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# DIFFUSIVE DECAY OF THE ENVIRONMENT VIEWED BY THE PARTICLE 

PAUL DE BUYER AND JEAN-CHRISTOPHE MOURRAT


#### Abstract

We prove an optimal diffusive decay of the environment viewed by the particle in random walk among random independent conductances, with, as a main assumption, finite second moment of the conductance. Our proof, using the analytic approach of Gloria, Neukamm and Otto, is very short and elementary.


## 1. Introduction

In this paper we are interested in the speed of convergence of the environment viewed by the particle in the context of random walk among random independent conductances. We refer to [Bis11] for an introduction. We will first introduce the model and then state our main theorem.

Consider the $d$-dimensional lattice $\left(\mathbb{Z}^{d}, \mathbb{B}^{d}\right), d \geq 1$, with $\mathbb{B}^{d}$ the set of unoriented edges connecting any two points of $\mathbb{Z}^{d}$ at Euclidian distance 1 and set $\Omega:=[0,+\infty)^{\mathbb{B}^{d}}$. Given an environment $\omega=\left(\omega_{x, y}\right)_{\{x, y\} \in \mathbb{B}^{d}} \in \Omega$ we shall consider the associated Markov process $\left(X_{t}^{\omega}\right)_{t \geqslant 0}$ with jump rate between $x$ and $y$ given by the conductance $\omega_{x, y}$ and write $P_{x}^{\omega}$ for its law starting from $x \in \mathbb{Z}^{d}$. In many places we may simply write $\left(X_{t}\right)_{t \geqslant 0}$ for simplicity. The environment itself will also be a random variable. In fact, throughout the paper, we will assume that the conductances $\omega_{x, y},\{x, y\} \in \mathbb{B}^{d}$, are i.i.d. with common law $\mu$, whose support is included in $[1, \infty)$. The law of the environment is therefore the product measure $\mathbb{P}:=\mu^{\mathbb{B}^{d}}$ and we denote by $\mathbb{E}$ the associated expectation. Since $\mathbb{P}$ is a product measure, standard percolation arguments guarantee that that the Markov process $\left(X_{t}\right)_{t \geqslant 0}$ is well defined for almost all $\omega \in \Omega$, for all times, see e.g. [KV86, DMFGW89, Mou11]. Moreover it is reversible with respect to the counting measure, i.e. for all $x, y \in \mathbb{Z}^{d}$, it holds $P_{x}^{\omega}\left(X_{t}=y\right)=P_{y}^{\omega}\left(X_{t}=x\right)$.

Now, define the translation operators $\left(\theta_{z}\right)_{z \in \mathbb{Z}^{d}}$ given by $\left(\theta_{z} \omega\right)_{x, y}:=\omega_{x+z, y+z}$. Then the Markov process $(\omega(t))_{t \geqslant 0}:=\left(\theta_{X_{t}} \omega\right)_{t \geqslant 0}$, called the environment viewed by the particle, is reversible with respect to $\mathbb{P}$.

The aim of this short paper is to give an optimal quantitative bound on the decay to equilibrium of $(\omega(t))_{t \geqslant 0}$.

In order to state our theorem, we need some more notations. A function $f: \mathbb{Z}^{d} \times \Omega \rightarrow \mathbb{R}$ is said to be translation invariant if $f(x, \omega)=f\left(0, \theta_{x} \omega\right)$ for all $x \in \mathbb{Z}^{d}$ and local if $f(0, \cdot)$ depends only on a finite set of conductances. The smallest set satisfying that property

[^0]is called the support of $f$ and is denoted by $\operatorname{supp}(f)$, while $\# \operatorname{supp}(f)$ stands for the size of the support (i.e. the number of sites of $\mathbb{Z}^{d}$ contained in $\left.\operatorname{supp}(f)\right)$. For any translation invariant function $f$, we set $\mathbb{E}[f]:=\mathbb{E}[f(0, \cdot)]$.

Our main theorem is the following.
Theorem 1.1. Let $d \geq 3$. Assume that the law $\mu$ of the conductance has support in $[1, \infty)$ and finite second moment $\mathbb{E}\left[\omega_{\cdot}^{2}, \cdot\right]<\infty$. Then there exists a constant $C>0$ that depends only on $d$ such that for all local translation invariant function $f: \mathbb{Z}^{d} \times \Omega \rightarrow \mathbb{R}$ with $\mathbb{E}[f]=0$, all $x \in \mathbb{Z}^{d}$, it holds

$$
\begin{equation*}
\mathbb{E}\left(E_{x}^{\omega}\left[f\left(X_{t}, \omega\right)\right]^{2}\right) \leqslant C \mathbb{E}\left[\omega_{r, \cdot}^{2}\right] \# \operatorname{supp}(f)^{2} \frac{\mathbb{E}\left[f^{2}\right]}{t^{d / 2}} \quad \forall t>0 \tag{1.1}
\end{equation*}
$$

Let us comment on Theorem 1.1.
We first observe that similar results already exist in the literature. One of us [Mou11] proved a polynomial decay in dimension 1 with exponent $1 / 2$, in dimension 2 with behavior $\log t / t$, in dimension 3 to 6 with $t^{-1}$ and in dimension 7 and higher with $t^{\frac{-d}{2}+2}$. Moreover, in the right hand side of (1.1) appears a much stronger norm than only the $L^{2}$ norm (the sum of the $L^{\infty}$ norm of $f$ and the so-called triple norm $\|\|f\|\|$ that involves the infinite sum of $L^{\infty}$-norm of local gradients of $f$, a natural norm largely used in the statistical mechanics literature, see for example [Mar99] and also [BZ99, JLQY99, Rob10]). On the other hand, the results of [Mou11] hold without the assumption on the finiteness of the second moment of the conductance. More recently, Gloria, Neukamm and Otto [GNO14a] obtained, among other results, a polynomial decay $t^{-\frac{d}{2}-1}$ when the function $f$ is the divergence of some other function. Because we follow closely their approach, our result is not far from theirs, although they need to assume that conductances are bounded, i.e. that $\mu$ has support contained in a compact set and that the function $f$ considered is in $L^{p}$, where $p$ is very large and not explicit. Finally, we mention that [Yur86, CK08, GO11, GO12] are related papers that deal with estimates on the diffusion matrix and we refer to [GNO14b, AM14] and references therein for recent results on homogenization of random operators.

Then, we observe that the power in (1.1) is the best possible, and we note that it can be obtained under stronger assumptions in [GNO14a]. On the other hand, we believe that the result should hold without the assumption on the finiteness of the second moment of the conductance, which appears as a technical fact in our proof. The second moment $\mathbb{E}\left[f^{2}\right]$ is also best possible. Indeed, a smaller norm would imply some regularization effect that would in turn imply a spectral gap estimate which is known not to hold. For the same reason, there must be some dependence, in the right hand side of (1.1), on the size of the support of $f$. Finally, we observe that the assumption on the support of $\mu$ is also necessary since it is known, for some specific choice of function $f$, with conductances taking small values, that the decay of the heat-kernel coefficients can be very slow, see [BBHK08], which would interfer with the decay in (1.1).

As for the proof, diffusive scaling is usually obtained, including other settings on graphs such as interacting particle systems (Kawasaki dynamics), using functional inequalities of Nash type, see e.g. [Lig91, BZ99, Mou11], Harnack [De199] or weak Poincaré inequalities [BZ10, Rob10], see also [BCLSC95, SC97]. Induction techniques can also be
used [JLQY99, CCR05], or specific aspects of the model such as attractivness [Deu94]. However, none of these techniques seem, to the best of our knowledge, to apply to our setting. Here instead, we will mainly follow the PDEs ideas from [GNO14a], but with many simplifications due to our non specific choice of class of functions.

As a result we believe that our approach of the decay to equilibrium of the environment viewed by the particle improves upon known results into different directions (class of function, assumption on the conductance, short and elementary proof), but, as a counterpart, we are not able to extend the result of Gloria, Neukamm and Otto [GNO14a] to unbounded conductances for divergence functions.

The paper is organized as follows. In the next subsection we will give some more notations. Then, we will give the proof of Theorem 1.1. Such a proof will rely on a series of lemmas that we will prove later on. Finally, in the last section, using the theory of completely monotonic functions, we comment on a possible short way of extending the results of Gloria, Neukamm and Otto [GNO14a] to our setting.

### 1.1. Notations. In this section we give some more notations.

Given an environment $\omega \in \Omega$, the infinitesimal generator of the process $\left(X_{t}\right)_{t \geq 0}$ and its associated semi-group, acting on functions $f: \mathbb{Z}^{d} \times \Omega \rightarrow \mathbb{R}$ are given, for any $x, \omega \in \mathbb{Z}^{d} \times \Omega$ by

$$
\mathcal{L}^{\omega} f(x, \omega)=\sum_{|z|=1} \omega_{x, x+z}(f(x+z, \omega)-f(x, \omega)),
$$

respectively by

$$
P_{t}^{\omega} f(x, \omega)=\exp \left(\mathcal{L}^{\omega} t\right) f(x, \omega)=\sum_{y \in \mathbb{Z}^{d}} P_{x}^{\omega}\left(X_{t}=y\right) f(y, \omega) .
$$

In some situations, we may want to work with a given fixed environment. In that case, and to emphasize this fact, we shall use the letter $m$ instead of $\omega$ and call that given environment the walk scheme. A walk scheme is well-defined when there is a threshold such that the set of conductances above it does not percolate. This will be used in particular to evaluate the behaviour of $\left(X_{t}^{m}\right)_{t \geqslant 0}$.

In many places we shall use the following equality $P_{0}^{\theta_{x} \omega}\left(X_{t}=y\right)=P_{x}^{\omega}\left(X_{t}=y+x\right)$ that holds for all $x, y \in \mathbb{Z}^{d}$ and all $\omega \in \Omega$.

For simplicity of notation, we set $f_{t}(x, \omega):=P_{t}^{\omega} f(x, \omega)=E_{x}^{\omega} f\left(X_{t}, \omega\right), t \geq 0, x \in \mathbb{Z}^{d}$, $\omega \in \Omega$, where $E_{x}^{\omega}$ is the mean associated to $P_{x}^{\omega}$, and in many places we shall omit the dependence in $\omega$ when there is no ambiguity.

Next, we define three different gradients. Denote by $e_{1}, \ldots, e_{d}$ the canonical orthonormal basis of $\mathbb{Z}^{d}$ (i.e. for all $i, e_{i}=(0, \ldots, 0,1,0, \ldots, 0)$ with 1 at the $i^{\text {th }}$ coordinate), set $e_{-i}=-e_{i}, i=1, \ldots, d$ and $e_{0}:=(0, \ldots, 0)$ for the origin. With that notation in hand, we define the following local gradient of $g: \mathbb{Z}^{d} \rightarrow \mathbb{R}$ :

$$
D_{i} g(y):=g\left(y+e_{i}\right)-g(y) \quad-d \leqslant i \leqslant d, \quad y \in \mathbb{Z}^{d} .
$$

In particular, if $f: \mathbb{Z}^{d} \times \Omega \rightarrow \mathbb{R}$, then $D_{i} f(y, \omega)=f\left(y+e_{i}, \omega\right)-f(y, \omega), \omega \in \Omega$, and $D_{i} P_{x}^{\omega}\left(X_{t}=y\right)$ stands for the gradient applied to the mapping $y \mapsto P_{x}^{\omega}\left(X_{t}=y\right)$. Similarly,

$$
\nabla^{i} P_{x}^{m}\left(X_{t}=y\right):=P_{x+e_{i}}^{m}\left(X_{t}=y\right)-P_{x}^{m}\left(X_{t}=y\right) \quad-d \leqslant i \leqslant d, \quad x, y \in \mathbb{Z}^{d} .
$$

Therefore, the infinitesimal generator of a random walk with scheme $m$ can be written as

$$
\mathcal{L}^{m} f(x, \omega)=\sum_{-d \leqslant i \leqslant d} m_{x, x+e_{i}} D_{i} f(x, \omega) \quad x \in \mathbb{Z}^{d}, \quad \omega \in \Omega
$$

Finally, for $x \in \mathbb{Z}^{d}$, let $a(x)=\left\{\omega_{x, x+e_{i}} ; 1 \leqslant i \leqslant d\right\}, \bar{a}(x)=\left\{\omega_{e}, e \in \mathbb{B}^{d}\right\} \backslash a(x)=$ $\cup_{y \neq x} a(y)$ and $\mathbb{E}^{(x)}$ be the conditional expectation given $\bar{a}(x)$. Then, for $y \in \mathbb{Z}^{d}$, we set

$$
\partial_{x} f(y, \omega):=f(y, \omega)-\mathbb{E}^{(x)}[f(y, \omega)]
$$

so that $\mathbb{E}^{(x)}\left[\left(\partial_{x} f\right)^{2}\right]$ is nothing else than the variance of $f$ with respect to the conditional expectation $\mathbb{E}^{(x)}$. Two sites $x, y$ are neighbors, a property we denote by $x \sim y$, if $\{x, y\} \in \mathbb{B}^{d}$. Also, given $f: \Omega \rightarrow \mathbb{R}$, we say that $x$ is a neighbor of $y$ with respect to $f$, and write $x \underset{f}{\sim} y$, if $a(x-y) \cap \operatorname{supp}(f) \neq \emptyset$ (observe that this is not an equivalence relation). In particular, observe that, if $x$ is not a neighbor of $y$ with respect to a local translation invariant function $f$ (i.e. if $a(x-y) \cap \operatorname{supp}(f)=\emptyset)$ then $\partial_{x} f(y, \omega)=0$.
2. Variance decay for unbounded conductances: proof of Theorem 1.1

In this section we prove Theorem 1.1. The idea, following [GNO14a], is to decompose the variance $\mathbb{E}\left(E_{x}^{\omega}\left[f\left(X_{t}, \omega\right)\right]^{2}\right)$, using Efron-Stein's inequality, into an infinite sum of terms of the type $\mathbb{E}\left[\left(\partial_{y} P_{t} f\right)^{2}\right]$, which, by Duhamel's formula, are split into two different terms that need to be analyzed separately (the core of the proof). The point in using Duhamel's Formula is to commute the operators $P_{t}$ and $\partial_{y}$. The proof ends by applying some sort of Gronwall Lemma.

Proof. [Proof of Theorem 1.1] Let $f: \Omega \rightarrow \mathbb{R}$ be a local translation invariant function with $\mathbb{E}[f]=0$, and assume, by homogeneity and for simplicity, that $\mathbb{E}\left[f^{2}\right]=1$. Following [GNO14a], we apply Efron-Stein's Inequality and the Duhamel formula (that we recall below, in Lemma 2.1, for completeness) to bound $\mathbb{E}\left[f_{t}^{2}\right]$ :

$$
\begin{align*}
\mathbb{E}\left[f_{t}^{2}\right] & \leqslant \mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(\partial_{y} P_{t} f\right)^{2}\right]=\mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(P_{t} \partial_{y} f+\int_{0}^{t} P_{t-s} h_{s}(0, y, \omega) d s\right)^{2}\right] \\
& \leqslant 2 \mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(P_{t} \partial_{y} f\right)^{2}\right]+2 \mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(\int_{0}^{t} P_{t-s} h_{s}(0, y, \omega) d s\right)^{2}\right] \tag{2.1}
\end{align*}
$$

where $h_{s}(x, y, \omega):=\mathbb{E}^{(y)}\left[\mathcal{L} f_{s}(x, \omega)\right]-\mathcal{L} \mathbb{E}^{(y)}\left[f_{s}(x, \omega)\right], x, y \in \mathbb{Z}^{d}, \omega \in \Omega, s \geq 0$. Next we analyze each term of the right hand side of the latter separately and start with the first one.

First recall that if $a(y-x) \cap \operatorname{supp}(f)=\emptyset$ then $\partial_{y} f(x, \omega)=0$. Hence

$$
\begin{align*}
\mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(P_{t} \partial_{y} f\right)^{2}\right] & =\mathbb{E}\left[\sum_{y}\left(\sum_{x \in \mathbb{Z}^{d}} P_{0}^{\omega}\left(X_{t}=x\right) \partial_{y} f(x, \omega)\right)^{2}\right] \\
& \leqslant \# \operatorname{supp}(f) \sum_{y} \sum_{\substack{x: y \sim x \\
f}} \mathbb{E}\left[P_{0}^{\omega}\left(X_{t}=x\right)^{2} \partial_{y} f(x, \omega)^{2}\right] \tag{2.2}
\end{align*}
$$

By invariance by translation we have

$$
\begin{aligned}
\mathbb{E}\left[P_{0}^{\omega}\left(X_{t}=x\right)^{2} \partial_{y} f(x, \omega)^{2}\right] & =\mathbb{E}\left[P_{0}^{\theta-x \omega}\left(X_{t}=x\right)^{2} \partial_{y} f\left(x, \theta_{-x} \omega\right)^{2}\right] \\
& =\mathbb{E}\left[P_{0}^{\omega}\left(X_{t}=-x\right)^{2} \partial_{y-x} f(0, \omega)^{2}\right]
\end{aligned}
$$

Therefore, changing variables ( $\operatorname{set} x^{\prime}=y-x$ and $y^{\prime}=x^{\prime}-y$ ), it holds

$$
\begin{aligned}
\mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(P_{t} \partial_{y} f\right)^{2}\right] & \leqslant \# \operatorname{supp}(f) \sum_{y} \sum_{\substack{x: y \sim x \\
f}} \mathbb{E}\left[P_{0}^{\omega}\left(X_{t}=-x\right)^{2} \partial_{y-x} f(0, \omega)^{2}\right] \\
& \leqslant \# \operatorname{supp}(f) \sum_{y} \sum_{\substack{x^{\prime}: x^{\prime} \sim 0 \\
f}} \mathbb{E}\left[P_{0}^{\omega}\left(X_{t}=x^{\prime}-y\right)^{2} \partial_{x^{\prime}} f(0, \omega)^{2}\right] \\
& \leqslant \# \operatorname{supp}(f) \sum_{\substack{x^{\prime}: x^{\prime} \sim 0 \\
f}} \sum_{y^{\prime}} \mathbb{E}\left[P_{0}^{\omega}\left(X_{t}=y^{\prime}\right)^{2} \partial_{x^{\prime}} f(0, \omega)^{2}\right]
\end{aligned}
$$

Finally, using Lemma 2.2 below and the fact that $\mathbb{E}\left[\partial_{x^{\prime}} f(0, \cdot)^{2}\right] \leqslant 2 \mathbb{E}\left[f(0, \cdot)^{2}\right]=$ $2 \mathbb{E}\left[f^{2}\right]=2$, we conclude that, for some constant $C$ that depends only on $d$,

$$
\begin{equation*}
\mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(P_{t} \partial_{y} f\right)^{2}\right] \leqslant C \frac{\# \operatorname{supp}(f)^{2}}{(t+1)^{d / 2}} \tag{2.3}
\end{equation*}
$$

Next we focus on the second term in the right hand side of (2.1). Using Lemma 2.3 we have

$$
\mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(\int_{0}^{t} P_{t-s} h_{s}(0, y, \omega) d s\right)^{2}\right]=\mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(\int_{0}^{t} \sum_{i=1}^{d} D_{i} P_{0}^{\omega}\left(X_{t-s}=y\right) g_{s}(y, y, \omega, i) d s\right)^{2}\right]
$$

so that, by Minkowski's integral inequality and the invariance by translation, it holds

$$
\begin{align*}
& \mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(\int_{0}^{t} P_{t-s} h_{s}(0, y, \omega) d s\right)^{2}\right]^{1 / 2} \\
& \leqslant \sqrt{d} \int_{0}^{t}\left(\sum_{y} \sum_{i} \mathbb{E}\left[\left(D_{i} P_{0}^{\omega}\left(X_{t-s}=y\right)\right)^{2} g_{s}(y, y, \omega, i)^{2}\right]\right)^{1 / 2} d s \\
&= \sqrt{d} \int_{0}^{t}\left(\sum_{y} \sum_{i} \mathbb{E}\left[\left(\nabla^{i} P_{0}^{\omega}\left(X_{t-s}=y\right)\right)^{2} g_{s}(0,0, \omega, i)^{2}\right]\right)^{1 / 2} d s \\
& \leqslant \sqrt{2 d} \int_{0}^{t}\left(\sum_{y} \sum_{i} \mathbb{E}\left[\left(\nabla^{i} P_{0}^{\omega}\left(X_{t-s}=y\right)\right)^{2} \mathbb{E}^{(0)}\left[\omega_{0, e_{i}} D_{i} f_{s}(0, \omega)\right]^{2}\right]\right. \\
&\left.+\mathbb{E}\left[\omega_{0, e_{i}}^{2}\left(\nabla^{i} P_{0}^{\omega}\left(X_{t-s}=y\right)\right)^{2} \mathbb{E}^{(0)}\left[D_{i} f_{s}(0, \omega)\right]^{2}\right]\right)^{1 / 2} d s \tag{2.4}
\end{align*}
$$

where $g_{s}(x, y, \omega, i):=\mathbb{E}^{(x)}\left[\omega_{y, y+e_{i}} D_{i} f_{s}(y, \omega)\right]-\omega_{y, y+e_{i}} \mathbb{E}^{(x)}\left[D_{i} f_{s}(y, \omega)\right], s \geq 0, x, y \in \mathbb{Z}^{d}$, $\omega \in \Omega$ and $i=1, \ldots, d$. Therefore, using twice that $(a+b)^{2} \leqslant 2 a^{2}+2 b^{2}$, Lemma 2.2
and Lemma 2.4 guarantee that, for some constant $C$ that depends only on $d$, it holds

$$
\begin{align*}
& \mathbb{E}\left[\sum_{y \in \mathbb{Z}^{d}}\left(\int_{0}^{t} P_{t-s} h_{s}(y, 0, \omega) d s\right)^{2}\right]^{1 / 2} \leqslant C \int_{0}^{t}\left((t-s+1)^{-\frac{d}{2}} \sum_{i} \mathbb{E}\left[\mathbb{E}^{(0)}\left[\omega_{0, e_{i}} D_{i} f_{s}(0, \omega)\right]^{2}\right]\right. \\
& \left.\quad+(t-s+1)^{-d / 2} \sum_{i} \mathbb{E}\left[\omega_{0, e_{i}}^{2} \mathbb{E}^{(0)}\left[D_{i} f_{s}(0, \omega)\right]^{2}\right]\right)^{1 / 2} d s \\
& (2.5)  \tag{2.5}\\
& \quad \leqslant \sqrt{2} C \int_{0}^{t}(t-s+1)^{-d / 4} \mathbb{E}\left[\omega_{0, e_{1}}^{2}\right]^{1 / 2}\left(-\partial_{s} \mathbb{E}\left[\left|f_{s}(0, \omega)\right|^{2}\right]\right)^{1 / 2} d s .
\end{align*}
$$

Plugging (2.3) and (2.5) into (2.1), we end up with

$$
\mathbb{E}\left[f_{t}^{2}\right]^{\frac{1}{2}} \leqslant C^{\prime} \# \operatorname{supp}(f) \mathbb{E}\left[\omega_{0, e_{1}}^{2}\right]^{\frac{1}{2}}\left((t+1)^{-\frac{d}{4}}+\int_{0}^{t}(t-s+1)^{-\frac{d}{4}}\left(-\partial_{s} \mathbb{E}\left[f_{s}^{2}\right]\right)^{\frac{1}{2}} d s\right)
$$

for some constant $C^{\prime}$ that depends only on $d$. The expected result will finally follow from Lemma 2.6 with $a(t):=\mathbb{E}\left[f_{t}^{2}\right]^{\frac{1}{2}}$ and $\alpha=\frac{d}{4}$. Indeed, $a(t)$ is non-increasing since, using classical computations for reversible Markov processes, $\partial_{t} \mathbb{E}\left[f_{t}^{2}\right]=2 \mathbb{E}\left[f_{t} \mathcal{L} f_{t}\right]=$ $-\sum_{i} \mathbb{E}\left[\omega_{0, e_{i}}\left(D_{i} f_{t}\right)^{2}\right] \leq 0$. This ends the proof.

In the proof of Theorem 1.1 we used the following series of lemma. The first two lemmas are well known results from Probability Theory and Analysis. The others are technical.

Lemma 2.1. [Efron-Stein's Inequality and the Duhamel formula] The following holds.
(Efron-Stein's Inequality): Let $n>1$ and $f$ be a function of $X_{1}, \ldots, X_{n}, n$ independent variables, then

$$
\operatorname{Var}(f) \leqslant \sum_{i=1}^{n} \mathbb{E}\left[\operatorname{Var}^{(i)}(f)\right]
$$

where $\mathbb{V a r}{ }^{(i)}$ is the conditional variance given $\left\{X_{1}, \ldots, X_{n}\right\} \backslash\left\{X_{i}\right\}$.
(Duhamel's Formula): For all $t \geqslant 0$ and almost all $\omega \in \Omega$ it holds

$$
\partial_{y} P_{t} f(x, \omega)=P_{t} \partial_{y} f(x, \omega)+\int_{0}^{t} P_{t-s} h_{s}(x, y, \omega) d s
$$

where $h_{s}(x, y, \omega):=\mathbb{E}^{(y)}\left[\mathcal{L} f_{s}(x, \omega)\right]-\mathcal{L} \mathbb{E}^{(y)}\left[f_{s}(x, \omega)\right], x, y \in \mathbb{Z}^{d}, \omega \in \Omega, s \geq 0$.
Proof. Efron-Stein's Inequality, also called tensorisation of the variance (see e.g. [ $\mathrm{ABC}^{+} 00$, Proposition 1.4.1]), is a well known result following from Cauchy-Schwarz' inequality. See [LO00] for an extension to general $\phi$-entropy, see also [Mas07].

As for the Duhamel formula, we observe that

$$
\partial_{y} P_{t} f(x, \omega)=P_{t} \partial_{y} f(x, \omega)+\int_{0}^{t} \partial_{s} P_{t-s} \partial_{y} P_{s} f(x, \omega) d s
$$

which leads to the expected result, since $h_{s}(x, y, \omega)=\left(\partial_{y} \mathcal{L}-\mathcal{L} \partial_{y}\right) P_{s} f$ and $\partial_{t} P_{t}=$ $\mathcal{L} P_{t}=P_{t} \mathcal{L}$.

Lemma 2.2. There exists a constant $C$ (that depends only on d) such that for all welldefined walk scheme $m$ and for all $t>0$ it holds

$$
\sum_{y \in \mathbb{Z}^{d}} P_{0}^{m}\left(X_{t}=y\right)^{2} \leqslant C(t+1)^{-d / 2}
$$

Proof. The proof of this inequality is a consequence of the reversibility and the fact that the invariant measure is the uniform measure. Indeed

$$
\sum_{y \in \mathbb{Z}^{d}} P_{0}^{m}\left(X_{t}=y\right)^{2}=\sum_{y \in \mathbb{Z}^{d}} P_{0}^{m}\left(X_{t}=y\right) P_{y}^{m}\left(X_{t}=0\right)=P_{0}^{m}\left(X_{2 t}=0\right)
$$

which gives the desired result combining [CKS87, Theorem 2.1] and [Mou11, Proposition 10.2] in its first arXiv version.

Lemma 2.3. Given $f: \mathbb{Z}^{d} \times \Omega \rightarrow \mathbb{R}$, define $h_{s}(x, y, \omega):=\mathbb{E}^{(y)}\left[\mathcal{L} f_{s}(x, \omega)\right]-\mathcal{L} \mathbb{E}^{(y)}\left[f_{s}(x, \omega)\right]$ and $g_{s}(x, y, \omega, i)=\mathbb{E}^{(y)}\left[\omega_{x, x+e_{i}} D_{i} f_{s}(x, \omega)\right]-\omega_{x, x+e_{i}} \mathbb{E}^{(y)}\left[D_{i} f_{s}(x, \omega)\right], x, y \in \mathbb{Z}^{d}, \omega \in \Omega$, $s \geq 0$ and $i=0, \ldots, d$. Then, for all $s, t>0$ and all $x, y$ it holds

$$
P_{t} h_{s}(x, y, \omega)=-\sum_{i=1}^{d} D_{i} P_{x}\left(X_{t}=y\right) g_{s}(y, y, \omega, i)
$$

Proof. Recall the definition of $D_{i}$. On the one hand, by definition, we have

$$
\mathbb{E}^{(y)}\left[\mathcal{L} f_{s}(x, \omega)\right]=\sum_{i=1}^{d} \mathbb{E}^{(y)}\left[\omega_{x, x+e_{i}}\left(D_{i} f_{s}(y, \omega)\right)\right]-\mathbb{E}^{(y)}\left[\omega_{x, x-e_{i}}\left(D_{i} f_{s}\left(x-e_{i}, \omega\right)\right)\right]
$$

and

$$
\mathcal{L} \mathbb{E}^{(y)}\left[f_{s}(x, \omega)\right]=\sum_{i=1}^{d} \omega_{x, x+e_{i}} \mathbb{E}^{(y)}\left[\left(D_{i} f_{s}(x, \omega)\right)\right]-\omega_{x, x-e_{i}} \mathbb{E}^{(y)}\left[\left(D_{i} f_{s}\left(x-e_{i}, \omega\right)\right)\right]
$$

On the other hand, $\omega_{x, x+e_{i}}$ is $\bar{a}(y)$-measurable iff $x \neq y$ and $\omega_{x, x-e_{i}}$ is $\bar{a}(y)$-measurable iff $x-e_{i} \neq y$. Therefore,

$$
h_{s}(x, y, \omega)= \begin{cases}\sum_{i=1}^{d} g_{s}(y, y, \omega, i) & \text { if } x=y \\ -g_{s}(y, y, \omega, i) & \text { if } x-e_{i}=y, i=1, \ldots, d \\ 0 & \text { otherwise }\end{cases}
$$

It finally follows that

$$
\begin{aligned}
P_{t} h_{s}(x, y, \omega) & =\sum_{z \in \mathbb{Z}^{d}} P_{x}\left(X_{t}=z\right) h_{s}(z, y, \omega) \\
& =P_{x}\left(X_{t}=y\right) h_{s}(y, y, \omega)-\sum_{i=1}^{d} P_{x}\left(X_{t}=y+e_{i}\right) h_{s}\left(y+e_{i}, y, \omega\right) \\
& =\sum_{i=1}^{d}\left(P_{x}\left(X_{t}=y\right)-P_{x}\left(X_{t}=y+e_{i}\right)\right) g_{s}(y, y, \omega, i)
\end{aligned}
$$

which is the desired result.

Lemma 2.4. Let $f: \mathbb{Z}^{d} \times \Omega \rightarrow \mathbb{R}$. Then, for all $s>0$ it holds

$$
\begin{aligned}
& \sum_{i=1}^{d} \mathbb{E}\left[\left(\mathbb{E}^{(0)}\left[\omega_{0, e_{i}} D_{i} f_{s}(0, \omega)\right]\right)^{2}\right] \leqslant-\mathbb{E}\left[\omega_{0, e_{1}}^{2}\right] \partial_{s} \mathbb{E}\left[f_{s}(0, \omega)^{2}\right] \\
& \sum_{i=1}^{d} \mathbb{E}\left[\left(\omega_{0, e_{i}} \mathbb{E}^{(0)}\left[D_{i} f_{s}(0, \omega)\right]\right)^{2}\right] \leqslant-\mathbb{E}\left[\omega_{0, e_{1}}^{2}\right] \partial_{s} \mathbb{E}\left[f_{s}(0, \omega)^{2}\right]
\end{aligned}
$$

Remark 2.5. The proof actually leads to a better bound in the first inequality above. Indeed, one can replace the second moment of the conductance by the first moment. This refinement will anyway not be useful for our purpose.
Proof. [Proof of Lemma 2.4] Using Cauchy-Schwarz' inequality and the fact that $\omega_{0, e_{i}}$ is $a$ (0)-measurable, we have

$$
\begin{aligned}
\mathbb{E}\left[\left(\mathbb{E}^{(0)}\left[\omega_{0, e_{i}} D_{i} f_{s}(0, \omega)\right]\right)^{2}\right] & \leqslant \mathbb{E}\left[\mathbb{E}^{(0)}\left[\omega_{0, e_{i}}\right] \mathbb{E}^{(0)}\left[\omega_{0, e_{i}}\left(D_{i} f_{s}(0, \omega)\right)^{2}\right]\right] \\
& =\mathbb{E}\left[\omega_{0, e_{i}}\right] \mathbb{E}\left[\omega_{0, e_{i}}\left(D_{i} f_{s}(0, \omega)\right)^{2}\right]
\end{aligned}
$$

Using that $\partial_{t} \mathbb{E}\left[f_{t}^{2}\right]=\partial_{t} \mathbb{E}\left[f_{t}(0, \omega)^{2}\right]=-\sum_{i=-d}^{d} \mathbb{E}\left[\omega_{0, e_{i}}\left(D_{i} f_{s}(0, \omega)\right)^{2}\right]$ (a classical consequence of the reversibility) and summing over $i=1, \ldots, d$, we get

$$
\sum_{i=1}^{d} \mathbb{E}\left[\left(\mathbb{E}^{(0)}\left[\omega_{0, e_{i}} D_{i} f_{s}(0, \omega)\right]\right)^{2}\right] \leqslant-\mathbb{E}\left[\omega_{0, e_{1}}\right] \partial_{s} \mathbb{E}\left[f_{s}(0, \omega)^{2}\right]
$$

which leads to the first inequality since $\omega_{0, e_{1}} \geq 1$.
For the second inequality by conditioning and Jensen's inequality, we have

$$
\begin{aligned}
\mathbb{E}\left[\left(\omega_{0, e_{i}} \mathbb{E}^{(0)}\left[D_{i} f_{s}(0, \omega)\right]\right)^{2}\right] & =\mathbb{E}\left[\mathbb{E}^{(0)}\left[\omega_{0, e_{i}}^{2}\right]\left(\mathbb{E}^{(0)}\left[D_{i} f_{s}(0, \omega)\right]\right)^{2}\right] \\
& =\mathbb{E}\left[\omega_{0, e_{i}}^{2}\right] \mathbb{E}\left[D_{i} f_{s}(0, \omega)^{2}\right]
\end{aligned}
$$

Now, since $\omega_{0, e_{1}} \geq 1$, one has $\sum_{i=1}^{d} \mathbb{E}\left[D_{i} f_{s}(0, \omega)^{2}\right] \leq \sum_{i=1}^{d} \mathbb{E}\left[\omega_{0, e_{1}} D_{i} f_{s}(0, \omega)^{2}\right]=$ $-\partial_{s} \mathbb{E}\left[f_{s}(0, \omega)^{2}\right]$ which leads to the desired result and ends the proof of the lemma.

The last lemma is related to Lemma 15 of [GNO14a]. However, due to our specific setting, considering $p=1 / 2$, its proof is a bit simpler. We give it for completeness.
Lemma 2.6. [[GNO14a]] Let $\alpha>1 / 2$ and $a: \mathbb{R}^{+} \rightarrow \mathbb{R}^{+}$be a $\mathcal{C}^{1}$ non-increasing function. Let $b(t)=\sqrt{-2 a^{\prime}(t) a(t)}, t \geq 0$. Assume that

$$
a(t) \leqslant C\left((t+1)^{-\alpha}+\int_{0}^{t}(t-s+1)^{-\alpha} b(s) d s\right) \quad \forall t \geq 0
$$

for some constant $C$. Then there exists a constant $C_{\alpha}$ that depends only on $\alpha$ such that $a(t) \leqslant C_{\alpha} \max (C, a(0))(t+1)^{-\alpha}$.
Proof. Throughout the proof we use that $u \lesssim v$ if there exists a constant $A$ that depends only on $\alpha$ such that $u \leq A v$. The expected result will follow from the fact that, for some $t_{o}>0$ that will be chosen later on, $\left[t_{o}, \infty\right) \ni t \mapsto \Lambda(t):=\sup _{t_{o} \leq s \leqslant t}(s+1)^{\alpha} a(s)$ is a bounded function, bounded by $C^{\prime} \max (C, a(0))$ for some $C^{\prime}$ depending only on $\alpha$.

Indeed, since $a$ is non-increasing $a(s) \leq a(0)$ for any $s \in\left[0, t_{o}\right]$ which, together with the bound on $\Lambda$ would lead to the desired conclusion.

Our starting point is the following inequality obtained using that $a$ is non-increasing.

$$
\begin{aligned}
a(t) & \leqslant \frac{2}{t} \int_{\frac{t}{2}}^{t} a(u) d u \leqslant \frac{2 C}{t} \int_{\frac{t}{2}}^{t} \frac{d u}{(u+1)^{\alpha}}+\frac{2}{t} \int_{\frac{t}{2}}^{t} \int_{0}^{u} \frac{b(s) d s d u}{(u+1-s)^{\alpha}} \leq \frac{C}{\left(\frac{t}{2}+1\right)^{\alpha}} \\
& +\frac{2}{t} \int_{\frac{t}{2}}^{t} \int_{0}^{T} \frac{b(s) d s d u}{(u+1-s)^{\alpha}}+\frac{2}{t} \int_{\frac{t}{2}}^{t} \int_{T}^{u / 2} \frac{b(s) d s d u}{(u+1-s)^{\alpha}}+\frac{2}{t} \int_{\frac{t}{2}}^{t} \int_{u / 2}^{u} \frac{b(s) d s d u}{(u+1-s)^{\alpha}} \\
6) & =\frac{C}{\left(\frac{t}{2}+1\right)^{\alpha}}+I+I I+I I I
\end{aligned}
$$

where $T \in[1, t / 4]$ is a parameter that will be chosen later on. In order to bound $I, I I$ and $I I I$, we will repeatedly use the following bounds, that holds for all $T_{1}<T_{2}$ and whose proof is given below

$$
\int_{T_{1}}^{T_{2}} b(s) d s \lesssim\left\{\begin{array}{l}
\sqrt{T_{2}-T_{1}} a\left(T_{1}\right)  \tag{2.7}\\
\Lambda\left(T_{2}\right) T_{1}^{\frac{1}{2}-\alpha}
\end{array} \quad \text { if } T_{1}>0\right.
$$

To prove the first inequality, we use Cauchy-Schwarz' inequality. Namely

$$
\frac{1}{T_{2}-T_{1}}\left(\int_{T_{1}}^{T_{2}} b(s) d s\right)^{2} \leqslant \int_{T_{1}}^{T_{2}} b^{2}(s) d s=-\int_{T_{1}}^{T_{2}} \frac{d}{d s} a^{2}(s) d s \leqslant a\left(T_{1}\right)^{2}
$$

To prove the second inequality, we repeatedly use the latter. set $N=\left\lceil\log _{2}\left(T_{2} / T_{1}\right)\right\rceil$, then we have

$$
\begin{aligned}
\int_{T_{1}}^{T_{2}} b(s) d s & =\sum_{n=0}^{N-1} \int_{2^{n} T_{1}}^{2^{n+1} T_{1}} b(s) d s \leqslant \sum_{n=0}^{N-1} \sqrt{2^{n} T_{1}} a\left(2^{n} T_{1}\right) \leqslant \sum_{n=0}^{N-1}\left(2^{n} T_{1}\right)^{\frac{1}{2}-\alpha} \Lambda\left(2^{n} T_{1}\right) \\
& \leqslant \Lambda\left(T_{2}\right) T_{1}^{\frac{1}{2}-\alpha} \sum_{n=0}^{N-1}\left(2^{\frac{1}{2}-\alpha}\right)^{n} \lesssim \Lambda\left(T_{2}\right) T_{1}^{\frac{1}{2}-\alpha}
\end{aligned}
$$

Now, since $T \leq t / 4$ and thanks to (2.7), we have

$$
I:=\frac{2}{t} \int_{\frac{t}{2}}^{t} \int_{0}^{T} \frac{b(s) d s d u}{(u+1-s)^{\alpha}} \leqslant \frac{2}{t} \int_{\frac{t}{2}}^{t} \int_{0}^{T} \frac{b(s) d s d u}{\left(\frac{t}{4}+1\right)^{\alpha}}=\frac{\int_{0}^{T} b(s) d s}{\left(\frac{t}{4}+1\right)^{\alpha}} \lesssim \frac{\sqrt{T} a(0)}{(t+1)^{\alpha}}
$$

Again using (2.7) we have

$$
I I:=\frac{2}{t} \int_{\frac{t}{2}}^{t} \int_{T}^{u / 2} \frac{b(s) d s d u}{(u+1-s)^{\alpha}} \leq \frac{2}{t} \int_{\frac{t}{2}}^{t} \frac{\int_{T}^{u / 2} b(s) d s}{\left(\frac{t}{4}+1\right)^{\alpha}} d u \leq \frac{\frac{2}{t} \int_{\frac{t}{2}}^{t} \Lambda\left(\frac{u}{2}\right) T^{\frac{1}{2}-\alpha} d u}{\left(\frac{t}{4}+1\right)^{\alpha}} \lesssim \frac{\Lambda(t) T^{\frac{1}{2}-\alpha}}{(t+1)^{\alpha}}
$$

In order to bound the third term, which is more intricate, we first use the Fubini Theorem, noting that $\mathbb{1}_{t / 2 \leqslant u \leqslant t} \mathbb{1}_{u / 2 \leqslant s \leqslant u} \leqslant \mathbb{1}_{t / 4 \leqslant s \leqslant t} \mathbb{1}_{s \leqslant u \leqslant t}$, and that $\alpha>1 / 2$ to get

$$
\begin{aligned}
I I I & \leqslant \frac{2}{t} \int_{\frac{t}{4}}^{t} \int_{s}^{t}(u+1-s)^{-\alpha} d u b(s) d s=\frac{2}{t} \int_{\frac{t}{4}}^{t} \int_{0}^{t-s}\left(t^{\prime}+1\right)^{-\alpha} d t^{\prime} b(s) d s \\
& \leqslant \frac{2}{t} \int_{0}^{t}\left(t^{\prime}+1\right)^{-\alpha} d t^{\prime} \times \int_{\frac{t}{4}}^{t} b(s) d s \lesssim(t+1)^{-\frac{1}{2}-\alpha} \Lambda(t)
\end{aligned}
$$

Therefore, plugging the previous bounds on $I, I I$ and $I I I$ into 2.6 , it holds

$$
(1+t)^{\alpha} a(t) \leq A \sqrt{T} \max (C, a(0))+A \Lambda(t)\left(T^{\frac{1}{2}-\alpha}+\frac{1}{\sqrt{t}}\right)
$$

for some constant $A$ that depends only on $\alpha$. Now, since $\alpha>1 / 2$, there exist $t_{o} \geq 4$ and $T \geq 1$ such that for all $t \geq t_{o}, 1 / \sqrt{t} \leq 1 /(4 A)$ and $T^{\frac{1}{2}-\alpha} \leq 1 /(4 A)$ so that, taking the supremum and using the monotonicity of $\Lambda$, it holds $\Lambda(t) \leq A \sqrt{T} \max (C, a(0))+$ $\frac{1}{2} \Lambda(t)$ for all $t \geq t_{o}$ which leads to the desired conclusion. The proof of the lemma is complete.

## 3. Additional Remarks

3.1. Completely monotonic functions. In this section, we prove some results on $P_{x}^{m}\left(X_{t}=y\right)$, for a given walk scheme $m$, using the notion of completely monotonic functions. Recall that a function $f:(0, \infty) \rightarrow \mathbb{R}$ is said to be completely monotonic if it possesses derivatives $f^{(n)}$ of all orders and if $(-1)^{n} f^{(n)}(x) \geq 0$ for all $x>0$ and all $n=0,1,2, \ldots$ (see e.g. [Fel71, MS01]).
Proposition 3.1. Let $C, \alpha>0$. Assume that $f:(0, \infty) \rightarrow \mathbb{R}$ is a completely monotonic function satisfying for all $t>0, f(t) \leqslant \frac{C}{t^{\alpha}}$. Then, for all $t>0$, it holds $-f^{\prime}(t) \leq \frac{C^{\prime}}{t^{\alpha+1}}$ for some constant $C^{\prime}$ that depends only on $C$ and $\alpha$.

Remark 3.2. At the price of some technicalities, the above result can be extended to more general decay (i.e. replacing $1 / t^{\alpha}$ by some general completely monotonic function $g$ with $\left.\lim _{\infty} g=0\right)$.

Proof. Without loss of generality assume that $f(0)=1$. It is well known (see [Fel71, Wid41]) that $f$ is the Laplace transform of a positive random variable $X$, namely that, for all $t>0, f(t)=\mathbb{E}\left[e^{-t X}\right]$ where $\mathbb{E}$ denotes the mean of the law of $X$. Using the Markov Inequality we get for all $\lambda>0$ and all $x>0$

$$
\mathbb{P}(X \leqslant x) \leqslant \mathbb{E}\left[e^{-\lambda X}\right] e^{\lambda x} \leqslant C \exp (\lambda x-\alpha \log \lambda)
$$

Optimizing over the $\lambda>0$ we get $\left(\operatorname{since}^{\inf }{ }_{\lambda>0}\{\lambda x-\alpha \log \lambda\}=\alpha-\alpha \log \alpha+\alpha \log x\right.$ (the minimum being reached at $\lambda=\alpha / x)) \mathbb{P}(X \leqslant x) \leqslant C e^{\alpha}\left(\frac{x}{\alpha}\right)^{\alpha}$. Therefore, using Fubini's theorem, for all $t>0$ it holds

$$
\begin{aligned}
-f^{\prime}(t) & =\mathbb{E}\left[X e^{-t X}\right]=\mathbb{E}\left[X \int_{X}^{\infty} t e^{-t x} d x\right]=\int_{0}^{\infty} t e^{-t x} \mathbb{E}\left[X \mathbb{1}_{X \leq x}\right] d x \\
& \leq \int_{0}^{\infty} t x e^{-t x} \mathbb{P}(X \leq x) d x \leq \frac{C_{\alpha}}{t^{\alpha}} \int_{0}^{\infty}(t x)^{\alpha+1} e^{-t x} d x=\frac{C_{\alpha} \Gamma(\alpha+2)}{t^{\alpha+1}}
\end{aligned}
$$

which ends the proof.
Corollary 3.3. There exists a constant $C$ such that for all well-defined walk scheme $m$ and all $t>0$ it holds $\sum_{y \in \mathbb{Z}^{d},|z|=1}\left(P_{0}^{m}\left(X_{t}=y\right)-P_{0}^{m}\left(X_{t}=y+z\right)\right)^{2} \leqslant C /(t+1)^{\frac{d}{2}+1}$.
Proof. For $p \geq 1$, set $\|f\|_{p}^{p}:=\sum_{x}|f(x)|^{p}$ and, given an operator $P$ acting on functions, $\|P\|_{p \rightarrow q}=\sup \|P f\|_{q}$, where the supremum is taken over all $f$ with $\|f\|_{p}=1$.

First, we observe that the quantity is bounded as a consequence of Lemma 2.2. As a matter of fact, for all $y \in \mathbb{Z}^{d}$ and $t \geqslant 0, \sum_{x} P_{y}^{m}\left(X_{t}=x\right)^{2}=P_{y}^{m}\left(X_{t}=y\right) \leqslant 1$.

Then, note that $t \mapsto\left\|P_{t}^{m} f\right\|_{2}^{2}$ is a completely monotonic function. It is due to the fact that $\mathcal{L}^{m}$ is self-adjoint ( $<f, \mathcal{L}^{m} g>=<\mathcal{L}^{m} f, g>$ ) and defines a negative quadratic form $\left(<\mathcal{L}^{m} f, f>\leqslant 0\right)$. Indeed, for any function $f, \partial_{t}<P_{t}^{m} f, P_{t}^{m} f>=2<\mathcal{L} P_{t}^{m} f, P_{t}^{m} f>\leqslant$ $0, \partial_{t} \partial_{t}<P_{t}^{m} f, P_{t}^{m} f>=4<\mathcal{L}^{m} P_{t}^{m} f, \mathcal{L}^{m} P_{t}^{m} f>\geqslant 0$, and because $\mathcal{L}^{m}$ commutes with $P_{t}^{m}$, we can conclude repeating these last arguments with the function $\mathcal{L}^{m} f$. Hence, using Proposition 3.1, there exists a constant $C^{\prime}$ that depends only on $d$ and $C$ such that for all $f \in \ell^{1}, \frac{d}{d t}\left\|P_{t}^{m} f\right\|_{2}^{2} \leqslant C^{\prime} / t^{\frac{d}{2}+1}$.

On the other hand, by definition of $\mathcal{L}^{m}$ and $P_{t}^{m}$, we have

$$
\begin{equation*}
\frac{d}{d t}\left\|P_{t}^{m} f\right\|_{2}^{2}=-\sum_{x,|z|=1} m_{x, x+z}\left(P_{t}^{m} f(x+z)-P_{t}^{m} f(x)\right)^{2} \tag{3.1}
\end{equation*}
$$

Observe that for $f=\mathbb{1}_{\{0\}}$ (the function that equals 1 at 0 and 0 elsewhere), we have $f \in \ell^{1}$ and $P_{t}^{m} f(x)=\sum_{y \in \mathbb{Z}^{d}} P_{x}^{m}\left(X_{t}=y\right) f(y)=P_{0}^{m}\left(X_{t}=x\right)$ which plugged in (3.1) gives the desired result.
3.2. Gloria, Neukamm and Otto with a fixed walk scheme. Using the monotonic function approach above, we can prove a stronger decay of the variance of the environment view by the particle, when the function $f$ is the divergence of an other function, but only when the walk scheme $m$ is fixed. This is a result (much weaker but) in the spirit of [GNO14a].
Proposition 3.4. There exists a constant $C>0$ such that for almost all walk scheme $m$, all $t \geqslant 0$ and all function $f=D_{i} g$, where $-d \leqslant i \leqslant d$, $g$ is local, translation-invariant and $\mathbb{E}\left[g^{2}\right]<\infty$, it holds $\mathbb{E}\left[\left(P_{t}^{m} f(0, \omega)\right)^{2}\right] \leqslant C \# \operatorname{supp}(g)^{2} \frac{\mathbb{E}\left[g^{2}\right]}{(t+1)^{\frac{d}{2}+1}}$.

Proof. Using the Cauchy-Schwarz Inequality and that $\mathbb{E}[g(x, \omega) g(y, \omega)]=0$ as soon as $\operatorname{supp}(g(x, \omega)) \cap \operatorname{supp}(g(y, \omega))=\emptyset$ we have

$$
\begin{aligned}
& \mathbb{E}\left[\left(P_{t}^{m} f(0, \omega)\right)^{2}\right]=\mathbb{E}\left[\left(\sum_{x \in \mathbb{Z}^{d}}\left(P_{0}^{m}\left(X_{t}=x-e_{i}\right)-P_{0}^{m}\left(X_{t}=x\right)\right) g(x, \omega)\right)^{2}\right] \\
& \leqslant \# \operatorname{supp}(g)^{2} \sum_{x \in \mathbb{Z}^{d}}\left(P_{0}^{m}\left(X_{t}=x\right)-P_{0}^{m}\left(X_{t}=x+e_{i}\right)\right)^{2} \mathbb{E}\left[g^{2}\right]
\end{aligned}
$$

which, combined with Corollary 3.3 gives the expected result.
3.3. Discussion about the polynomial decay. In the introduction, we told that $t^{-d / 2}$ is the optimal decay for local functions in $L^{2}$. Indeed, in [DMFGW89], they proved that, under the annealed law, the walker $X_{t}$ converges to a Brownian motion, suggesting a diffusive behaviour and this rate of decay as optimal. In [Mou11], it is suggested that, using spectral theory, we can construct a non-local function $f$ such that $\mathbb{E}\left[f_{t}^{2}\right]$ decays as fast or as slow as we want. We note here that one can construct a local function with a faster decay than $t^{-d / 2}$. Indeed, because $\mathbb{E}\left[f_{t}^{2}\right]$ is a completely monotonic
function, it is a consequence of corrolary 3.1. For example, consider a function $g \in L^{2}$ such that $f=\mathcal{L} g$, then

$$
\mathbb{E}\left[f_{t}^{2}\right]=\mathbb{E}\left[\left(\mathcal{L} g_{t}\right)^{2}\right]=\partial_{t} \partial_{t} \mathbb{E}\left[g_{t}^{2}\right] \leqslant C_{g} t^{-d / 2-2}
$$

Iterating the process, considering $f=\mathcal{L}^{n} g$, we have that $\mathbb{E}\left[f_{t}^{2}\right] \leqslant C_{g} t^{-d / 2-2 n}$. Note that in this case, $f$ might not be in $L^{2}$, even if $f_{t}$ would.

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