\begin{equation}
\max_{i \in [1,n]} \|x_i\| \leq \sqrt{Kn^{\theta/2}e^{k/2}}.
\end{equation}

**Claim 2.** There are three consecutive points in our path in a relatively small cylinder: there exists $i \in [1, n-2]$ satisfying

\begin{equation}
t_{i+2} - t_i \leq \frac{4}{n-2} \quad \text{and} \quad \|x_{i+1} - x_i\|^2 + \|x_{i+2} - x_{i-1}\|^2 \leq \frac{16Kn^\theta e^k}{(n-2)^2}.
\end{equation}

Before proving the two claims (4.36)–(4.37), let us use them to conclude the proof of Lemma 4.18. We can cover $[0, 1] \times [-\sqrt{Kn^{\theta/2}e^{k/2}}, \sqrt{Kn^{\theta/2}e^{k/2}}]d$ with a collection of $C_Kn^{1+d}$ overlapping cylinders of the type $[t, t + \frac{8}{n-2}] \times [x, x + \frac{\theta}{n-2}Kn^{\theta/2}e^{k/2}]d$ for a collection of $t$'s distant by $\frac{4}{n-2}$ and of $x$'s distant by $\frac{4}{n-2}Kn^{\theta/2}e^{k/2})$. Thanks to the above claims, if $A_{k,n}$ is satisfied then there exists some path $\sigma \in \mathcal{P}(\omega)$ of length $n$ satisfying (4.36)–(4.37), meaning that at least one of the constructed cylinders contains three points in $\pi(\omega_k)$. By a union bound, since each cylinder is of area bounded by $C_Kn^{(d/2)-(d+1)}e^{kd/2}$, we therefore get that

$$\mathbb{P}(A_{k,n}) \leq C_Kn^{1+d} \times \left(C_Kn^{(d/2)-(d+1)}e^{kd/2}\lambda_k\right)^3 \leq C_{n}^\theta (\lambda_k e^{kd/2})^3 n^{(3d/2)\theta-(2d+1)}.$$

We conclude simply by observing that this upper bound is summable over $k$ and $n$, thanks to (4.35) and since $(3d/2)\theta - 2(d+1) < -\frac{1}{2}(d+1)$ thanks to our choice $\theta < 1 + 1/d$.

The first claim (4.36) just follows by observing that the entropy of the path is larger than $\|x_i\|^2/t_i \geq 2$ for every $i \in [1, n]$. For the second claim (4.37), we observe that we have $\sum_{i=1}^{n-2} (t_{i+2} - t_i) \leq 2$, which means that setting

$$I := \{i \in [1, n-2], t_{i+2} - t_i \leq 4/(n-2)\},$$

we have $|I| \geq \frac{n-2}{2}$. We also have, by definition of $I$,

$$\sum_{i \in I} \|x_{i+1} - x_i\|^2 + \|x_{i+2} - x_{i+1}\|^2 \leq \frac{4}{n-2} \sum_{i \in I} \frac{\|x_{i+1} - x_i\|^2}{t_{i+1} - t_i} + \frac{\|x_{i+2} - x_{i+1}\|^2}{t_{i+2} - t_{i+1}} \leq \frac{8H(\sigma)}{n-2}.$$

Then (4.37) is simply a consequence of the fact that the smallest element of the sum is smaller than the average and that $H(\sigma) \leq Kn^\theta e^k$.

**4.6. Proof of Lemma 4.3.** We now adapt Proposition 4.1 to prove the convergence of $Z^{a, n}_{\omega}(f)$ and $Z^{a, n}_{\omega}$ without the condition $\int_{[1, \infty)} \lambda(dt) < \infty$, i.e., we prove Lemma 4.3. Recall the definition (2.26) and set for $f \in \mathcal{B}$

\begin{equation}
Z^{a, n}_{\omega}(f) := 1 + \sum_{k=1}^{\infty} \beta^k \int_{X^d \times \mathbb{R}^d} \varphi(t, x, f) \prod_{i=1}^{k} \mathcal{E}_{j}^{a, n}(dt_i, dx_i).
\end{equation}
Note that Proposition 4.1 (or rather Corollary 4.2), applied to the measure $\lambda_b$ defined by
\begin{equation}
\lambda_b(A) = \lambda(A \cap [0,b))
\end{equation}
automatically yields the following convergence of $Z_{\omega,a}^{(1)}(f)$.

**Corollary 4.19.** — Under the assumption (2.10), for any $f \in \mathcal{B}$ and any fixed $b \geq 1$ we have that $(Z_{\omega,a}^{(1)}(f))_{a \in (0,1]}$ is a uniformly integrable time-reversed martingale. Hence the following convergence therefore holds almost surely and in $L_1$:
\begin{equation*}
\lim_{a \to 0} Z_{\omega,a}^{(1)}(f) =: Z_{\omega}^{(1)}(f).
\end{equation*}

Note also that we have
\begin{equation*}
E[Z_{\omega,a}^{(1)}(f)] = e^{\beta \mu_b} Q(f) \text{ for all } a \in (0,1], \text{ with } \mu_b := \int_{(1,b)} \nu \lambda(d\nu).
\end{equation*}

We now prove the convergence of $Z_{\omega,a}^{(2)}(f)$ for $f \in \mathcal{B}$. Repeating the argument from the proof of Proposition 3.1, there exists $b_0(\omega,f)$ such that for $b \geq b_0$, we have $Z_{\omega,a}^{(2)}(f) = Z_{\omega}^{(2)}(f)$ for every $a \in (0,1]$. Thus we have, from Corollary 4.19,
\begin{equation*}
\lim_{a \to 0} Z_{\omega,a}^{(2)}(f) = Z_{\omega}^{(2)}(f).
\end{equation*}

It only remains to prove that $\lim_{a \to 0} Z_{\omega,a}^{(2)}(f)$ exists and is finite for $f \in \mathcal{B}$ when (2.9) also holds. We focus on the case $f \equiv 1$ for simplicity but the argument adapts immediately to the case of non-negative bounded $f$, so there is no loss of generality. The convergence of $Z_{\omega,a}^{(2)}$ is a consequence of the following statement, valid for any $\epsilon > 0$,
\begin{equation}
\lim_{b \to \infty} P\left(\sup_{a \in (0,1]} (Z_{\omega,a}^{(2)} - Z_{\omega}^{(2)}_{\beta} - Z_{\omega}^{(2)}_{\beta}) > \epsilon\right) = 0.
\end{equation}

Indeed, by monotonicity in $b$ (cf. Lemma 3.3) we have for every $b \in (1,\infty)$,
\begin{equation*}
\lim_{a \to 0} Z_{\omega,a}^{(2)} \geq Z_{\omega}^{(2)}_{\beta}.
\end{equation*}

Thus to ensure that $Z_{\omega,a}^{(2)}$ converges to $Z_{\omega}^{(2)}_{\beta}$ and that the latter is finite, the only thing that needs be proved is that for every $\epsilon > 0$, there almost surely exists $b_0(\omega,\epsilon)$ such that
\begin{equation}
\limsup_{a \to 0} Z_{\omega,a}^{(2)} \leq Z_{\omega}^{(2)}_{\beta} + \epsilon.
\end{equation}

Since $\sup_{a \in (0,1]} (Z_{\omega,a}^{(2)} - Z_{\omega}^{(2)}_{\beta})$ is non-increasing in $b$, (4.40) implies that
\begin{equation}
\lim_{b \to \infty} \sup_{a \in (0,1]} (Z_{\omega,a}^{(2)} - Z_{\omega}^{(2)}_{\beta}) = 0 \quad \mathbb{P}-\text{a.s.},
\end{equation}
and this readily implies that (4.41) holds for some large $b_0$.

To prove (4.40), we observe that the process \{\epsilon \wedge (Z_{\omega,a}^{(2)}_{\beta} - Z_{\omega}^{(2)}_{\beta} - Z_{\omega,a}^{(2)}_{\beta}) : a \in (0,1]\} is a time-reversed positive (for $b' > b > 1$) super-martingale, thanks to Lemma 3.5.
Using Doob’s maximal inequality we thus obtain that for any fixed \( b' > b > 1 \)
\[
\mathbb{P}
\left( \sup_{a \in (0,1)} \left( \mathbb{F}_{\beta}^\omega \{ \mathbb{F}_{\beta}^{a,b'} - \mathbb{F}_{\beta}^{a,b} \} > \varepsilon \right) \right) \leq \frac{1}{\varepsilon} \mathbb{E}
\left[ \left( \mathbb{F}_{\beta}^{1,b'} - \mathbb{F}_{\beta}^{1,b} \right) \wedge \varepsilon \right].
\]
Sending \( b' \) to infinity on both sides and noting that \( \mathbb{F}_{\beta}^{a,b} = \lim_{b' \to \infty} \mathbb{F}_{\beta}^{a,b'} \) by monotone convergence (see Lemma 3.3), we obtain
\[
\mathbb{P}
\left( \sup_{a \in (0,1)} \left( \mathbb{F}_{\beta}^{a,b} - \mathbb{F}_{\beta}^{a,b} \right) > \varepsilon \right) \leq \frac{1}{\varepsilon} \mathbb{E}
\left[ \left( \mathbb{F}_{\beta}^{1,b} - \mathbb{F}_{\beta}^{1,b} \right) \wedge \varepsilon \right].
\]
The right-hand side goes to zero by dominated convergence, thanks to Proposition 2.5, using again that \( \mathbb{F}_{\beta}^{1,b} = \lim_{b' \to \infty} \mathbb{F}_{\beta}^{1,b'} \). \( \square \)

4.7. Almost sure positivity of \( \mathbb{F}_{\beta}^{\omega}(f) \). — In this section, we prove Proposition 4.4. Let \( f \in \mathcal{B}_b \) be non-negative and such that \( \mathbb{Q}(f) > 0 \). Recall the definitions (2.26) and (4.38) of the truncated noise and of the corresponding partition function. We are going to show first that the positivity of \( \mathbb{F}_{\beta}^{\omega,\{0,b\}}(f) \) does not depend on the value of \( b \), for any \( 0 < b < b' \leq 1 \).
\[
\mathbb{P}
\left( \{ \mathbb{F}_{\beta}^{\omega,\{0,b\}}(f) > 0 \} \triangle \{ \mathbb{F}_{\beta}^{\omega,\{0,b'\}}(f) > 0 \} \right) = 0.
\]
where \( \triangle \) stands for the symmetric difference (in other words the events are equal in the \( L_1 \) sense and in particular have the same probability). Applying Lemma 3.3 to the measure \( \lambda_b \) (recall (4.39)) we obtain that
\[
\mathbb{F}_{\beta}^{\omega,\{a,b\}}(f) = e^{-\beta (\kappa_a - \kappa_b)} \sum_{\sigma \in \mathcal{P}(\omega)} \beta^{1|\sigma|} \theta(t, x, f) \prod_{i=1}^{1|\sigma|} u_i 1\{u_i \in [a,b]\}.
\]
This last expression implies that for every \( a < b' \leq b \) we have almost surely
\[
\mathbb{F}_{\beta}^{\omega,\{a,b'\}}(f) \geq e^{-\beta \kappa_b} \mathbb{F}_{\beta}^{\omega,\{a,b\}}(f) \geq e^{-\beta \kappa_b} \mathbb{F}_{\beta}^{\omega,\{a,b'\}}(f),
\]
and taking the limit when \( a \) tends to zero we obtain
\[
\mathbb{F}_{\beta}^{\omega,\{0,b'\}}(f) \geq e^{-\beta \kappa_b} \mathbb{F}_{\beta}^{\omega,\{0,b\}}(f) \geq e^{-\beta \kappa_b} \mathbb{F}_{\beta}^{\omega,\{0,b'\}}(f).
\]
This yields
\[
\mathbb{P}
\left( \{ \mathbb{F}_{\beta}^{\omega,\{0,b'\}}(f) > 0 \} \triangle \{ \mathbb{F}_{\beta}^{\omega,\{0,b\}}(f) > 0 \} \right) = 0.
\]
On the other hand, the same argument as in Lemma 3.5 yields that \( \{ \mathbb{F}_{\beta}^{\omega,\{a,b\}}(f) \}_{a \in (0,1]} \) is a martingale (in \( b \)) for the filtration \( \mathcal{G}_b := \sigma\{(t, x, v) \in \omega : v < b\} \).
Taking \( a \) to zero in the conditional expectation (using uniform integrability cf. Corollary 4.19) we obtain that for \( 1 \geq b' > b \)
\[
\mathbb{E}
\left[ \mathbb{F}_{\beta}^{\omega,\{0,b\}}(f) \big| \mathcal{G}_b \right] = \mathbb{F}_{\beta}^{\omega,\{0,b'\}}(f),
\]
which yields the second inclusion of (4.44)
\[
\mathbb{P}
\left( \{ \mathbb{F}_{\beta}^{\omega,\{0,b\}}(f) > 0 \} \triangle \{ \mathbb{F}_{\beta}^{\omega,\{0,b'\}}(f) > 0 \} \right) = 0.
\]
Now let us fix a decreasing sequence \(b_n \in (0,1]\) with \(b_n \downarrow 0\) and consider the event
\[
A(f) := \bigcap_{m \geq 0} \bigcup_{n \geq m} \{2^\omega_{\beta,[0,b_n]}(f) > 0\}.
\]
An immediate consequence of (4.44) is that
\[
\mathbb{P}(2^\omega_{\beta,[0,1]}(f) > 0) = \mathbb{P}[A(f)].
\]
Now \(A(f)\) is measurable with respect to the \(\sigma\)-algebra \(\mathcal{H}_0 := \bigcap_{n \geq 1} \mathcal{H}_n\). Therefore, by Kolmogorov’s \(0\)–\(1\) law, it has probability either \(0\) or \(1\); indeed, for any \(n \geq 1\), \(\mathcal{H}_n\) is independent of \(\mathcal{H}_0 = \sigma(\{(t,x,v) \in \omega: v \geq b_n\})\): since \(A \in \mathcal{H}_0\) and \(A \in \sigma(\bigcup_{n \geq 0} \mathcal{H}_n)\), we get that \(A\) is independent of itself. From Corollary 4.19 the martingale \((2^\omega_{\beta,[n,1]}(a))_{a \in (0,1]}\) is uniformly integrable and thus
\[
\mathbb{E}[2^\omega_{\beta,[0,1]}(f)] = \mathbb{Q}(f) > 0.
\]
Combining these two facts we obtain that necessarily \(\mathbb{P}(2^\omega_{\beta,[0,1]}(f) > 0) = 1\), which concludes the proof using (4.45).

**Remark 4.20.** — The proofs in Section 4.6–4.7 apply verbatim to the point-to-point partition function. They give that for any \((t,x) \in \mathbb{R}_+^* \times \mathbb{R}^d\) we have the a.s. convergence \(\lim_{a \to \infty} 2^\omega_{\beta,[a,1]}(t,x) = 2^\omega_{\beta}(t,x)\), with \(2^\omega_{\beta}(t,x)\) positive and finite almost surely.

### 4.8. Proof of Proposition 4.5.

We now show that \((\mathbb{Q}_{\beta,a}^{\omega,a})_{a \in (0,1]}\) is tight. We need to find a sequence of compact sets \(\mathcal{K}_N\) such that almost surely
\[
\lim_{N \to \infty} \sup_{a \in (0,1]} \mathbb{Q}_{\beta,a}^{\omega,a}(\mathcal{K}_N) = 0.
\]
Since \(2^\omega_{\beta,a}\) converges to a positive limit, this is of course equivalent to proving
\[
\lim_{N \to \infty} \sup_{a \in (0,1]} 2^\omega_{\beta,a}(1_{\mathcal{K}_N}) = 0.
\]
In the case \(\mu := \int_{[1,\infty)} v \lambda(\mathrm{d}v) < \infty\), since \(\mathbb{E}[2^\omega_{\beta,a}(1_A)] = e^{\beta \mu} \mathbb{Q}(A)\), using Doob’s maximal inequality we have
\[
\mathbb{P}\left(\sup_{a \in (0,1]} 2^\omega_{\beta,a}(1_{\mathcal{K}_N}) \geq \varepsilon\right) \leq \frac{1}{\varepsilon} e^{\beta \mu} \mathbb{Q}(\mathcal{K}_N).
\]
It is then easy to show that (4.46) holds for an arbitrary increasing sequence of compacts satisfying \(\lim_{N \to \infty} \mathbb{Q}(\mathcal{K}_N) = 1\) (using the monotonicity of \(f \mapsto 2^\omega_{\beta,a}(f)\), cf. Lemma 3.3). For instance one can take
\[
\mathcal{K}_N := \{\varphi \in C_0([0,1]) : |\varphi(t) - \varphi(s)| \leq N|t-s|^{1/4} \forall s,t \in [0,1]\}.
\]
In the case \(\int_{[1,\infty)} v \lambda(\mathrm{d}v) = \infty\), we proceed analogously with the truncated partition function \(2^\omega_{\beta,[a,b]}\). We obtain that for any \(b > 0\),
\[
\lim_{N \to \infty} \sup_{a \in (0,1]} 2^\omega_{\beta,[a,b]}(1_{\mathcal{K}_N}) = 0.
\]
We then conclude using (4.42), to get that a.s.
\[
\lim_{b \to \infty} \sup_{a \in (0,1]} \left(2^\omega_{\beta,a}(1_{\mathcal{K}_N}) - 2^\omega_{\beta,[a,b]}(1_{\mathcal{K}_N})\right) \leq \lim_{b \to \infty} \sup_{a \in (0,1]} \left(2^\omega_{\beta,a} - 2^\omega_{\beta,[a,b]}\right) = 0.
\]
5. Degeneracy of the partition function: Propositions 2.6 and 2.12

5.1. Proof of Proposition 2.6. — Let us assume that \( \int_{[1, \infty)} (\log v)^{d/2} \lambda(dv) = \infty \) and show that the partition function \( Z^\omega_\beta \) is a.s. infinite. We use the representation \( (3.8) \). Keeping only paths \( \sigma \) with cardinality one in the sum and keeping only those with \( t \in [1/2, 1] \) we have

\[
Z^\omega_\beta \geq e^{-\beta \kappa_a} \sum_{(t, x, \nu) \in \omega^\alpha} \rho_t(x) v \geq e^{-\beta \kappa_a} \sum_{(t, x, \nu) \in \omega^\alpha, t \geq 1/2} v e^{-\|x\|^2}.
\]

Hence to conclude it is sufficient to show that almost surely

\[
\sup_{(t, x, \nu) \in \omega, t \in [1/2, 1]} \{ \log v - \|x\|^2 \} = \infty.
\]

For this it is sufficient to check that almost surely, the event \( A_j \) defined by

\[
A_j := \{ \exists (t, x, \nu) \in \omega : t \in [1/2, 1], \|x\|_\infty \in [2^{j-1}, 2^j), \log v \geq 4^{j+1} \},
\]

is satisfied for infinitely many \( j \). By Borel–Cantelli, since the \( A_j \) are by construction independent it suffices to show that \( \sum_{j=1}^{\infty} P(A_j) = \infty \). The number of points \( (t, x, \nu) \in \omega \) such that \( t \in [1/2, 1], \|x\|_\infty \in [2^{j-1}, 2^j) \) and \( \log v \geq 4^{j+1} \) is a Poisson random variable with mean

\[
\lambda_j = \frac{1}{2} 2^{dj} (1 - 2^{-d}) \lambda(\exp(4^{j+1}), \infty)).
\]

Hence we have \( P(A_j) = 1 - e^{-\lambda_j} \) and we simply need to show that \( \sum_{j=1}^{\infty} \lambda_j = \infty \). But this is a direct consequence of our assumption \( \int_{[1, \infty)} (\log v)^{d/2} \lambda(dv) = \infty \) since

\[
\int_{[1, \infty)} (\log v)^{d/2} \lambda(dv) \leq \sum_{j=1}^{\infty} \int_{\exp(4^{j+1})}^{\exp(4^{j+2})} (4^{j+2})^{d/2} \lambda(dv) \leq \frac{2}{1 - 2^{-d}} 4^{d} \sum_{j=1}^{\infty} \lambda_j.
\]

5.2. Proof of Proposition 2.12. — Since the use of the size-biased measure is at the heart of our proof we are going first to assume that \( \mu := \int_{[1, \infty)} v \lambda(dv) < \infty \), in order to be able to use Lemma 3.7. At the end of the proof we explain how to deal with the case \( \mu = \infty \).

Note that \( Z^\omega_\beta \) converges almost surely, as a consequence of the martingale property: we only need to prove that it converges to zero in probability. Since \( \mathbb{E} Z^\omega_\beta := e^{-\beta \mu} Z^\omega_\beta \) is a positive variable with mean \( 1 \) it is sufficient to identify a sequence of events \( J_a \) such that

\[
\lim_{a \to 0} \mathbb{E} \left[ \mathbb{E}^a Z^\omega_\beta \mathbf{1}_{J_a} \right] = 0 \quad \text{and} \quad \lim_{a \to 0} P(J_a) = 1,
\]

as it implies that \( \mathbb{E}^a Z^\omega_\beta \mathbf{1}_{J_a} \) and thus \( Z^\omega_\beta \) converge to zero in probability. This is equivalent to proving that the total variation distance between the two measures \( \mathbb{P} \) and \( \mathbb{P}^a \) goes to 1. That is, according to Lemma 3.7, we need to prove that

\[
\lim_{a \to 0} \left\| \mathbb{P} (\omega \in \cdot) - \mathbb{Q} \otimes \mathbb{E} \otimes \mathbb{E}^a (\varpi \in \cdot) \right\|_{TV} = 1.
\]
Our proof’s strategy relies on finding a statistic that helps to distinguish between $\omega$ and $\tilde{\omega}$ for most realizations of the Brownian trajectory. More precisely, we use the second moment method. We are going to define a functional $Y_a(\omega)$ which satisfies

$$\lim_{a \to 0} \frac{\mathbb{E}(Q \otimes E \otimes E'_a | Y_a(\tilde{\omega}) - \mathbb{E}[Y_a(\omega)])^2}{\text{Var}_\mathbb{P}(Y_a(\omega)) + \text{Var}_Q \otimes \mathbb{E}_e(Y_a(\omega))} = \infty.$$  

(5.2)

The above implies that asymptotically $Y_a(\tilde{\omega})$ and $Y_a(\omega)$ concentrate around different values and thus that (5.1) holds (see [42, Prop. 7.12] for a quantitative statement and its proof). We treat separately the cases $d \geq 3$, $d = 1$ and $d = 2$, in that order.

5.2.1. The case $d \geq 3$. — We assume that $\int_{(0,1)} v^{1+2/d} \lambda(du) = +\infty$. In order to find a statistic that allows us to distinguish between $\mathbb{P}$ and $\tilde{\mathbb{P}}_\beta$, the idea here is to find a region of $\mathbb{R} \times \mathbb{R}^d \times \mathbb{R}_+$ where $\tilde{\mathbb{P}}_\beta$ displays significantly more points than $\mathbb{P}$. We consider a sequence $R_a$ going (slowly) to infinity (we set its value later on) and we set

$$Y_a(\omega) := \#\{ (t, x, u) \in \omega : \|x\|_\infty \leq R_a, u \geq (a \vee t^{d/2}) \}.$$ 

Under $\mathbb{P}$, $Y_a$ is a Poisson variable with mean given by

$$\mathbb{E}[Y_a(\omega)] := (2R_a)^d \int_0^1 t^{d/2} \lambda([a \vee t^{d/2}, \infty)) dt.$$ 

Note that our assumption on $\lambda$ readily implies that $\int_0^1 t^{d/2} \lambda([a \vee t^{d/2}, \infty)) dt$ and hence $\mathbb{E}[Y_a(\omega)]$ go to infinity as $a \downarrow 0$. On the other hand, conditionally on $(B_t)_{t \in [0,1]}$, under $\mathbb{P} \otimes \tilde{\mathbb{P}}'_a$, we have that $Y_a(\tilde{\omega})$ is a Poisson random variable with mean

$$\mathbb{E} \otimes \tilde{\mathbb{P}}'_a [Y_a(\tilde{\omega})] = (2R_a)^d \int_0^1 t^{d/2} \lambda([a \vee t^{d/2}, \infty)) dt + \beta \mathcal{F}_a,$$

(5.3)

where we have set

$$\mathcal{F}_a := \int_0^1 \mathbf{1}_{\{|B_t| \leq R_a \sqrt{t}\}} \int_{(a \vee t^{d/2}, \infty)} v \lambda(du) dt.$$ 

Before averaging with respect to the Brownian motion, since $R_a$ tends to infinity, notice that we can almost replace $\mathbf{1}_{\{|B_t| \leq R_a \sqrt{t}\}}$ by 1, so $\mathcal{F}_a$ is close to $m_a := \int_0^1 \int_{(a \vee t^{d/2}, \infty)} v \lambda(du) dt$. With this in mind (and the fact that the variance of a Poisson variable is equal to its expectation), the important part that has to be checked for (5.2) to hold is that

$$\lim_{a \to 0} \frac{m_a^2}{(R_a)^d \int_0^1 t^{d/2} \lambda([a \vee t^{d/2}, \infty)) dt} = \infty.$$  

(5.4)

Since

$$m_a = \int_0^1 \int_{(a \vee t^{d/2}, \infty)} v \lambda(du) dt \geq \int_0^1 t^{d/2} \lambda([a \vee t^{d/2}, \infty)) dt$$

and

$$\int_0^1 t^{d/2} \lambda([t^{d/2}, \infty)) dt = +\infty.$$
by assumption, the condition (5.4) is satisfied as long as \( R_a \) diverges slowly enough. We can choose for instance
\[
R_a = \left( \int_0^1 t^{d/2} \lambda([a \lor t^{d/2}, \infty)) dt \right)^{1/2d}.
\]
Now that all the notation have been set, let us complete the proof of (5.2). Setting \( q_a := Q(|B_1| \leq R_a) \), we have by that Brownian scaling that
\[
Q(\mathcal{F}_a) = q_a \int_0^1 \int_{[a \lor t^{d/2}, \infty)} v \lambda(dv) dt = q_a m_a,
\]
\[
\text{Var}_Q(\mathcal{F}_a) = Q(\mathcal{F}_a^2) - Q(\mathcal{F}_a)^2 \leq (1 - q_a^2) m_a^2.
\]
As a consequence, recalling (5.3), we have that
\[
Q \otimes E \otimes E'_{\omega} [Y_a(\omega)] - E[Y_a(\omega)] = \beta q_a m_a.
\]
We also have
\[
\text{Var}_{Q \otimes P \otimes P'_{\omega}} [Y_a(\omega)] = Q(\text{Var}_{P \otimes P'_\omega}(Y_a(\omega))) + \beta^2 \text{Var}_Q(\mathcal{F}_a)
\]
\[
\leq (2 R_a)^d \int_0^1 \int_{[0 \lor t^{d/2}, \infty)} \lambda([a \lor t^{d/2}, \infty)) dt + \beta q_a m_a + \beta^2 (1 - q_a^2) m_a^2.
\]
Now we can conclude that (5.2) holds, simply by using (5.4) and the fact that \( q_a \) tends to 1 as \( a \searrow 0 \) (using also that \( m_a = o(m_2) \) since \( m_a \) goes to \( \infty \)).

5.2.2. The case \( d = 1 \). We assume that \( \int_{(0,1)} v^2 \lambda(dv) = \infty \). In this case we set
\[
Y_a(\omega) := \sum_{(t,x,v) \notin \omega} v \mathbf{1}_{\{v \in [a,1], t \in [0,1], \|x\|_\omega \leq R_a\}},
\]
where again \( R_a \) is a sequence going to infinity sufficiently slowly (it is chosen below). Then, we have
\[
E[Y_a(\omega)] = 2 R_a \int_{[a,1]} v \lambda(dv), \quad \text{Var}(Y_a(\omega)) = 2 R_a \int_{[a,1]} v^2 \lambda(dv).
\]
Additionally, setting this time \( q_a := \int_0^1 Q(|B_t| \leq R_a) dt \) we have
\[
Q \otimes E \otimes E'_{\omega} [Y_a(\omega)] = 2 R_a \int_{[a,1]} v \lambda(dv) + \beta q_a \int_{[a,1]} v^2 \lambda(dv),
\]
and using a variant of the argument used in (5.5)
\[
\text{Var}_{Q \otimes P \otimes P'_{\omega}} [Y_a(\omega)]
\]
\[
\leq 2 R_a \int_{[a,1]} v^2 \lambda(dv) + \beta q_a \int_{[a,1]} v^3 \lambda(dv) + \beta^2 (1 - q_a^2) \left( \int_{[a,1]} v^3 \lambda(dv) \right)^2
\]
\[
\leq (2 R_a + \beta q_a) \int_{[a,1]} v^2 \lambda(dv) + \beta^2 (1 - q_a^2) \left( \int_{[a,1]} v^2 \lambda(dv) \right)^2,
\]
where we simply used that \( v \leq 1 \) for the second inequality. Now, since \( q_a \) goes to 1 as \( a \searrow 0 \), to conclude that (5.2) holds it is sufficient to have \( R_a = o \left( \int_{[a,1]} v^2 \lambda(dv) \right) \) which can be obtained by setting \( R_a = \left( \int_{[a,1]} v^2 \lambda(dv) \right)^{1/2} \).
5.2.3. The case \(d = 2\). — We assume that \(\int_{(0,1)} v^2 |\log v| \lambda(\mathrm{d}v) = \infty\). In this case we define

\[
Y_a(\omega) := \sum_{(t,x,v) \in \omega} \frac{v}{t \vee v} 1_{\{v \in [a,1), t \in [0,1], \|x\|_\infty \leq R_a, t\}}
\]

where \(R_a\) goes to infinity slowly enough (it is chosen below). In that case, we have

\[
\mathbb{E}[Y_a(\omega)] = (2R_a)^2 \int_{[a,1]} \int_0^1 \frac{v^2t}{t \vee v} \mathrm{d}t \lambda(\mathrm{d}v)
\]

\[
\text{Var}(Y_a(\omega)) = (2R_a)^2 \int_{[a,1]} \int_0^1 \frac{v'^2t}{(t \vee v)^2} \mathrm{d}t \lambda(\mathrm{d}v).
\]

Now with \(q_a := \mathbb{Q}\{|B_1| \leq R_a\}\) we have,

\[
\mathbb{Q} \otimes \mathbb{E} \otimes \mathbb{E}_a[Y_a(\tilde{\omega})] = (2R_a)^2 \int_{[a,1]} \int_0^1 \frac{v^2t}{t \vee v} \mathrm{d}t \lambda(\mathrm{d}v) + q_a \int_{[a,1]} \int_0^1 \frac{v'^2t}{t \vee v} \mathrm{d}t \lambda(\mathrm{d}v).
\]

Using again a variant of (5.5) to bound the variance from above we obtain

\[
\text{Var}_{\mathbb{Q} \otimes \mathbb{P} \otimes \mathbb{P}_a}[Y_a(\tilde{\omega})] \leq (2R_a)^2 \int_{[a,1]} \int_0^1 \frac{v'^3t}{(t \vee v)^2} \mathrm{d}t \lambda(\mathrm{d}v)
\]

\[
+ q_a \beta \int_{[a,1]} \int_0^1 \frac{v'^3t}{(t \vee v)^2} \mathrm{d}t \lambda(\mathrm{d}v) + \beta^2(1 - q_a)^2 \left( \int_{[a,1]} \int_0^1 \frac{v'^2t}{t \vee v} \mathrm{d}t \lambda(\mathrm{d}v) \right)^2.
\]

To conclude we need to check that (5.2) holds. It is not difficult to show that the second and third term appearing in the variance of \(Y_a(\tilde{\omega})\) can be neglected (recall that \(q_a\) goes to 1) and hence to conclude one only needs to ensure that

\[
\lim_{a \to 0} \frac{\left( \int_{[a,1]} \int_0^1 \frac{v'^2t}{t \vee v} \mathrm{d}t \lambda(\mathrm{d}v) \right)^2}{(2R_a)^2 \int_{[a,1]} \int_0^1 \frac{v'^2t}{(t \vee v)^2} \mathrm{d}t \lambda(\mathrm{d}v)} = \infty.
\]

Now this can be done by setting \(R_a = (\int_{[a,1]} v^2 |\log v| \mathrm{d}v)^{1/4}\) since both integrals in the numerator and the denominator are comparable to \(\int_{[a,1]} v^2 |\log v| \mathrm{d}v\).

5.2.4. Conclusion of the proof of Proposition 2.12. — When \(\int_{[1,\infty)} v^2 \lambda(v) \mathrm{d}v < \infty\), we have shown that if (2.14) is not satisfied, then \(\lim_{a \to 0} \mathcal{F}_B^{\omega,a}_\beta = 0\) almost surely. If \(f \in \mathcal{R}\), simply using that \(|\mathcal{F}_\alpha^{\omega,a}_\beta(f)| \leq \|f\|_\infty \mathcal{F}_\alpha^{\omega,a}_{\beta} \) gives us \(\lim_{a \to 0} \mathcal{F}_\alpha^{\omega,a}_\beta(f) = 0\).

Let us now turn to the case \(\int_{[1,\infty)} v \lambda(v) \mathrm{d}v = \infty\). For \(f \in \mathcal{R}_b\) we can replace the noise \(\xi^{(a)}\) by a truncated one \(\xi^{(a,b)}\) (recall (2.26)), like in the proof of Proposition 3.1, using that \(f\) has a bounded support. We therefore conclude that we also have \(\lim_{a \to 0} \mathcal{F}_\alpha^{\omega,a}_\beta(f) = 0\).

It remains to show that if (2.9) holds, then we also have \(\lim_{a \to 0} \mathcal{F}_\alpha^{\omega,a}_\beta = 0\) a.s. We set

\[
f_n(B) := 1_{\{\max_{x \in [0,1]} |B_1| \leq n\}} \quad \text{and} \quad f_n = 1 - f_n.
\]
We have for any \( n \) \( \lim_{a \to 0} \mathcal{Z}^{\omega,a}_1(f_n) = 0 \) a.s. and we can thus conclude if we prove that

\[
\lim_{n \to \infty} \sup_{a \in (0,1]} \mathcal{Z}^{\omega,a}_1(f_n) = 0
\]

Using Doob’s maximal inequality for the super-martingale \( (\mathcal{Z}^{\omega,a}_1(f_n))_{a \in (0,1]} \), we get

\[
P(\sup_{a \in (0,1]} \mathcal{Z}^{\omega,a}_1(f_n) > \varepsilon) \leq \frac{1}{\varepsilon} \mathbb{E}[\mathcal{Z}^{\omega,a}_1(f_n) \wedge \varepsilon].
\]

Taking \( b \) to infinity on both sides we get

\[
P(\sup_{a \in (0,1]} \mathcal{Z}^{\omega,a}_1(f_n) > \varepsilon) \leq \frac{1}{\varepsilon} \mathbb{E}[\mathcal{Z}^{\omega,a}_1(f_n) \wedge \varepsilon],
\]

which proves (5.6) by dominated convergence, thanks to Proposition 2.5.

6. Properties of the continuum directed polymer in Lévy noise

In this section, we prove various properties of the measure \( \mathbb{Q}^\omega_T \) constructed in Theorem 2.8. We always suppose that Assumptions (2.9)–(2.10) are satisfied.

6.1. Proof of Proposition 2.16. — The proof is based on the following technical result.

**Lemma 6.1.** — For any fixed \( f \in \mathcal{B} \) we have almost surely

\[
\mathbb{Q}^\omega_T(f) = \lim_{a \to 0} \mathbb{Q}^{\omega,a}_T(f).
\]

Let us briefly comment on the result above before we present its proof. Lemma 4.3 and Proposition 4.4 ensure that \( \mathbb{Q}^{\omega,a}_T(f) \) converges a.s. as a quotient: for any \( f \in \mathcal{B} \), we have almost surely

\[
\lim_{a \to 0} \mathbb{Q}^{\omega,a}_T(f) = \frac{\mathcal{Z}^{\omega}_T(f)}{\mathcal{Z}^{\omega}_T(f)}.
\]

What requires a proof is that for a fixed \( f \in \mathcal{B} \) we almost surely have

\[
\mathcal{Z}^{\omega}_T(f) = \mathcal{Z}^{\omega}_T(f) =: \mathcal{Z}^{\omega}_T(f).
\]

This is a non-trivial statement since the two terms correspond to different definitions: \( \mathcal{Q}^{\omega}_T(f) \) is the expectation of \( f \) with respect to the weak limit of \( \mathbb{Q}^{\omega,a}_T(f) \). The definition of weak convergence only implies that (6.3) holds when \( f \in \mathcal{C} \).

Note also here that the position of quantifiers is important. The convergence (6.1) holds with probability one, simultaneously for every \( f \in \mathcal{C} \), but not necessarily simultaneously for all bounded measurable functions. The latter statement would correspond to convergence in total variation and does not hold when \( \int_{(0,1)} v \lambda(dv) = \infty \) (see Remark 2.19).
Proof. — To prove (6.1), we only need to verify that for any fixed measurable bounded set $A$, we almost surely have

$$\lim_{a \to 0} Q_{\beta}^{\omega,a}(A) = Q_{\beta}^{\omega}(A). \tag{6.4}$$

Indeed, (6.4) implies that (6.1) is satisfied for non-negative simple functions with bounded support. Then given a non-negative $f \in \mathcal{B}$, consider a non-negative sequence of simple functions with bounded support $(f_n)_{n \geq 1}$ which is such that $f_n \uparrow f$. Using monotone convergence, we have for every $m \geq 1$

$$\frac{\mathcal{Z}_{\beta}(f)}{\mathcal{Z}_{\beta}(f)} = \lim_{a \to 0} Q_{\beta}^{\omega,a}(f) = \lim_{a \to 0} \lim_{n \to \infty} Q_{\beta}^{\omega,a}(f_n) \geq \lim_{a \to 0} Q_{\beta}^{\omega,a}(f_m) = Q_{\beta}^{\omega}(f_m).$$

Letting $m \to \infty$ and using monotone convergence again, we get $\mathcal{Z}_{\beta}(f) \geq \mathcal{Z}_{\beta}(f)$. Since we also have (for the same reason) $\mathcal{Z}_{\beta}(g) \geq \mathcal{Z}_{\beta}(g)$ for $g = (|f| - f)$, we obtain the equality (6.1).

Let us now prove (6.4). Since $Q$ is a regular measure, one can find an increasing sequence of closed set $(A_n^{(1)})_{n \geq 1}$ and a decreasing sequence of bounded open sets $(A_n^{(2)})_{n \geq 1}$ such that

$$\forall n \geq 1, A_n^{(1)} \subset A \subset A_n^{(2)}, \quad \text{and} \quad \lim_{n \to \infty} Q(A_n^{(2)} \setminus A_n^{(1)}) = 0. \tag{6.5}$$

The inclusions implies that

$$\lim_{a \to 0} Q_{\beta}^{\omega,a}(A_n^{(1)}) \leq \lim_{a \to 0} Q_{\beta}^{\omega,a}(A) \leq \lim_{a \to 0} Q_{\beta}^{\omega,a}(A_n^{(2)}).$$

Now, by the Portmanteau theorem, we have for every $n \geq 1$

$$\lim_{a \to 0} Q_{\beta}^{\omega,a}(A_n^{(1)}) \leq Q_{\beta}^{\omega}(A_n^{(1)}) \leq Q_{\beta}^{\omega}(A_n^{(2)}) \leq \lim_{a \to 0} Q_{\beta}^{\omega,a}(A_n^{(2)}).$$

Hence to conclude that $\lim_{a \to 0} Q_{\beta}^{\omega,a}(A) = Q_{\beta}^{\omega}(A)$, it is sufficient prove that

$$\lim_{n \to \infty} \lim_{n' \to \infty} Q_{\beta}^{\omega}(A_n^{(2)} \setminus A_n^{(1)}) = (2\beta)^{-1} \lim_{n \to \infty} Q_{\beta}^{\omega}(A_n^{(2)} \setminus A_n^{(1)}) = 0, \tag{6.6}$$

where the first equality is just a consequence of the definitions (see (6.2)). If we assume that $\mu := \int_{[1, \infty)} v \lambda(dv) < \infty$, then using successively Fatou’s lemma, Corollary 4.2 and (6.5), we obtain that

$$\mathbb{E} \left[ \lim_{n \to \infty} 2\beta \left(1_{A_n^{(2)} \setminus A_n^{(1)}} \right) \right] \leq \lim_{n \to \infty} \mathbb{E} \left[ 2\beta \left(1_{A_n^{(2)} \setminus A_n^{(1)}} \right) \right] = e^{\beta \mu} \lim_{n \to \infty} Q(A_n^{(2)} \setminus A_n^{(1)}) = 0,$$

proving (6.6). Now if $\int_{[1, \infty)} v \lambda(dv) = \infty$, the previous argument yields that for any $b$, $

$$\lim_{n \to \infty} 2\beta A_n^{(2)} \setminus A_n^{(1)} = 0.$$ Since $A_n^{(2)} \setminus A_n^{(1)} \subset A_n^{(2)}$ for all $n$ with $A_n^{(2)}$ bounded, we have that for $b$ sufficiently large,

$$2\beta A_n^{(2)} \setminus A_n^{(1)} = 2\beta(0,b) A_n^{(2)} \setminus A_n^{(1)}$$

for every $n$. This allows to conclude that $2\beta A_n^{(2)} \setminus A_n^{(1)}$ goes to 0 also in that case. \hfill \Box
Proof of Proposition 2.16. — Given $A$ such that $Q(A) = 0$, we have a.s. $Q^\omega_{\emptyset}(A) = 0$, for any $a \in (0, 1]$. If $\mu < \infty$, this follows from (3.3). If $\mu = \infty$, one has a.s. $\mathcal{Z}^\omega_{\emptyset}([a,b)} (1_A) = 0$, for any $b > 1$: by monotonicity $\mathcal{Z}^\omega_{\emptyset} (1_A) = \lim_{b\to\infty} \mathcal{Z}^\omega_{\emptyset}([a,b)} (1_A) = 0$. Thus Lemma 6.1 applied to $1_A$ implies that $Q^\omega_{\emptyset}(A) = 0$ for almost every $\omega$, giving that $P \times Q^\omega_{\emptyset}(A) = 0$. □

6.2. Proof of Proposition 2.18. — Let us start with item (i). $Q(\emptyset_{\text{empty}}(\omega)) = 1$, which is the easiest statement. For any $t > 0$ and $x \in \mathbb{R}^d$ we have $Q(B_t = x) = 0$, hence,

$$Q(\emptyset_{\text{empty}}) \leq \sum_{(t,x,\nu) \in \omega} Q(B_t = x) = 0.$$ 

Now, to prove item (ii), we notice that when $\kappa_0 := \int_{[0,1]} v \lambda(dv) < \infty$ then almost surely the $Q^\omega_{\emptyset}$-probability of an event $A$ is given by

$$Q^\omega_{\emptyset}(A) = \frac{1}{Z^\omega_{\emptyset}} \sum_{\sigma \in \mathcal{P}(\omega)} w_{0,\beta} (\sigma, 1_A),$$

where $w_{0,\beta} (\sigma, 1_A)$ is defined as in (3.6) with $a = 0$. Note that we have $w_{0,\beta} (\sigma, f) = \lim_{a \to 0} w_{a,\beta} (\sigma, f)$ provided that $\kappa_0 < +\infty$ (otherwise $\lim_{a \to 0} w_{a,\beta} (\sigma, f) = 0$). Notice that the sum in the r.h.s. of (6.7) is finite since, by monotone convergence (note that $e^{\beta \kappa_0} w_{a,\beta} (\sigma, 1_A)$ is non-increasing in $a$),

$$e^{\beta \kappa_0} \sum_{\sigma \in \mathcal{P}(\omega)} w_{0,\beta} (\sigma, 1_A) = \lim_{a \to 0} e^{\beta \kappa_0} \sum_{\sigma \in \mathcal{P}(\omega)} w_{a,\beta} (\sigma, 1_A) = \lim_{a \to 0} e^{\beta \kappa_0} \mathcal{Z}^\omega_{\emptyset} (1_A) = e^{\beta \kappa_0} \mathcal{Z}^\omega_{\emptyset} (1_A),$$

the last term being finite thanks to Lemma 4.3. We can then conclude that (6.7) holds using Lemma 6.1. In particular, from (6.7) we have

$$Q^\omega_{\emptyset} (\emptyset_{\text{empty}}) = \frac{e^{-\beta \kappa_0}}{Z^\omega_{\emptyset}} \in (0, 1).$$

For $\emptyset_{\infty}$ the same argument as in item (i) shows that $w_{0,\beta} (\sigma, 1_{\emptyset_{\infty}}) = 0$, since $\emptyset_{\infty}$ requires that the trajectory visits at least one point outside of $\sigma$ (recall that $\sigma$ is finite), and hence $Q^\omega_{\emptyset} (\emptyset_{\infty}) = 0$.

Let us now turn to the more delicate item (iii). Our idea is to find a sequence $A_n$ of sets in $C_0([0,1])$ which are such that

$$\lim_{n \to \infty} Q^\omega_{\emptyset} (A_n^C) = 0 \quad \text{a.s. and} \quad \lim \sup A_n := \bigcup_{n \geq 0} A_n \subset \emptyset_{\text{dense}}.$$ 

We will then get that almost surely

$$Q_{\emptyset}^\omega (\emptyset_{\text{dense}}) \geq \lim_{n \to \infty} Q_{\emptyset}^\omega (\bigcup_{m \geq n} A_n) = 1.$$ 

By (6.1) the first requirement in (6.8) is equivalent to $\lim_{n \to \infty} \lim_{a \to 0} Q_{\emptyset}^\omega \omega_{\emptyset} (A_n^C) = 0$ and thus to

$$\lim_{n \to \infty} \lim_{a \to 0} \mathcal{Z}^\omega_{\emptyset} (1_{A_n^C}) = 0,$$

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since \( X_n^\omega, a \) converges to a positive limit and is thus bounded away from 0. The obvious way to bound \( X_n^\omega, a(A_n^\beta) \) is via the computation of its expectation. For this reason we first assume that \( \mu = \int_{[1, \infty)} v \lambda (dv) < \infty \). Let us denote \( z(a, \varphi, \omega) \) the maximal spacing in the times of visit to points in \( \omega(a) \), i.e.,

\[
z(a, \varphi, \omega) := \sup \{ s : \exists t \in [0, 1 - s], \Delta(\varphi, \omega(s)) \cap [t, t + s] = \emptyset \},
\]

where we recall that \( \Delta(\varphi, \omega) = \{ t \in [0, 1] : \exists v > 0, (t, \varphi(t), v) \in \omega \} \) (see (2.16)). Let us also set

\[
a_n := \sup \{ a : \kappa_a \geq n \},
\]

which goes to 0 as \( n \) tends to infinity. To lighten notation, let us set \( \kappa := \kappa_n \geq n \) and \( \pi_n := \mu + \kappa_n \). We then define

\[
A_n := \{ \varphi : z(a_n, \varphi, \omega) \leq \kappa_n^{-1/2} \},
\]

and notice that \( \mathcal{A}_n \) is satisfied as soon as infinitely many \( A_n \)'s are satisfied. Now, from Lemma 3.5 we have that \( (X_n^\omega, a(A_n^\beta))_{a \in (0, a_n)} \) is a martingale which is uniformly integrable: we can therefore extend it at 0. By Markov's inequality we have

\[
\mathbb{P} \left( \lim_{a \to 0} X_n^\omega, a(A_n^\beta) \geq n^{-1} \right) \leq n \mathbb{E} [X_n^\omega, a(A_n^\beta)].
\]

The r.h.s. can be computed explicitly: we have (recall that \( t_0 = 0, t_{s+1+1} = 1 \))

\[
X_n^{\omega, a_n}(A_n^\beta) = \sum_{\sigma \in \mathcal{A}(\omega(a_n)), i \in (0, s], t_{s+1+1} > \kappa_n^{-1/2}} w_{a_n, \beta}(\sigma).
\]

Hence we have

\[
\mathbb{E} [X_n^{\omega, a_n}(A_n^\beta)] \leq e^{-\beta \kappa_n} \sum_{k=0}^{\infty} (\beta \pi_n)^k \int_{X^k} \mathbf{1}_{\{i \in [0, k], t_{s+1+1} > \kappa_n^{-1/2}\}} dt.
\]

Now, by symmetry (the roles of \( t_{s+1+1} - t_i \) can be exchanged), we have for any \( k \geq 1 \)

\[
\int_{X^k} \mathbf{1}_{\{i \in [0, k], t_{s+1+1} > \log n \}} dt \leq (k + 1) \int_{X^k} \mathbf{1}_{\{i \in [0, k], t_{s+1+1} > \kappa_n^{-1/2}\}} dt = \frac{(k + 1)(1 - \kappa_n^{-1/2})^k}{k!}.
\]

Thus, bounding \( k + 1 \leq 2k \) for \( k \geq 1 \), we have

\[
\mathbb{E} [X_n^{\omega, a_n}(A_n^\beta)] \leq e^{-\beta \kappa_n} (1 + 2\beta \pi_n e^{\beta \pi_n(1 - \kappa_n^{-1/2})}) \leq e^{-\beta \kappa_n} + 2 \beta \pi_n e^{\beta \pi_n e^{-\beta \pi_n}} \leq n^{-3},
\]

where the last inequality holds for \( n \) sufficiently large. We therefore get that for \( n \) sufficiently large

\[
\mathbb{P} \left( \lim_{a \to 0} X_n^\omega, a(A_n^\beta) \geq n^{-1} \right) \leq n^{-2},
\]

which is summable, so we conclude that (6.9) holds a.s. by Borel–Cantelli lemma. When \( \int_{[1, \infty)} v \lambda (dv) = \infty \), in order to prove (6.9) (for the same \( A_n \)), we observe that

\[
X_n^{\omega, a}(A_n^\beta) \leq (X_n^\omega, a(A_n^\beta) + X_n^\omega, a(b))(A_n^\beta).
\]

The second term goes to zero by the above proof and the first one can be made arbitrarily small by choosing \( b \) large, thanks to Proposition 2.5 which ensures that \( X_n^{\omega, a} \) decreases.
6.3. Proof of Proposition 2.21. — Recall that (see (2.20))

\[ \mathcal{F}_\beta^\omega([t_1, x_1], [t_2, x_2]) := \limsup_{a \to 0} \mathcal{F}_{\beta}^{\omega, a}([t_1, x_1], [t_2, x_2]). \]

We use a lim sup so that the quantity is defined everywhere (which is necessary for integration). However, as a consequence of Proposition 2.20 (proved in Section 4) and of translation invariance, for a fixed choice of end points, the lim sup can almost surely be replaced by a limit, and hence for any fixed \( 0 < t_1 < \cdots < t_k = 1 \) and \( (x_1, \ldots, x_k) \)

\[ P \left( \exists i \in [1, k], \mathcal{F}_{\beta}^\omega([t_{i-1}, x_{i-1}], (t_i, x_i)] > \liminf_{a \to 0} \mathcal{F}_{\beta}^{\omega, a}([t_{i-1}, x_{i-1}], (t_i, x_i)) \right) = 0. \]

Thus, as a consequence of Fubini’s theorem, the set

\[ \{(x_1, \ldots, x_k) : \exists i \in [1, k], \mathcal{F}_{\beta}^\omega([t_{i-1}, x_{i-1}], (t_i, x_i)] > \liminf_{a \to 0} \mathcal{F}_{\beta}^{\omega, a}([t_{i-1}, x_{i-1}], (t_i, x_i)) \} \]

has almost surely zero Lebesgue measure. Now, let \( g \) be a bounded continuous function of \( k \) variables in \( \mathbb{R}^d \), satisfying \( 0 \leq g \leq 1 \). Applying (2.13) with \( \Pi(\varphi) := g(\varphi(t_1), \ldots, \varphi(t_k)) \), we have

\[ \mathcal{Q}_\beta^\omega(g(B_{t_1}, \ldots, B_{t_k})) = \lim_{a \to 0} \mathcal{Q}_\beta^{\omega, a}(g(B_{t_1}, \ldots, B_{t_k})), \quad \text{a.s.} \]

For any \( a > 0 \) we have

\[ \mathcal{Q}_\beta^{\omega, a}(g(B_{t_1}, \ldots, B_{t_k})) = \frac{1}{\mathcal{F}_{\beta}^\omega} \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^k \mathcal{F}_{\beta}^{\omega, a}([t_{i-1}, x_{i-1}], (t_i, x_i)) \, dx. \]

To conclude, we only need to show that almost surely

\[ \lim_{a \to 0} \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^k \mathcal{F}_{\beta}^{\omega, a}([t_{i-1}, x_{i-1}], (t_i, x_i)) \, dx_i = \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^k \mathcal{F}_{\beta}^\omega([t_{i-1}, x_{i-1}], (t_i, x_i)) \, dx. \]

In particular, taking \( g \equiv 1 \), this will give that (recall (3.10))

\[ \mathcal{F}_{\beta}^\omega = \int_{(\mathbb{R}^d)^k} \prod_{i=1}^k \mathcal{F}_{\beta}^\omega([t_{i-1}, x_{i-1}], (t_i, x_i)) \, dx_i. \]

Let us first treat the case \( \int_{[1, \infty)} v \lambda(du) < \infty \). In that case we have

\[ \mathbb{E} \left[ \left| \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^k \mathcal{F}_{\beta}^{\omega, a}([t_{i-1}, x_{i-1}], (t_i, x_i)) \, dx \right. \right. \]

\[ - \left. \left. \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^k \mathcal{F}_{\beta}^\omega([t_{i-1}, x_{i-1}], (t_i, x_i)) \, dx \right] \right] \leq \int_{(\mathbb{R}^d)^k} g(x) \mathbb{E} \left[ \left| \prod_{i=1}^k \mathcal{F}_{\beta}^{\omega, a}([t_{i-1}, x_{i-1}], (t_i, x_i)) - \prod_{i=1}^k \mathcal{F}_{\beta}^\omega([t_{i-1}, x_{i-1}], (t_i, x_i)) \right| \right] \, dx. \]
Moreover, we have (recall (3.4))
\[
\lim_{a \to 0} E \left[ \prod_{i=1}^{k} \mathcal{Z}^{\omega,a}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) - \prod_{i=1}^{k} \mathcal{Z}^{\omega}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) \right] = 0.
\]

and thus the convergence (6.10) holds in \( L_1 \) by dominated convergence. The fact that the convergence is also almost sure comes from the fact that the l.h.s. in (6.10) is a martingale.

In the case \( \int_{[1, \infty)} v \lambda (dv) = \infty \), we can apply (6.10) for the truncated environment and obtain that almost surely
\[
\lim_{a \to 0} \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^{k} \mathcal{Z}^{\omega,a}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) dx
\]
\[
= \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^{k} \mathcal{Z}^{\omega,0}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) dx.
\]

Bounding above \( \mathcal{Z}^{\omega,a}_{\beta} \) by \( \mathcal{Z}^{\omega}_{\beta} \) in the l.h.s. and using monotone convergence for the r.h.s., we obtain that
\[
\lim_{a \to 0} \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^{k} \mathcal{Z}^{\omega,a}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) dx
\]
\[
\geq \int_{(\mathbb{R}^d)^k} g(x) \prod_{i=1}^{k} \mathcal{Z}^{\omega}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) dx.
\]

Since the same inequality is also valid for \( 1 - g \), to conclude it is sufficient to check that we have equality when \( g \equiv 1 \). This corresponds to checking that
\[
\int_{(\mathbb{R}^d)^k} \prod_{i=1}^{k} \mathcal{Z}^{\omega}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) dx = \mathcal{Z}^{\omega}_{\beta}.
\]

But thanks to (6.10) (see in particular (6.11)), we have
\[
\int_{(\mathbb{R}^d)^k} \prod_{i=1}^{k} \mathcal{Z}^{\omega,0}_{\beta} ((t_{i-1}, x_{i-1}), (t_i, x_i)) dx = \mathcal{Z}^{\omega,0}_{\beta},
\]
for all \( b \), so (6.14) follows by monotone convergence.

7. stochastic heat equation with Lévy noise: proof of Proposition 2.26

Recall that Proposition 2.24 has been proved in Section 4 (see Remark 4.16). It remains to show that for fixed \( t \in [0,1] \) and \( x \in \mathbb{R}^d \), we have
\[
\lim_{a \to 0} \int_{\mathbb{R}^d} \mathcal{Z}^{\omega,a}_{\beta} ([0, y), (t,x)] u_0 (dy) = \int_{\mathbb{R}^d} \mathcal{Z}^{\omega}_{\beta} ([0, y), (t,x)] u_0 (dy)
\]

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and that the right-hand-side is finite. For simplicity, we assume that \( u_0 \) is a positive measure, since otherwise we simply treat the positive and negative parts of \( u_0 \) separately.

In the case \( \int_{[1,\infty)} v \lambda(dv) < \infty \), we can repeat the proof of Proposition 2.21. We have

\[
\mathbb{E}[|u^\alpha(t, x) - u(t, x)|] \leq \int_\mathbb{R}^d \mathbb{E}\left[\left|\mathcal{D}^{\alpha}_{\beta}[0, y]((0, y), (t, x)) - \mathcal{Z}_{\beta}^{\alpha}((0, y), (t, x))\right|\right] u_0(dy).
\]

Using (6.12)–(6.13) with \( k = 1 \) together with the fact that \( \int_{\mathbb{R}^d} \rho(y - x) u_0(dy) < \infty \) thanks to our assumption (2.25), we conclude by dominated convergence that

\[
\lim_{\alpha \to 0} u^\alpha(t, x) = u(t, x)
\]

in \( L_1 \) and almost surely (since \( (u^\alpha(t, x))_{\alpha \in [0,1]} \) is a martingale).

Let us now turn to the case \( \int_{[1,\infty)} v \lambda(dv) = \infty \). Recall the definition (2.27) of \( u^{(a,b)} \) and notice that for all \( a \in (0,1] \) and \( b \geq 1 \) we have \( u^{(a,b)}(t, x) \leq u^a(t, x) < \infty \) almost surely. Applying the \( L_1 \) and a.s. convergence with the truncated environment, we get that

\[
\lim_{\alpha \to 0} u^{(a,b)}(t, x) = u^{(a,b)}(t, x) := \int_{\mathbb{R}^d} \mathcal{D}^{\alpha}_{\beta}([0, y], (t, x)) u_0(dy), \quad \text{a.s.}
\]

To conclude, we need to show that we can take the limit \( b \to \infty \) uniformly for \( a \in (0,1] \). More precisely, similarly to (4.40), we show that for any \( \varepsilon > 0 \) we have

\[
\lim_{b \to \infty} P\left(\sup_{a \in [0,1]} (u^a(t, x) - u^{(a,b)}(t, x)) > \varepsilon\right) = 0.
\]

Indeed, for any \( b' > b \geq 1 \), considering the super-martingale

\[
\varepsilon \land (u^{(a,b')}(t, x) - u^{(a,b)}(t, x))_{a \in [0,1]}
\]

and applying Doob’s inequality, we get

\[
P\left(\sup_{a \in (0,1]} (u^{(a,b')}(t, x) - u^{(a,b)}(t, x)) > \varepsilon\right) \leq \frac{1}{\varepsilon} \mathbb{E}\left[\varepsilon \land (u^{(1,b')}(t, x) - u^{(1,b)}(t, x))\right].
\]

Sending \( b' \) to infinity we therefore get by monotone convergence (analogously to (4.43))

\[
P\left(\sup_{a \in (0,1]} (u^a(t, x) - u^{(a,b)}(t, x)) > \varepsilon\right) \leq \frac{1}{\varepsilon} \mathbb{E}\left[\varepsilon \land (u^1(t, x) - u^{(1,b)}(t, x))\right].
\]

Then, since \( u^1(t, x) < \infty \), the limit (7.1) follows by dominated convergence. As a by-product, this shows that \( u(t, x) < \infty \) a.s.

\[
\square
\]

**Appendix A. On Lévy noises**

Recall that we consider a Poisson process \( \omega \) on \( \mathbb{R} \times \mathbb{R}^d \times \mathbb{R}_+ \) with intensity \( dt \otimes dx \otimes \lambda(dv) \), where \( \lambda \) is a positive measure on \((0,\infty)\) with \( \lambda([1,\infty)) < \infty \). Recall the definition (2.5) of the truncated and centered measure \( \xi^{(a)}_\omega \) on \( \mathbb{R} \times \mathbb{R}^d \).
Proof and hence when

\[ \chi \left( \int_{0}^{1} v^{2} \lambda( dv) < \infty \right) \text{ converges in } H_{\text{loc}}^s(\mathbb{R} \times \mathbb{R}^d) \text{ with } s > (d + 1)/2, \text{ i.e., there exists a distribution } \xi_{\omega} \in H_{\text{loc}}^s(\mathbb{R} \times \mathbb{R}^d) \text{ such that for any non-negative smooth compactly supported } \psi \]

\[ \lim_{a \to 0} \| \psi(\xi_{\omega}^{(a)} - \xi_{\omega}) \|_{H^{1/2}} = 0. \]

Remark A.2. — The optimality of the criterion above on \( \lambda \) can be checked by computing the characteristic function of the random variable \( \int \theta d\xi_{\omega}^{(a)} \) for a smooth \( \theta \) with compact support. We have

\[ \chi^{(a)}(\theta) := \mathbb{E}[e^{i \int \theta d\xi_{\omega}^{(a)}}] \]

\[ = \exp \left( \int_{\mathbb{R}^d} e^{i \theta(t,x)} - 1 - i \theta(t,x) \frac{1}{(\theta(t,x))^2} \right) \, dt \, dx \, \lambda( dv) \].

When \( \int_{0}^{1} v^{2} \lambda( dv) = \infty \), we have \( \lim_{a \to 0} \chi^{(a)}(\theta) = \mathbb{1}_{(\theta)}(\theta) \) for any non-trivial \( \theta \), and hence \( \int \theta d\xi_{\omega}^{(a)} \) does not converge (even in distribution).

Proof. — Let \( \psi \) be non-negative, \( C^\infty \) and compactly supported. Let us show that \( (\psi \xi_{\omega}^{(a)})_{a \in (0,1)} \) is a Cauchy sequence in \( H^{-s}(\mathbb{R}^{1+d}) \). In this proof only, we denote \( \mathbb{R}^\mathbb{R} \).

First, we observe that \( \psi \xi_{\omega}^{(a)} \) belongs to \( H^{-s}(\mathbb{R}^{1+d}) \) almost surely because its Fourier transform is smaller than \( |\xi_{\omega}^{(a)}(\psi) < +\infty, \text{ and } (1 + |z|^2)^{-s} \) is integrable as \( 2s > \theta \). Now, to show that \( \{\xi_{\omega}^{(a)}\} \) is a Cauchy sequence, we need to prove that

\[ \lim_{a \to 0} \sup_{b \in (0,a]} \| \psi(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)}) \|_{H^{-s}} = 0. \]

We have (recall the definition (1.7))

\[ \| \psi(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)}) \|_{H^{-s}}^2 = \int_{\mathbb{R}^\mathbb{R}} (1 + |z|^2)^{-s} \left| \int_{\mathbb{R}^\mathbb{R}} e^{iz \cdot x} \psi(x)(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)})(dx) \right|^2 \, dz \]

and hence by Fatou’s lemma

\[ \sup_{b \in (0,a]} \| \psi(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)}) \|_{H^{-s}}^2 \leq \int_{\mathbb{R}^\mathbb{R}} (1 + |z|^2)^{-s} \sup_{b \in (0,a]} \left| \int_{\mathbb{R}^\mathbb{R}} e^{iz \cdot x} \psi(x)(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)})(dx) \right|^2 \, dz. \]

Now, as \( \int (\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)})(dx) e^{iz \cdot x} \psi(x)(dx) \) is a (complex valued) time reversed martingale with càdlàg path, Doob’s maximal inequality yields

\[ \mathbb{E} \left[ \sup_{b \in (0,a]} \left| \int_{\mathbb{R}^\mathbb{R}} e^{iz \cdot x} \psi(x)(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)})(dx) \right|^2 \right] \leq \lim_{b \to 0} \mathbb{E} \left[ \left| \int_{\mathbb{R}^\mathbb{R}} e^{iz \cdot x} \psi(x)(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)})(dx) \right|^2 \right]. \]

The right-hand side is then straightforward to compute. Expanding the square and using the fact that the off-diagonal integral

\[ \int_{\mathbb{R}^\mathbb{R} \times \mathbb{R}^\mathbb{R}} \psi(x) \psi(y) \mathbb{1}_{\{x \neq y\}} e^{iz \cdot (x - y)}(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)})(dx)(\xi_{\omega}^{(a)} - \xi_{\omega}^{(b)})(dy) \]


has zero mean, we obtain that for any $a > b > 0$
\[
\mathbb{E} \left[ \int_{\mathbb{R}^d} e^{iz \cdot x} \psi(x) \left( \xi_\omega^{(a)} - \xi_\omega^{(b)} \right)(dx) \right]^2 = \mathbb{E} \left[ \sum_{(x,v) \in \omega} v^2 1_{\{v \in (b,a)\}} \psi(x)^2 \right].
\]
We can compute the last expression using the formula for Poisson point process (see Proposition 3.8): we obtain
\[
\mathbb{E} \left[ \sum_{(x,v) \in \omega} v^2 1_{\{v \in (b,a)\}} \psi(x)^2 \right] \leq \mathbb{E} \left[ \sum_{(x,v) \in \omega} v^2 1_{\{v \in (0,a)\}} \psi(x)^2 \right] = \int_{\mathbb{R}^d} \psi(x)^2 dx \int_{(0,a)} v^2 \lambda(dv).
\]
Hence we obtain from (A.1)
\[
\mathbb{E} \left[ \sup_{b \in (0,a]} ||\psi(\xi_\omega^{(a)}) - \xi_\omega^{(b)}||_{H^{-s}}^2 \right] \leq \int_{\mathbb{R}^d} (1 + |z|^2)^{-s} dx \int_{(0,a)} v^2 \lambda(dv),
\]
which is sufficient to conclude that $\lim_{a \to \infty} \sup_{b \in (0,a]} ||\psi(\xi_\omega^{(a)}) - \xi_\omega^{(b)}||_{H^{-s}}^2 = 0 \ a.s.,$
by Fatou’s lemma.

\section*{Appendix B. Proofs of some properties of $\mathcal{Z}_\beta^{\omega,a}$}

\subsection*{B.1. Alternative representation of $\mathcal{Z}_\beta^{\omega,a}$: proof of Lemma 3.3.}

For $\sigma \in \mathcal{P}(\omega)$, we let $\mathcal{E}^k(\sigma)$ denote the collection of $k$ space-time points which include $(t_i, x_i)_{i=1}^{\sigma}$ and no other space-points of the Poisson process, that is
\[
\mathcal{E}^k(\sigma) := \{(t, x) \in \mathbb{R}^d : \{(t_i, x_i)\}_{i=1}^{\sigma} \cap \pi(\omega) = \pi(\sigma)\},
\]
where $\pi$ denote the projection on the first two coordinates. The following technical result, which immediately implies Lemma 3.3, establishes that $w_{\alpha,\beta}(\sigma, f)$ (recall its definition (3.6)) corresponds to the contribution to the partition function of the integral over the disjoint unions of $\mathcal{E}^k(\sigma)$, $k \geq |\sigma|$. The idea is that the integration over $\mathcal{E}^k(\sigma)$ decouples the Poisson and the Lebesgue parts of $\xi_\omega^{(a)}$, i.e., the contribution of points respectively inside and outside of $\sigma$.

\begin{lemma}

For any $f \in \mathcal{B}$ and any given $\sigma \in \mathcal{P}(\omega)$ we have
\[
w_{\alpha,\beta}(\sigma, f) := e^{-\beta \kappa_\alpha \beta |\sigma|} \varrho(t, x, f) \prod_{i=1}^{\sigma} u_i 1_{(u_i, \infty)} = \sum_{k \geq |\sigma|} \beta^k \int_{\mathcal{E}^k(\sigma)} \varrho(t, x, f) \prod_{i=1}^{k} \xi_\omega^{(a)}(dt_i, dx_i).
\]

When $\sigma = \emptyset$ the term $k = 0$ in the sum is by convention equal to $Q(f)$.
\end{lemma}

In order to lighten notations, we write the proof only in the case of a function $f \equiv 1$. We assume that $\int_{[1, \infty)} v \lambda(dv) < \infty$ so that all the integrals below are well defined (recall Proposition 3.1). The general case with a function $f \in \mathcal{B}_b$ (for which all terms are well defined assuming that $\lambda([a, \infty)) < \infty$, thanks to Proposition 3.1) is a mere adaptation of notation.
Proof of Lemma B.1. — Let $\sigma$ be a fixed set of points with $|\sigma| = \ell \geq 1$ (the case $\sigma = \emptyset$ can be checked separately) and let $(t_i, x_i, u_i)_{i=1}^{\ell}$ denote the (time ordered) points in $\sigma$. Given $i \in [0, \ell]$, and $(s, y) \in \mathcal{E}^{(k)}(\sigma)$, we let $s_j^{(i)}$, $y_j^{(i)}$, $j \leq k$, denote the space time points of $(s, y)$ in the time interval $(t_i, t_{i+1})$ (note that $k_i$ here is a function of $(s, y)$).

Then by splitting the integrals according the the value of the $k_i$'s, grouping the terms and factoring, we obtain that $\sum_{k_i \geq 1} \beta^{k_i} \int_{\mathcal{E}^{(k)}(\sigma)} \rho(t, x) \prod_{i=1}^{k} \xi_j^{(a)}(dt_i, dx_i)$ is equal to

$$\begin{align*}
\beta' \prod_{i=0}^{\ell-1} d_{t_{i+1}} 1_{\{u_{i+1+} \leq a\}} \sum_{k_i = 0}^\infty (-\beta \kappa_a)^{k_i} \\
\times \int_{t_i < s_1^{(i)} < \cdots < s_k^{(i)} < t_{i+1}} \left( \prod_{j=1}^{k_i+1} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \right) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)})
\end{align*}$$

with the convention $\Delta_j y_j^{(i)} := y_j^{(i)} - s_j^{(i)}$, $s_0^{(i)} = t_i$, $s_{k_i+1}^{(i)} = t_{i+1}$ and analogously for $y_j^{(i)}$. When $k_i = 0$ resp. $k_\ell = 0$, the value of the above integrals are by convention

$$\int_{t_i < s_1^{(i)} < \cdots < s_k^{(i)} < t_{i+1}} \left( \prod_{j=1}^{k} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \right) \prod_{j=1}^{k} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) = 1.$$

Now, one can check that

$$\begin{align*}
\sum_{k_i = 0}^\infty (-\beta \kappa_a)^{k_i} \int_{t_i < s_1^{(i)} < \cdots < s_k^{(i)} < t_{i+1}} \left( \prod_{j=1}^{k_i+1} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \right) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \\
= \rho_{t_{i+1}-t_i}(x_{i+1} - x_i) \left( 1 + \sum_{k_i = 1}^{\infty} (-\beta \kappa_a)^{k_i} \int_{t_i < s_1^{(i)} < \cdots < s_k^{(i)} < t_{i+1}} \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \right) \\
= \rho_{t_{i+1}-t_i}(x_{i+1} - x_i) e^{-\beta \kappa_a(t_{i+1} - t_i)},
\end{align*}$$

and similarly

$$\sum_{k_i = 0}^\infty (-\beta \kappa_a)^{k_i} \int_{t_i < s_1^{(i)} < \cdots < s_k^{(i)} < t_{i+1}} \left( \prod_{j=1}^{k_i+1} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \right) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)}) \prod_{j=1}^{k_i} \rho_{\Delta_j y_j^{(i)}}(\Delta_j y_j^{(i)})$$

is equal to $e^{-\beta \kappa_a(t_{i+1} - t_i)}$, which concludes the proof. 

Proof of (3.9). — The proof of (3.9) works exactly as above when $\int_{[1, \infty)} \psi \lambda(dv) < \infty$. For the general case, one needs first to have an identity for a positive integral so
that we have no trouble with our definition. Truncating the environment and using monoton convergence we have

\[ \rho_t(x) + \sum_{k=1}^{\infty} \beta^k \int_{0 < t_1 < \cdots < t_k < t} \int_{(\mathbb{R}^d)^k} \varphi(t, x) \rho_{t-k}(x-x_k) \prod_{i=1}^{k} |\xi_{\lambda_i}^{(\omega)}|(dt_i, dx_i) \]

\[ = e^{2\alpha_1 t} \sum_{\mathcal{P}_{0,1}(\omega)} w_{a}(\sigma(t, x)). \]

This ensures that the sum and integrals in (2.17) are convergent if and only if we have \( \sum_{\mathcal{P}_{0,1}(\omega)} \omega(\sigma(t, x)) < \infty \). Then, repeating the proof above, we obtain that (3.9) holds.

\[ \Box \]

**B.2. The size-biased measure: proof of Lemma 3.7.** — Let us recall here, for the sake of clarity, the content of Lemma 3.7. The size-biased measure \( \hat{P}_\beta \) is defined as \( \hat{P}_\beta(J) = \mathbb{E}[\mathcal{P}_\beta^{\omega, a} 1_J] \). Then Lemma 3.7 states that for all bounded measurable function \( f \),

\[ \hat{P}_\beta[f(\omega)] = \mathbb{P} \otimes \mathbb{P}_\alpha' \otimes \mathbb{Q} \left[f(\tilde{\omega}(\omega, \omega', B))\right], \]

where \( \tilde{\omega}(\omega, \omega', B) = \omega \cup \{(t, B_1, u) : (t, u) \in \omega'\} \), with \( \mathbb{Q} \) the distribution of a standard Brownian motion \( B \), \( \mathbb{P}_\alpha' \) the distribution of a Poisson point process \( \omega'_a \) on \([0, 1] \times \mathbb{R}^+\) with intensity \( \alpha \otimes \beta \mathbb{1}_{\{\upsilon \geq a\}}(du) \) and \( \mathbb{P} \) the distribution of the Poisson point process \( \omega \) introduced in (1.5). Recall that we assume that \( \mu := \int_{[1, \infty)} v \lambda(du) < \infty \).

**Proof of Lemma 3.7.** — It is sufficient to check that the distributions of the two point processes in Equation (3.12) coincide when restricted to \([0, 1] \times \mathbb{R}^d \times [a, \infty)\), since their distributions outside of this set are unaltered by the size-biasing and remain independent of the rest.

Given a bounded measurable subset \( A \) of \([0, 1] \times \mathbb{R}^d \times [a, \infty)\), we define

\[ \mathcal{N}_A := \#(\omega \cap A). \]

Our proof starts with the observation that the distribution of simple point processes is completely characterized by \( \mathbb{P}(\mathcal{N}_A = 0) \) for all bounded and measurable set \( A \), see [41, Th.6.11], and hence a fortiori by the distribution of \( \mathcal{N}_A \).

Hence, setting \( \tilde{\mathcal{N}}_A := \#(\tilde{\omega} \cap A) \), it is sufficient for us to prove that for every set \( A \) and any \( k \geq 0 \)

\[ \frac{1}{k!} \hat{P}_\beta[\mathcal{N}_A(\mathcal{N}_A - 1) \cdots (\mathcal{N}_A - k)] = \frac{1}{k!} \mathbb{P} \otimes \mathbb{P}_\alpha' \otimes \mathbb{Q} \left[\tilde{\mathcal{N}}_A(\tilde{\mathcal{N}}_A - 1) \cdots (\tilde{\mathcal{N}}_A - k)\right], \]

and that the quantities above do not grow faster than exponentially, so that the distributions of \( \mathcal{N}_A \) and \( \tilde{\mathcal{N}}_A \) are indeed characterized by their moments.

Let us define \( f = f_{k,A} : ([0, 1] \times \mathbb{R}^d \times [a, \infty])^k \to \mathbb{R} \) by

\[ f_{k,A}(\{t_i, x_i, u_i\}_{i=1}^k) := \mathbb{1}_{\{t_1 < t_2 < \cdots < t_k\}} \prod_{i=1}^{k} \mathbb{1}_A(t_i, x_i, u_i). \]
Since almost surely, there are no two points in \( \omega \) with the same time coordinate we have almost-surely

\[
\sum_{(t_i, x_i, u_i)_{i=1}^k \in \omega^k} f_{k, A}((t_i, x_i, u_i)_{i=1}^k) = \frac{1}{k!} \mathcal{N}_A(\mathcal{N}_A - 1) \cdots (\mathcal{N}_A - k + 1)
\]

and the analogous identity is valid for \( \mathcal{N}_A \). Hence we can check that the identity (3.12) holds simply by applying Mecke’s formula (Proposition 3.8) to each side in (B.1).

Let us start with the (easier) case of \( \tilde{\omega} \), i.e., the right-hand side of (B.1). We set

\[
A_k := \{(t_i, x_i, u_i)_{i=1}^k \in A^k : t_1 < t_2 < \cdots < t_k \}
\]

and we obtain (recalling the definition of \( P_{a} \))

\[
P \otimes P_{a}' \left[ \sum_{(t_i, x_i, u_i)_{i=1}^k \in \tilde{\omega}^k} f_{k, A}((t_i, x_i, u_i)_{i=1}^k) \right] = \int_{A_k} \prod_{i=1}^k (dt_i \, dx_i \, \lambda(du_i) + \beta dt_i \delta_{B_i}(dx_i) \lambda(du_i)).
\]

Now expanding the product and averaging with respect to the Brownian motion \( B \) we obtain that the right-hand side in (B.1) is equal to

\[
\int_{A_k} \sum_{I \subseteq [1, k]} \beta^{|I|} u_I \rho_I(t, x) \prod_{i=1}^k dt_i \, dx_i \, \lambda(du_i),
\]

where \( u_I = \prod_{i \in I} u_i \) and \( \rho_I(t, x) := \prod_{j=1}^{|I|} \rho_{j, j-1}(x_j - x_{j-1}) \) with \( (i_j)_{j=1}^{|I|} \) the ordered indices of \( I \) (by convention \( i_0 = 0 \) and \( t_0 = 0, x_0 = 0 \)).

Now let us move to the left-hand side of (B.1), that is the expectation with respect to the size-biased measure. Recalling the definition (3.11) of \( P_{a}^\beta \) and the representation (3.8) of \( Z_{\mathcal{N}_A}^{\beta, a} \), it is equal to

\[
\sum_{(t_i, x_i, u_i)_{i=1}^k \in \omega^k} f_{k, A}((t_i, x_i, u_i)_{i=1}^k)
\]

\[
e^{-\beta \mu} \mathbb{E} \left[ \sum_{\sigma \in \mathcal{P}(\omega)} w_{a, \beta}(\sigma) \sum_{(t_i, x_i, u_i)_{i=1}^k} f_{k, A}((t_i, x_i, u_i)_{i=1}^k) \right].
\]

We are going to decompose the sum above according to how \( \sigma \) intersects with the points \( (t_i, x_i, u_i)_{i=1}^k \) that are arguments of \( f_{k, A} \). For any given \( I \subseteq [1, k] \) with \( |I| = m \) and \( \ell = (\ell_{i})_{i=1}^{m+1} \), we set

\[
\mathcal{I}_{I, \ell} = \left\{ \sigma \in \mathcal{P}(\omega) : (\sigma \cap (t_i, x_i, u_i)_{i=1}^k) = (t_{i_j}, x_{i_j}, u_{i_j})_{j \in I} \right. \\
and \forall j \in [1, m+1], \# \{(t, x, u) \in \sigma, t \in (t_{i_{j-1}}, t_{i_{j}})\} = \ell_{j} \}
\]

where \( (i_j)_{j=1}^m \) are the ordered elements of \( I \), with \( i_0 = 0 \) and \( i_{m+1} = k + 1 \) (and \( t_0 = 0, t_{k+1} = 1 \)) by convention. For this computation we introduce

\[
\kappa_a = \int_{[a, \infty)} v \lambda(\text{d}v) = \kappa_a + \mu.
\]

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Using again Mecke’s formula and recalling the definition (3.5) of $w_{a,\beta}(\sigma)$, we have, for any such $I$ and $\ell$,

$$
E \left[ \sum_{a \in \mathcal{P}_I} \sum_{(t_i, x_i, u_i)_{i=1}^m \in \omega^k} w_{a,\beta}(\sigma) \int_{A_k} f_{k,A}((t_i, x_i, u_i)_{i=1}^k) \right] = e^{-\beta \kappa_a} \int_{A_k} H_{\beta}^a(t_m, x_m, \ell_m) \prod_{j=1}^m \frac{G_{\beta}^a(t_{j-1}, t_j, x_{j-1}, x_j, \ell_j)}{\beta t_j! \ell_j!} \prod_{i=1}^k dt_i dx_i \lambda^k(du_i),
$$

where we set

$$
G_{\beta}^a(t, x, x', \ell) := \beta \ell \int_{t < s_1 < \cdots < s_{\ell} < t'} \int_{(\mathbb{R}^d)^{\ell}} \int_{(a,\infty)^{\ell}} \left( \prod_{i=1}^{\ell+1} \rho_{\Delta_i,s}(\Delta_i y) \right) \prod_{i=1}^{\ell} ds_i dy_i \lambda( dv_i ) = (\beta \nu_{\alpha})^\ell \frac{(t'-t)^\ell}{\ell!} \rho_{\ell-t}(x' - x),
$$

with $\Delta_i s = s_i - s_{i-1}$ ($s_0 = t$, $s_{\ell+1} = t'$) and $\Delta_i y = y_i - y_{i-1}$ ($y_0 = x$, $y_{\ell+1} = x'$) by convention; and we also set

$$
H_{\beta}^a(t, x, \ell) := \beta \ell \int_{t < s_1 < \cdots < s_{\ell} < t} \int_{(\mathbb{R}^d)^{\ell}} \int_{(a,\infty)^{\ell}} \left( \prod_{i=1}^{\ell} \rho_{\Delta_i,s}(\Delta_i y) \right) \prod_{i=1}^{\ell} ds_i dy_i \lambda( dv_i ) = (\beta \nu_{\alpha})^\ell \frac{(1-t)^\ell}{\ell!}.
$$

Replacing $G_{\beta}^a$ and $H_{\beta}^a$ by their value we obtain (recall that $t_{m+1} = 1$ by convention)

$$
E \left[ \sum_{a \in \mathcal{P}_I} \sum_{(t_i, x_i, u_i)_{i=1}^m \in \omega^k} w_{a,\beta}(\sigma) \int_{A_k} f_{k,A}((t_i, x_i, u_i)_{i=1}^k) \right] = e^{-\beta \kappa_a} \int_{A_k} \prod_{j=1}^{m+1} (\beta \nu_{\alpha})^{\ell_j} \frac{(t_j - t_{j-1})^{\ell_j}}{\ell_j!} \beta^{\ell_j} t_j! \rho_{\ell_j}^a(t, x) \prod_{i=1}^k dt_i dx_i \lambda(du_i).
$$

Summing over all the possible $\ell_j$ just results in a factor $e^{\beta \nu_{\alpha}} = e^{\beta \mu} e^{\beta \kappa_a}$ and thus after summing over $I$ we obtain

$$
\mathbb{E}_{\beta} \left[ \sum_{a \in \mathcal{P}_I} \sum_{(t_i, x_i, u_i)_{i=1}^m \in \omega^k} f_{k,A}((t_i, x_i, u_i)_{i=1}^k) \right] = e^{\beta \mu} \sum_{I \subseteq [1,k]} \int_{A_k} \beta^{\ell_I} t_I! \rho_I^a(t, x) \prod_{i=1}^k dt_i dx_i \lambda(du_i).
$$

All together, we find that (B.3) is equal to (B.2). This proves (B.1), which concludes the proof of Lemma 3.7. \qed

**Appendix C. Stochastic comparisons**

We provide here two results enabling us to compare some integrals with respect to the measure $\lambda$ to integrals with respect to the Lebesgue measure. In particular, they establish the two claims (4.21)–(4.22).

**Proposition C.1.** — Assume that

$$
\mu := \int_{[1,\infty)} v \lambda( dv ) < \infty \quad \text{and also} \quad \int_{(0,1)} v^p \lambda( dv ) < \infty
$$

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for some $p \in (1, 2)$. Then for $q \geq 1$ there is a constant $c_q$, satisfying $\lim_{q \to \infty} c_q q^{1-p} = 0$, such that for every $m \geq 1$ and any non-decreasing function $g : \mathbb{R}^m \to \mathbb{R}_+$, i.e., non-decreasing in every coordinate, with $\text{Support}(g) \subset (\varepsilon, \infty)^m$ for some positive $\varepsilon > 0$, we have

$$
\int_{(0, \mathbb{Q})^m} g(u_1, \ldots, u_m) \prod_{i=1}^m u_i \lambda(du_i) \leq (c_q)^m \int_{(0, \mathbb{Q})^m} g(u_1, \ldots, u_m) \prod_{i=1}^m u_i^{-p} du_i.
$$

**Proof.** — Let us begin with a few observations. First, we only need to treat the case $m = 1$ since applying the result successively to the functions $u_i \mapsto f(u_1, \ldots, u_m)$ concludes the proof. Second, we can work with a differentiable and bounded function $g$, the general case being obtained by monotone convergence.

By an integration by part, defining $\overline{p}_q(u) := \int_{(u, \mathbb{Q})} v \lambda(du)$, we get that

$$
(C.1) \quad \int_{(0, \mathbb{Q})} g(u) u \lambda(du) = \int_{[\varepsilon, \mathbb{Q})} g(u) u \lambda(du) = \int_{[\varepsilon, \mathbb{Q})} g'(u) \overline{p}_q(u) du.
$$

Now, an important observation is that under our assumptions we have that

$$
(C.2) \quad \overline{p}_q(u) := \int_{(u, \mathbb{Q})} v \lambda(du) \leq c_q u^{1-p}, \quad \forall u \in (0, \mathbb{Q}),
$$

for a constant $c_q$ satisfying $\lim_{q \to \infty} c_q q^{1-p} = 0$. Let us postpone the proof of (C.2), but we can already see that plugged in (C.1) and using that $g'(u) \geq 0$, it implies that

$$
\int_{(0, \mathbb{Q})} g(u) u \lambda(du) \leq c_q \int_{[\varepsilon, \mathbb{Q})} g'(u) u^{1-p} du = c_q g(q) q^{1-p} + (p-1) c_q \int_{[\varepsilon, \mathbb{Q})} g(u) u^{-p} du,
$$

where we have used another integration by parts in the last identity. Using again that $g$ is non-decreasing, we get that

$$
\int_{[\eta, \mathbb{Q})} g(u) u^{-p} du \geq \frac{1}{p-1} (1 - 2^{1-p}) g(q) q^{1-p},
$$

and we therefore end up with

$$
\int_{(0, \mathbb{Q})} g(u) u \lambda(du) \leq c_p c_q \int_{(0, \mathbb{Q})} g(u) u^{-p} du,
$$

where the constant $c_p$ only depends on $p$. This concludes the proof.

It remains to see why (C.2) is true. We consider the cases $u < 1$ and $u \geq 1$ separately. If $u < 1$, we use the fact that $c'_p := \int_{(0, 1)} v^p \lambda(du) < +\infty$ to get that

$$
\overline{p}_q(u) \leq \int_{(u, 1)} v^{1-p} e_p \lambda(dv) + \int_{[1, \infty)} v \lambda(dv) \leq c'_p u^{1-p} + \mu \leq c''_p u^{1-p},
$$

since $p > 1$. If $u \geq 1$, we simply use that

$$
\overline{p}_q(u) \leq c''_q u^{1-p} \quad \text{with} \quad c''_q := \sup_{u \in [1, \mathbb{Q})} \overline{p}_q(u) / u^{1-p}
$$

and notice that since $\overline{p}_q(u)$ is non-increasing and goes to 0 as $u \to \infty$ this implies that $\lim_{q \to \infty} q^{1-p} c''_q = 0$. Combining the above estimates gives (C.2). \hfill \Box
Proposition C.2. — Assume that $\int_{(0,1)} v^p \lambda(dv) < \infty$ for some $p \in (1,2)$. Then for any $q \geq 1$ there is a constant $C_q$, such that for every $m \geq 1$ and any non-increasing function $g : \mathbb{R}^m \to \mathbb{R}_+$ we have

$$\int_{(0,q)^m} g(u_1, \ldots, u_m) \prod_{i=1}^m u_i^q \lambda(du_i) \leq (C_q)^m \int_{(0,q)^m} g(u_1, \ldots, u_m) \prod_{i=1}^m u_i^{1-p} du_i.$$

Proof. — The proof is similar to that of Proposition C.1 above. Again, we only have to treat the case $m = 1$ and of a bounded and differentiable function $g$, with $\|g\|_{\infty} \leq 1$ to simplify.

Setting $F(u) := \int_{(0,u]} v^2 \lambda(dv)$ (which is finite for any $u \geq 0$), an integration by parts gives that

$$\int_{(0,q)} g(u) u^2 \lambda(du) = g(q) F(q) - \int_{(0,q)} g'(u) F(u) du.$$

Now, notice that there is a constant $C_q := \int_{(0,q]} v^p \lambda(dv) < \infty$ such that

$$F(u) = \int_{(0,u]} v^{2-p} v^p \lambda(dv) \leq C_q u^{2-p}, \quad \forall u \in (0,q].$$

Using that $g'(u) \leq 0$, we therefore get that

$$\int_{(0,q)} g(u) u^2 \lambda(du) \leq C_q g(q) q^{2-p} - C_q \int_{(0,q)} g'(u) u^{2-p} du = (2 - p) C_q \int_{(0,q)} g(u) u^{1-p} du,$$

where we used another integration by parts for the last identity. \qed

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