

An Eyring–Kramers law for the stochastic Allen–Cahn equation in dimension two

Nils Berglund* Giacomo Di Gesù† Hendrik Weber‡

Abstract

We study spectral Galerkin approximations of an Allen–Cahn equation over the two-dimensional torus perturbed by weak space-time white noise of strength $\sqrt{\varepsilon}$. We introduce a Wick renormalisation of the equation in order to have a system that is well-defined as the regularisation is removed. We show sharp upper and lower bounds on the transition times from a neighbourhood of the stable configuration -1 to the stable configuration 1 in the asymptotic regime $\varepsilon \rightarrow 0$. These estimates are uniform in the discretisation parameter N , suggesting an Eyring–Kramers formula for the limiting renormalised stochastic PDE. The effect of the “infinite renormalisation” is to modify the prefactor and to replace the ratio of determinants in the finite-dimensional Eyring–Kramers law by a renormalised Carleman–Fredholm determinant.

Keywords: Stochastic partial differential equations; metastability; Kramers’ law; renormalisation; potential theory; capacities; spectral Galerkin approximation; Wick calculus.

AMS MSC 2010: Primary 60H15; 35K57, Secondary 81S20; 82C28.

Submitted to EJP on April 27, 2016, final version accepted on April 23, 2017.

Supersedes arXiv:1604.05742v3.

Supersedes HAL:hal-01304559.

1 Introduction

Metastability is a common physical phenomenon in which a system spends a long time in metastable states before reaching its equilibrium. One of the most classical mathematical models where this phenomenon has been studied rigorously is the overdamped motion of a particle in a potential V , given by the Itô stochastic differential equation

$$dx(t) = -\nabla V(x(t)) dt + \sqrt{2\varepsilon} dw(t). \tag{1.1}$$

*MAPMO–CNRS UMR 7349, Fédération Denis Poisson FR 2964, Université d’Orléans, 45067 Orléans, France. E-mail: nils.berglund@univ-orleans.fr <http://www.univ-orleans.fr/mapmo/membres/berglund/>

†CERMICS – Ecole des Ponts ParisTech, F-77455 Marne-la-Vallée Cedex 2, France. Current address: TU Vienna, E101 - Institut für Analysis und Scientific Computing, 1040 Vienna, Austria. E-mail: giacomo.di.gesù@tuwien.ac.at <https://sites.google.com/site/giacomodigesu>

‡Mathematics Institute, University of Warwick, Coventry CV4 7AL, United Kingdom. E-mail: hendrik.weber@warwick.ac.uk https://www2.warwick.ac.uk/fac/sci/math/people/staff/hendrik_weber/

For small noise strength ε solutions of (1.1) typically spend long stretches of time near local minima of the potential V with occasional, relatively quick transitions between these local minima. The mean transition time between minima is then governed by the Eyring–Kramers law [12, 22]: If τ denotes the expected hitting time of a neighbourhood of a local minimum y of the solution of (1.1) started in another minimum x , and under suitable assumptions on the potential V , the Eyring–Kramers law gives the asymptotic expression

$$\mathbb{E}[\tau] = \frac{2\pi}{|\lambda_0(z)|} \sqrt{\frac{|\det D^2V(z)|}{\det D^2V(x)}} e^{[V(z)-V(x)]/\varepsilon} [1 + \mathcal{O}_\varepsilon(1)] , \tag{1.2}$$

where z is the *relevant saddle* between x and y , and $\lambda_0(z)$ is the (by assumption) unique negative eigenvalue of the Hessian $D^2V(z)$ (more precise bounds on the error term $\mathcal{O}_\varepsilon(1)$ are available). The right exponential rate in this formula was established rigorously using large deviation theory [15]. Rigorous proofs including the prefactor were given in [30, 8, 20], see [4] for a recent survey.

It is natural to study metastability in high- or infinite-dimensional systems and to seek an extension of the Eyring–Kramers law. In this direction the Allen–Cahn equation perturbed by a small noise term is a natural model to study. It is given by the stochastic PDE

$$\partial_t \phi(t, x) = \Delta \phi(t, x) - (\phi(t, x)^3 - \phi(t, x)) + \sqrt{2\varepsilon} \xi(t, x) , \tag{1.3}$$

where ξ is a noise term to be described below. Just like in (1.1) the deterministic part of this equation (set $\xi = 0$ in (1.3)) has gradient form and the relevant potential is given by

$$V(\phi) = \int \left(\frac{1}{2} |\nabla \phi|^2 - \frac{1}{2} \phi^2 + \frac{1}{4} \phi^4 \right) dx . \tag{1.4}$$

The natural choice of noise term ξ is (at least formally) given by space-time white noise because this choice is compatible with the scalar product used to define the deterministic gradient flow and it makes the dynamics given by (1.3) reversible (in the sense that the system satisfies detailed balance when in its statistical equilibrium state). The constant profiles $\phi_\pm(x) = \pm 1$ are stable solutions of the deterministic system and it is natural to ask how long a small noise term typically needs to move the system from one of these stable profiles to the other one.

In the case where equation (1.3) is solved over a $1 + 1$ -dimensional time-space domain $(t, x) \in [0, \infty) \times [0, L]$ this question was first studied in [13] on the level of large deviations, yielding the correct exponent in (1.2). The problem of obtaining sharper asymptotics with the correct prefactor was first considered in [3], and infinite-dimensional versions of the Eyring–Kramers formula were established in [2, 6]. Let τ denote the first-hitting time of a solution of (1.3) starting near the constant profile ϕ_- of a neighbourhood of the constant profile $\phi_+ = 1$. In [6] it was shown, for example in the case where (1.3) is endowed with periodic boundary conditions on $[0, L]$ and $L < 2\pi$, that

$$\mathbb{E}[\tau] = \frac{2\pi}{|\lambda_0|} \prod_{k \in \mathbb{Z}} \sqrt{\frac{|\lambda_k|}{\nu_k}} e^{[V(\phi_0) - V(\phi_-)]/\varepsilon} [1 + \mathcal{O}_\varepsilon(1)] , \tag{1.5}$$

where k plays the role of a wave number in a Fourier decomposition. The purpose of the condition $L < 2\pi$ is to ensure that the constant profile $\phi_0 = 0$ is the relevant saddle between the two stable minima ϕ_\pm ; but situations for longer intervals and different choices of boundary conditions are also described in [6]. The sequences λ_k, ν_k appearing in this expression are the eigenvalues of the Hessian of V around ϕ_0 and ϕ_- , the operators $-\partial_x^2 - 1$ and $-\partial_x^2 + 2$, both endowed with periodic boundary condition (the corresponding eigenfunctions being simply Fourier modes). All of these eigenvalues are

strictly positive, except for $\lambda_0 = -1$. Leaving out the factor $k = 0$, the infinite product in (1.5) can be written as

$$\prod_{k \neq 0} \sqrt{\frac{\lambda_k}{\nu_k}} = \prod_{k \neq 0} \sqrt{\left(1 + \frac{\nu_k - \lambda_k}{\lambda_k}\right)^{-1}} = \frac{1}{\sqrt{\det(\text{Id} + 3P_{\perp}(-\partial_x^2 - 1)^{-1})}}, \quad (1.6)$$

where P_{\perp} projects on the complement of the $k = 0$ mode and the operator $(-\partial_x^2 - 1)$ acts on zero-mean functions. This expression converges, because $P_{\perp}(-\partial_x^2 - 1)^{-1}$ is a trace-class operator, so that the infinite-dimensional (Fredholm) determinant is well-defined (see for instance [14, 23]).

When trying to extend this result to higher spatial dimensions two problems immediately present themselves. First, for spatial dimension $d \geq 2$ the Allen–Cahn equation as stated in (1.3) fails to be well-posed: in this situation already the linear stochastic heat equation (drop the non-linear term $-(\phi^3 - \phi)$ in (1.3)) has distribution-valued solutions due to the irregularity of the white noise ξ . In this regularity class $-(\phi^3 - \phi)$ does not have a canonical definition. As an illustration for the problems caused by this irregularity, it was shown in [18] that for fixed noise strength ε finite-dimensional spectral Galerkin approximations¹

$$\partial_t \phi_N = \Delta \phi_N - (P_N \phi_N^3 - \phi_N) + \sqrt{2\varepsilon} \xi_N \quad (1.7)$$

defined over a two-dimensional torus converge to a trivial limit as the approximation parameter N goes to ∞ (precise definitions of the finite-dimensional noise ξ_N and the projection operator P_N are given in Section 2 below). A second related problem is, that for $d \geq 2$ the infinite product appearing in (1.5) converges to 0, corresponding to the fact that for $d \geq 2$ the operator $3P_{\perp}(-\Delta - 1)^{-1}$ fails to be trace-class so that the Fredholm determinant $\det(\text{Id} + 3P_{\perp}(-\Delta - 1)^{-1})$ is not well-defined.

On the level of the $N \rightarrow \infty$ limit for fixed ε the idea of renormalisation, inspired by ideas from Quantum Field Theory (see e.g. [16]), has been very successful over the last years. Indeed, in [9] it was shown that in the two-dimensional case, if the approximations in (1.7) are replaced by

$$\partial_t \phi_N = \Delta \phi_N - (P_N \phi_N^3 - 3\varepsilon C_N \phi_N - \phi_N) + \sqrt{2\varepsilon} \xi_N, \quad (1.8)$$

for a particular choice of logarithmically divergent constants C_N (see (2.6) below), the solutions do converge to a non-trivial limit which can be interpreted as *renormalised solutions* of (1.3). This result (for a different choice of C_N) was spectacularly extended to three dimensions in Hairer’s pioneering work on regularity structures [17]. For spatial dimension $d \geq 4$, equation (1.3) fails to satisfy a *subcriticality condition* (see [17]) and non-trivial renormalised solutions are not expected to exist.

Note that formally the extra term $3\varepsilon C_N \phi_N$ moves the stable solutions further apart to $\pm\sqrt{3\varepsilon C_N + 1}$ (and ultimately to $\pm\infty$ as $N \rightarrow \infty$). Note furthermore that while the constants C_N diverge as N goes to ∞ , for fixed N they are multiplied with a small factor ε . This suggests that in the small-noise regime the renormalised solutions may still behave as perturbations of the Allen–Cahn equation, despite the presence of the infinite renormalisation constant. In [19] this intuition was confirmed on the level of large deviations. There it was shown that both in two and three dimensions the renormalised stochastic PDE satisfies a large-deviation principle as $\varepsilon \rightarrow 0$, with respect to a suitable topology and with rate functional given by

$$\mathcal{I}(\phi) = \frac{1}{2} \int_0^T \int (\partial_t \phi - (\Delta \phi - (\phi^3 - \phi)))^2 dx dt, \quad (1.9)$$

¹In fact, in [18, 9] the nonlinearity ϕ^3 is not projected onto a finite dimensional space, but this does not affect the result.

which is exactly the “correct” rate functional one would obtain by formally applying Freidlin–Wentzell theory to (1.3) without any regard to renormalisation. Results in a similar spirit had previously been obtained in [21, 1].

The purpose of this article is to show that the renormalised solutions have the right asymptotic small-noise behaviour even beyond large deviations, and to establish an Eyring–Kramers formula in this framework. As remarked also in [28] nothing seems to be known at this level so far. The key observation is that the introduction of the infinite constant not only permits to define the dynamics, but that it also fixes the problem of vanishing prefactor in the Eyring–Kramers law (1.5). More precisely, we argue that in two dimensions the correct Eyring–Kramers formula for the renormalised SPDE is

$$\mathbb{E}[\tau] = \frac{2\pi}{|\lambda_0|} \sqrt{\prod_{k \in \mathbb{Z}^2} \frac{|\lambda_k|}{\nu_k} \exp\left\{\frac{\nu_k - \lambda_k}{|\lambda_k|}\right\}} e^{[V(\phi_0) - V(\phi_-)]/\varepsilon} [1 + \mathcal{O}_\varepsilon(1)], \quad (1.10)$$

where as above the λ_k and ν_k are the eigenvalues of $-\Delta - 1$ and $-\Delta + 2$, now indexed by a vectorial wave number $k \in \mathbb{Z}^2$. In functional-analytic terms this means that due to the presence of the infinite renormalisation constant the regular determinant from (1.5) is replaced by a renormalised or Carleman–Fredholm determinant of the operator $\text{Id} + 3P_\perp(-\Delta - 1)^{-1}$. Unlike the “usual” determinant, the Carleman–Fredholm determinant is defined for the class of Hilbert–Schmidt perturbations of the identity and not only for the smaller class of trace-class perturbations of the identity. Recall, that $(-\Delta - 1)^{-1}$ is Hilbert–Schmidt both in two and three dimensions, but not for $d \geq 4$. It is striking to note that these are exactly the dimensions in which a renormalised solution to the Allen–Cahn (or Φ^4) equation can be constructed.

In order to illustrate our result in the easiest possible situation we only consider the case of the Allen–Cahn equation in a small domain $\mathbb{T}^2 = [0, L]^2$ of size $L < 2\pi$ with periodic boundary conditions. As in the one-dimensional case this assumption guarantees that the constant profile ϕ_0 is the relevant saddle. We make use of the ± 1 symmetry of the system to simplify some arguments. Throughout the article, we work in the framework of the finite-dimensional spectral Galerkin approximation (1.8) and derive asymptotic bounds for the expected transition time as $\varepsilon \rightarrow 0$ which are uniform in the approximation parameter $N \rightarrow \infty$.

On the technical level, our analysis builds on the potential-theoretic approach developed in [8], which allows to express expected transition times in terms of committor functions and capacities, that can be estimated using a variational principle. As we work in finite dimensions throughout, we can avoid making any use of the analytic tools developed in recent years to deal with singular SPDEs. A crucial idea is to change point of view with respect to the usual finite dimensional setting as presented in [7, 8] and to regard capacities and partition functions as expectations of random variables under Gaussian measures, which are well-defined in infinite dimension. This idea already appeared in [11] for the analysis of metastability of the Allen–Cahn equation in space dimension $d = 1$. In the present setting ($d = 2$) the new point of view is particularly powerful, since expectations under Gaussian measures can be estimated using Wick calculus, and in particular the classical Nelson argument [25] which permits to bound expectations of exponentials of Hermite polynomials. Another key argument is the observation from [5] that the field ϕ can be decomposed into its average and fluctuating part and that the (non-convex) potential V is convex in the transverse directions (see Lemma 5.3). An additional key idea, following [11, Sec. 3.2], is to avoid using Hausdorff–Young inequalities in the discussion of Laplace asymptotics (see [3, 6]) and rather use Taylor expansions and global estimates, which lead to much better error estimates, both in ε and in N . The rest of this paper is structured as follows: in Sections 2 and 3 we give

the precise assumptions, state our main theorem and give the necessary background from potential theory. Lower bounds on the expected transition time are proved in Section 4, upper bounds are proved in Section 5. Some well-known facts about Hermite polynomials and Wick powers are collected in Appendix A.

Acknowledgments. The authors gratefully acknowledge the hospitality and financial support they received during reciprocal visits at the University of Warwick and the Université d’Orléans. GDG gratefully acknowledges the support of the European Research Council under the European Union’s Seventh Framework Programme (FP/2007-2013) / ERC Grant Agreement number 614492. HW gratefully acknowledges support from the EPSRC through an EPSRC First Grant and the Royal Society through a University Research Fellowship. Finally, the authors would like to thank the two anonymous referees for their careful reading of the first version of this manuscript, which led to substantial improvements in the presentation.

2 Results

Let $\mathbb{T}^2 = \mathbb{R}^2 / (L\mathbb{Z})^2$ denote the two-dimensional torus of size $L \times L$. We are interested in the renormalised Allen–Cahn equations

$$\partial_t \phi = \Delta \phi + [1 + 3\varepsilon C_N] \phi - \phi^3 + \sqrt{2\varepsilon} \xi_N \tag{2.1}$$

for $\phi = \phi(t, x) : \mathbb{R}^+ \times \mathbb{T}^2 \rightarrow \mathbb{R}$, where ξ_N approximates space-time white noise on the scale $1/N$. In fact, we will consider spectral Galerkin approximations of the above equation. Let $e_k(x) = L^{-1} e^{i(2\pi/L)k \cdot x}$ denote L^2 -normalised Fourier basis vectors of $L^2(\mathbb{T}^2)$, where $k \in \mathbb{Z}^2$ is the wave vector. Denote by P_N the projection on Fourier modes with wave number k satisfying $|k| = \max\{|k_1|, |k_2|\} \leq N$, that is,

$$(P_N \phi)(x) = \sum_{k \in \mathbb{Z}^2: |k| \leq N} \langle \phi, e_k \rangle e_k(x). \tag{2.2}$$

Then we consider the sequence of equations

$$\partial_t \phi = \Delta \phi + [1 + 3\varepsilon C_N] \phi - P_N \phi^3 + \sqrt{2\varepsilon} \xi_N \tag{2.3}$$

where $\xi_N = P_N \xi$ is the spectral Galerkin approximation of space-time white noise ξ .

We will assume periodic boundary conditions (b.c.), with a domain size satisfying $0 < L < 2\pi$. This assumption guarantees that the identically zero function plays the role of the transition state, which separates the basins of attraction of the two stable solutions $\phi_{\pm} = \pm 1$ of the deterministic Allen–Cahn equation.

Note that the deterministic system (take $\varepsilon = 0$ in (2.3)) is a gradient system, with potential

$$V_N[\phi] = \frac{1}{2} \int_{\mathbb{T}^2} [\|\nabla \phi(x)\|^2 - \phi^2(x)] dx + \frac{1}{4} \int_{\mathbb{T}^2} [\phi^4(x) - 6\varepsilon C_N \phi^2(x) + 3\varepsilon^2 C_N^2] dx. \tag{2.4}$$

The measure $e^{-V_N/\varepsilon}$ is an invariant, reversible measure of the stochastic system (2.3), and we will denote by $Z_N(\varepsilon)$ its normalisation (the partition function of the system). The constant term $3\varepsilon^2 C_N^2$ in the second integral is of course irrelevant for the dynamics, but it will simplify notations. This is related to the fact that

$$\phi^4(x) - 6\varepsilon C_N \phi^2(x) + 3\varepsilon^2 C_N^2 = H_4(\phi(x), C_N) \tag{2.5}$$

is the so-called Wick renormalisation of $\phi^4(x)$ with respect to the centered Gaussian measure having covariance $|P_N[-\Delta - 1]^{-1}|$, often denoted $:\phi^4(x):$, where H_4 is the

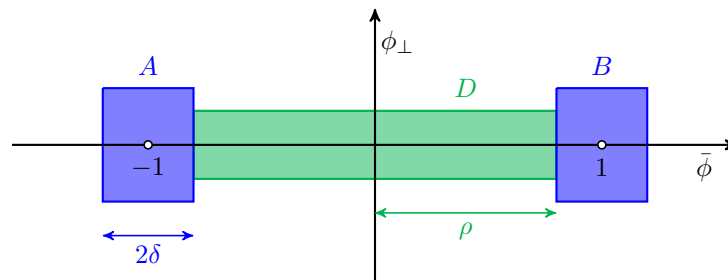


Figure 1: Geometry of the neighbourhoods A and B of the deterministic stable solutions $\phi_{\pm}(x)$. The set D will be needed later in the proof, cf. Section 5.3.

fourth Hermite polynomial (see Section 4 and Appendix A). The renormalisation constant C_N is given by

$$C_N = \frac{1}{L^2} \text{Tr}(|P_N[-\Delta - 1]^{-1}|) := \frac{1}{L^2} \sum_{k \in \mathbb{Z}^2: |k| \leq N} \frac{1}{|\lambda_k|} \tag{2.6}$$

where $\lambda_k = (2\pi/L)^2(k_1^2 + k_2^2) - 1$. Therefore, C_N diverges logarithmically as

$$C_N \asymp \frac{2\pi}{L^2} \log(N) . \tag{2.7}$$

The choice of C_N is somewhat arbitrary, as adding a constant independent of N to C_N will also yield a well-defined limit equation as $N \rightarrow \infty$. See Remark 2.5 below for the effect of such a shift on the results.

In the deterministic case $\varepsilon = 0$, the Allen–Cahn equation (2.3) has two stable stationary solutions given by $\phi_-(x) = -1$ and $\phi_+(x) = 1$. We are interested in obtaining sharp asymptotics on the expectation of the first-hitting time τ_B of a neighbourhood B of ϕ_+ , when starting near ϕ_- . The neighbourhood B should have a minimal size. More precisely, we decompose any function $\phi : \mathbb{T}^2 \rightarrow \mathbb{R}$ into its mean and oscillating part by setting

$$\phi(x) = \bar{\phi} + \phi_{\perp}(x) \tag{2.8}$$

where $\bar{\phi} = L^{-2} \int_{\mathbb{T}^2} \phi(x) dx$ and the integral of ϕ_{\perp} over \mathbb{T}^2 is zero. Then we define two symmetric sets A and B as follows (see Figure 1).

Definition 2.1. Fix constants $\delta \in (0, 1)$ and $s < 0$, and let D_{\perp} be a ball of radius $r\sqrt{\varepsilon \log(\varepsilon^{-1})}$ in the Sobolev space $H^s(\mathbb{T}^d)$, where r is a sufficiently large numerical constant. Then

$$\begin{aligned} A &= \{ \phi \in H^s(\mathbb{T}^d) : \bar{\phi} \in [-1 - \delta, -1 + \delta], \phi_{\perp} \in D_{\perp} \} , \\ B &= \{ \phi \in H^s(\mathbb{T}^d) : \bar{\phi} \in [1 - \delta, 1 + \delta], \phi_{\perp} \in D_{\perp} \} . \end{aligned} \tag{2.9}$$

Remark 2.2. The definition of D_{\perp} ensures that $A \cup B$ contains most of the mass of the invariant probability measure $\mathcal{Z}_N^{-1} e^{-V_N/\varepsilon}$ of the equation. In fact, the result remains true if we replace D_{\perp} by any sufficiently regular set containing D_{\perp} , see Lemma 5.9.

Our main result for periodic b.c. is as follows (recall from the introduction that $\nu_k = (2\pi/L)^2(k_1^2 + k_2^2) + 2 = \lambda_k + 3$).

Theorem 2.3. Assume $0 < L < 2\pi$. There exists a sequence $\{\mu_N\}_{N \geq 1}$ of probability

measures concentrated on ∂A such that in the case of periodic b.c.,

$$\begin{aligned} \limsup_{N \rightarrow \infty} \mathbb{E}^{\mu_N} [\tau_B] &\leq \frac{2\pi}{|\lambda_0|} \sqrt{\prod_{k \in \mathbb{Z}^2} \frac{|\lambda_k|}{\nu_k} \exp\left\{\frac{\nu_k - \lambda_k}{|\lambda_k|}\right\}} e^{[V(\phi_0) - V(\phi_-)]/\varepsilon} [1 + c_+ \sqrt{\varepsilon}] , \\ \liminf_{N \rightarrow \infty} \mathbb{E}^{\mu_N} [\tau_B] &\geq \frac{2\pi}{|\lambda_0|} \sqrt{\prod_{k \in \mathbb{Z}^2} \frac{|\lambda_k|}{\nu_k} \exp\left\{\frac{\nu_k - \lambda_k}{|\lambda_k|}\right\}} e^{[V(\phi_0) - V(\phi_-)]/\varepsilon} [1 - c_- \varepsilon] , \end{aligned} \quad (2.10)$$

where the constants c_{\pm} are uniform in ε .

Since $\nu_k = \lambda_k + 3$ and $V(\phi_0) - V(\phi_-) = L^2/4$, the leading term in (2.10) can also be written as

$$2\pi \left(\frac{e^{3/|\lambda_0|}}{|\lambda_0|(\lambda_0 + 3)} \prod_{k \neq 0} \frac{e^{3/\lambda_k}}{1 + 3/\lambda_k} \right)^{1/2} e^{L^2/4\varepsilon} . \quad (2.11)$$

The infinite product indeed converges, since for $x > 0$

$$0 < \log\left(\frac{e^x}{1+x}\right) = x - \log(1+x) \leq \frac{1}{2}x^2 , \quad (2.12)$$

and the sum over \mathbb{Z}^2 of λ_k^{-2} converges, unlike the sum of λ_k^{-1} that would arise without the regularising term e^{3/λ_k} . On a more abstract level, as already mentioned in the introduction, this is due to the fact that we have replaced the usual Fredholm determinant $\det(\text{Id} + T)$ (with $T = 3|(-\Delta - 1)^{-1}|$) by the Fredholm–Carleman determinant

$$\det_2(\text{Id} + T) = \det(\text{Id} + T) e^{-\text{Tr } T} , \quad (2.13)$$

which is defined for every Hilbert–Schmidt perturbation T of the identity, without the requirement of T to be trace-class [29, Chapter 5].

Remark 2.4. An analogous result holds for zero-flux Neumann b.c. $\partial_{x_i} \phi(t, x) = 0$ whenever $x_i \in \{0, L\}$ for $i = 1$ or 2 , provided $0 < L < \pi$. The only difference is that one has to replace $2\pi/L$ by π/L in the definition of the λ_k , and that the sums are over $k \in \mathbb{N}_0^2$ instead of $k \in \mathbb{Z}^2$.

Remark 2.5. The definition (2.6) of the renormalisation constant C_N is not unique, and one would still obtain a well-defined limit for (2.3) if C_N were replaced by $C_N + \theta/L^2$ for some constant $\theta \in \mathbb{R}$ (or even by $C_N + \theta_N/L^2$, where θ_N converges to a limit θ as $N \rightarrow \infty$). One easily checks that the effect of such a shift in the renormalisation constant is to multiply the expected transition time by a factor $e^{3\theta/2|\lambda_0|}$.

Remark 2.6. The measures μ_N appearing in Theorem 2.3 are the normalised equilibrium measures $\mu_{A,B}$ of the finite dimensional dynamics of ϕ_N . These measures appear in the result, because the method of proof is based on an asymptotic analysis as $\varepsilon \rightarrow 0$ of the exact formula (3.5) for $\mathbb{E}^{\mu_N} [\tau_B]$ which is available with this initial distribution. One would expect that the precise starting distribution does not affect the final result and that equation (2.10) remains true with μ_N replaced by any distribution which is supported on a sufficiently small neighbourhood of ϕ_- , in particular a deterministic initial condition at ϕ_- . For the one-dimensional SPDE such a result was indeed obtained in [6], by post-processing a bound involving the equilibrium measure, building on a coupling technique developed in [24]. It would be interesting to see if a similar post-processing is also possible in the context of our renormalised SPDE.

Remark 2.7. In a similar spirit, it might seem more natural to state the bound (2.10) in the limit $N \rightarrow \infty$, i.e. for solutions of the full renormalized SPDE rather than stating a uniform-in- N bound over the Galerkin approximations. In the one-dimensional case, such

a result was obtained in [6, Proposition 3.4], using various a priori estimates available for $d = 1$. Using the technique from [9] it is straightforward to show that the approximate solutions ϕ_N (say all starting in ϕ_-) and the corresponding stopping times τ_N converge in probability as $N \rightarrow \infty$. One can then pass to the limit in the expectations $\mathbb{E}^{\phi_-} [\tau_N]$ (for $N \rightarrow \infty$ keeping ε fixed) provided the random variables τ_N are uniformly integrable, which is implied, for example, by a bound of the type $\sup_N \mathbb{E}^{\phi_-} [\tau_N^p] < \infty$ for some $p > 1$. We expect that a much stronger bound of the type $\sup_N \mathbb{E}^{\phi_-} [\exp(\lambda \tau_N)] < \infty$ for λ small enough can be obtained, using the results derived in [31].

Remark 2.8. The error term in $\sqrt{\varepsilon}$ in the upper bound for the expected transition time is due to our using less sharp approximations in the Laplace asymptotics. It is in principle possible, as was done in the one-dimensional case in [11], to derive further terms in the asymptotic expansion. In particular it is expected that the leading error term has order ε .

3 Some potential theory

When considering the spectral Galerkin approximations (2.3), it will sometimes be convenient to use Fourier variables z_k defined by

$$\phi_N(t, x) = \sum_{k \in \mathbb{Z}^2: |k| \leq N} z_k(t) e_k(x). \tag{3.1}$$

In order to ensure that ϕ_N is real-valued, the coefficients z_k are chosen to take values in

$$\{(z_k) \in \mathbb{C}^{(-N, \dots, N)^2} : z_{-k} = \overline{z_k} \text{ for all } k\} \tag{3.2}$$

which we identify with $\mathbb{R}^{(2N+1)^2}$ throughout. In particular, we will always interpret gradients and integration with respect to Lebesgue measure dz in terms of this identification. The spectral Galerkin approximation with cut-off N is equivalent to the system of Itô SDEs

$$dz(t) = -\nabla V_N(z(t)) dt + \sqrt{2\varepsilon} dW_t, \tag{3.3}$$

where the potential, obtained by evaluating (2.4) in ϕ_N , is given by

$$V_N(z) = \frac{1}{2} \sum_{|k| \leq N} \lambda_k |z_k|^2 + \frac{1}{4} \left(\frac{1}{L^2} \sum_{\substack{k_1+k_2+k_3+k_4=0 \\ |k_i| \leq N}} z_{k_1} z_{k_2} z_{k_3} z_{k_4} - 6\varepsilon C_N \sum_{|k| \leq N} |z_k|^2 + 3L^2 \varepsilon^2 C_N^2 \right). \tag{3.4}$$

Arguments based on potential theory (see [8] and [7, Corollary 7.30]) show that for any finite N one has the relation

$$\mathbb{E}^{\mu_{A,B}} [\tau_B] = \frac{1}{\text{cap}_A(B)} \int_{B^c} h_{A,B}(z) e^{-V_N(z)/\varepsilon} dz. \tag{3.5}$$

Here $h_{A,B}(z)$ is the *committor function* (or *equilibrium potential*)

$$h_{A,B}(z) = \mathbb{P}^z \{ \tau_A < \tau_B \}, \tag{3.6}$$

where τ_A denotes the first-hitting time of a set $A \subset \mathbb{R}^{(2N+1)^2}$. The term $\text{cap}_A(B)$ is the so-called *capacity*, which admits several equivalent expressions:

$$\text{cap}_A(B) = \varepsilon \int_{(A \cup B)^c} \|\nabla h_{A,B}(z)\|^2 e^{-V_N(z)/\varepsilon} dz \tag{3.7}$$

$$= \varepsilon \inf_{h \in \mathcal{H}_{A,B}} \int_{(A \cup B)^c} \|\nabla h\|^2 e^{-V_N(z)/\varepsilon} dz \tag{3.8}$$

$$= \int_{\partial A} e^{-V_N(z)/\varepsilon} \rho_{A,B}(dz), \tag{3.9}$$

where $\mathcal{H}_{A,B}$ is the set of functions $h \in H^1$ such that $h = 1$ in A and $h = 0$ in B , and $\rho_{A,B}(dz)$ is a measure concentrated on ∂A , called the *equilibrium measure*. The density of this measure (with respect to the surface measure on ∂A) is equal to the exterior normal derivative of the committor function $h_{A,B}$. Finally, $\mu_{A,B}$ is the probability measure on ∂A obtained by normalising $\rho_{A,B} e^{-V/\varepsilon}$:

$$\mu_{A,B}(dz) = \frac{1}{\text{cap}_A(B)} e^{-V_N(z)/\varepsilon} \rho_{A,B}(dz) . \tag{3.10}$$

The following symmetry argument allows us to link the expected transition time to the partition function of the system.

Lemma 3.1. *If A and B are symmetric with respect to the plane $z_0 = 0$ then*

$$\int_{B^c} h_{A,B}(z) e^{-V_N(z)/\varepsilon} dz = \frac{1}{2} \int_{\mathbb{R}^{(2N+1)^2}} e^{-V_N(z)/\varepsilon} dz =: \frac{1}{2} \mathcal{Z}_N(\varepsilon) . \tag{3.11}$$

Proof. Consider the reflection S given by

$$S(z_0, z_\perp) = (-z_0, z_\perp) .$$

The potential V_N satisfies the symmetry

$$V_N(Sz) = V_N(z)$$

which implies

$$\int_{\{z_0 < 0\}} e^{-V_N(z)/\varepsilon} dz = \int_{\{z_0 > 0\}} e^{-V_N(z)/\varepsilon} dz = \frac{1}{2} \mathcal{Z}_N(\varepsilon) . \tag{3.12}$$

Assuming we choose A and B such that $B = SA$, the committor satisfies

$$h_{A,B}(z) = h_{B,A}(Sz) .$$

In addition, we always have

$$h_{A,B}(z) = 1 - h_{B,A}(z) .$$

Now observe that we have

$$\begin{aligned} & \int_{\mathbb{R}^{(2N+1)^2}} h_{A,B}(z) e^{-V_N(z)/\varepsilon} dz \\ &= \int_{\{z_0 < 0\}} h_{A,B}(z) e^{-V_N(z)/\varepsilon} dz + \int_{\{z_0 > 0\}} (1 - h_{B,A}(z)) e^{-V_N(z)/\varepsilon} dz \\ &= \int_{\{z_0 < 0\}} h_{A,B}(z) e^{-V_N(z)/\varepsilon} dz + \int_{\{z_0 > 0\}} e^{-V_N(z)/\varepsilon} dz - \int_{\{z_0 < 0\}} h_{A,B}(z) e^{-V_N(z)/\varepsilon} dz \\ &= \int_{\{z_0 > 0\}} e^{-V_N(z)/\varepsilon} dz , \end{aligned}$$

and the conclusion follows from (3.12). □

As a consequence of (3.11), we can rewrite (3.5) in the form

$$\mathbb{E}^{\mu_{A,B}} [\tau_B] = \frac{1}{2 \text{cap}_A(B)} \mathcal{Z}_N(\varepsilon) . \tag{3.13}$$

Note that this relation can also be written as

$$\frac{1}{\mathbb{E}^{\mu_{A,B}} [\tau_B]} = 2 \mathbb{E}^{\pi_N(\varepsilon)} [\varepsilon \|\nabla h_{A,B}\|^2] , \tag{3.14}$$

where $\pi_N(\varepsilon)$ is the probability measure on $\mathbb{R}^{(2N+1)^2}$ with density $\mathcal{Z}_N(\varepsilon)^{-1} e^{-V_N(z)/\varepsilon} dz$.

The result will follow if we are able to prove the estimate

$$\text{cap}_A(B) = \sqrt{\frac{|\lambda_0|\varepsilon}{2\pi}} \prod_{0 < |k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k}} [1 + r(\varepsilon)] \tag{3.15}$$

on the capacity with $-c_- \sqrt{\varepsilon} \leq r(\varepsilon) \leq c_+ \varepsilon$, as well as an estimate on the partition function of the form

$$\frac{1}{2} \mathcal{Z}_N(\varepsilon) = \prod_{|k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k + 3}} e^{L^2/4\varepsilon} e^{3L^2 C_N/2} [1 - r(\varepsilon)] . \tag{3.16}$$

Here the exponent $L^2/4$ is the value of the original potential (1.4) at the stable stationary solution ϕ_- , while the term $3L^2 C_N/2$ is due to the renormalisation, which makes the potential well deeper and deeper as $N \rightarrow \infty$. Note that owing to the expression (2.6) of C_N , this is exactly the extra term transforming the divergent Fredholm determinant into a converging Carleman–Fredholm determinant.

4 Lower bound on the expected transition time

We will start by deriving a lower bound on the expected transition time, i.e., we will prove an upper bound on the capacity given by (3.15) with $r(\varepsilon) \leq c_+ \varepsilon$ (in Section 4.1) and a lower bound on the partition function given by (3.16) with the same bound on $r(\varepsilon)$ (in Section 4.2). These bounds are somewhat simpler to obtain than the matching lower and upper bound that we will derive in Section 5.

For the upper bound on $\text{cap}_A(B)$ we choose a specific test-function h_+ (defined in (4.1)) and plug it into the variational characterisation (3.8) of the capacity. The specific form of h_+ is similar to the one used in [8]. It is given by (an approximation to) the explicit minimiser of the one-dimensional version of (3.8) in the z_0 direction and chosen to be constant in all transversal directions. Then an explicit calculation with this test function leads to an upper bound on the capacity which consists of the desired pre-factor multiplied by the Gaussian expectation of the exponential of a negative fourth Wick power (see equation (4.9) below). Here the renormalisation introduces a difficulty, because the Wick power is not bounded from below uniformly in N . In order to analyse this expression, we invoke techniques from constructive Quantum Field Theory, more precisely, the Nelson argument (see [25]). This is implemented in two steps in Proposition 4.1 and Proposition 4.2 leading to the desired bound.

In order to derive the lower bound on the $\mathcal{Z}_N(\varepsilon)$ we use the symmetry of the system to reduce to the integration over a half space. We then recentre the field around the minimiser 1 and perform the corresponding change of coordinates in the expression for V_N . At this point we use the binomial identity for Wick powers (see (A.3)) as well as the transformation rules for changing the “mass” in a Wick power (see (A.5)), resulting again in a Gaussian expectation of the exponential of a quartic Wick polynomial. This time the Gaussian covariance changes with respect to Section 4.1 and the Wick polynomial is defined with respect to this modified measure and also has a non-trivial cubic term. The “renormalisation” factor $e^{3L^2 C_N/2}$ in (3.16) appears at this point, because the change of Gaussian measure is given by a quadratic Wick power which produces an extra term in the Gaussian normalisation constant. Finally, the Gaussian expectation is treated with a relatively simple argument using Jensen’s inequality in Proposition 4.4.

4.1 Upper bound on the capacity

One can obtain an upper bound on the capacity by inserting any test function in the right-hand side of (3.8). Let $\delta > 0$ be a small constant and define

$$h_+(z) = \begin{cases} 1 & \text{if } z_0 \leq -\delta, \\ \frac{\int_{z_0}^{\delta} e^{-|\lambda_0|t^2/2\varepsilon} dt}{\int_{-\delta}^{\delta} e^{-|\lambda_0|t^2/2\varepsilon} dt} & \text{if } -\delta < z_0 < \delta, \\ 0 & \text{if } z_0 \geq \delta. \end{cases} \tag{4.1}$$

Although $|\lambda_0| = 1$, we will keep λ_0 in the notation as it allows to keep track of its influence on the result. Observe that

$$\|\nabla h_+(z)\|^2 = \begin{cases} \frac{e^{-|\lambda_0|z_0^2/\varepsilon}}{\left(\int_{-\delta}^{\delta} e^{-|\lambda_0|t^2/2\varepsilon} dt\right)^2} & \text{if } -\delta < z_0 < \delta, \\ 0 & \text{otherwise.} \end{cases} \tag{4.2}$$

Standard asymptotics of Gaussian integrals yields

$$\left(\int_{-\delta}^{\delta} e^{-|\lambda_0|t^2/2\varepsilon} dt\right)^2 = \frac{2\pi\varepsilon}{|\lambda_0|} [1 + \mathcal{O}(e^{-\delta^2/2\varepsilon})], \tag{4.3}$$

where here and throughout the paper, the notation $f(\varepsilon) = \mathcal{O}(g(\varepsilon))$ indicates that there exists $\varepsilon_0 > 0$ such that $|f(\varepsilon)|$ is bounded by a constant times $|g(\varepsilon)|$ uniformly in $\varepsilon \in (0, \varepsilon_0)$. Note that here the parameter δ is small but fixed. Inserting (4.3) in (3.8) we get

$$\begin{aligned} \text{cap}_A(B) &\leq \frac{|\lambda_0|}{2\pi} \int_{\mathbb{R}^{(2N+1)^2}} e^{-[V_N(z)+|\lambda_0|z_0^2]/\varepsilon} dz [1 + \mathcal{O}(e^{-\delta^2/2\varepsilon})] \\ &= \frac{|\lambda_0|}{2\pi} \varepsilon^{\frac{1}{2}(2N+1)^2} \int_{\mathbb{R}^{(2N+1)^2}} e^{-[V_N(\sqrt{\varepsilon}y)+\varepsilon|\lambda_0|y_0^2]/\varepsilon} dy [1 + \mathcal{O}(e^{-\delta^2/2\varepsilon})], \end{aligned} \tag{4.4}$$

where we used the scaling $z = \sqrt{\varepsilon}y$, which amounts to working with the scaled field

$$\phi_N = \sqrt{\varepsilon}\hat{\phi}_N. \tag{4.5}$$

Inserting this scaling in (3.4) shows that the exponent in (4.4) can be written in the form

$$\frac{1}{\varepsilon} [V_N(\sqrt{\varepsilon}y) + \varepsilon|\lambda_0|y_0^2] = g_N(y) + \varepsilon w_N(y), \tag{4.6}$$

where

$$\begin{aligned} g_N(y) &= \frac{1}{2}|\lambda_0|y_0^2 + \frac{1}{2} \sum_{0 < |k| \leq N} \lambda_k |y_k|^2, \\ w_N(y) &= \frac{1}{4} \int_{\mathbb{T}^2} (\hat{\phi}_N^4(x) - 6C_N \hat{\phi}_N^2(x) + 3C_N^2) dx. \end{aligned} \tag{4.7}$$

The quadratic form g_N allows us to define a Gaussian probability measure $d\gamma(y) = \mathcal{N}^{-1} e^{-g_N(y)} dy$, with normalisation

$$\mathcal{N} = \sqrt{\frac{2\pi}{|\lambda_0|}} \prod_{k \neq 0, |k| \leq N} \sqrt{\frac{2\pi}{\lambda_k}}. \tag{4.8}$$

We can thus rewrite the upper bound (4.4) in the form

$$\text{cap}_A(B) \leq \sqrt{\frac{|\lambda_0|\varepsilon}{2\pi}} \prod_{k \neq 0, |k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k}} \mathbb{E}^\gamma [e^{-\varepsilon w_N}] [1 + \mathcal{O}(e^{-\delta^2/2\varepsilon})]. \quad (4.9)$$

The term $\mathbb{E}^\gamma [e^{-\varepsilon w_N}]$ can be estimated using the Gaussian calculus developed in Appendix A. Indeed, the law of the field $\hat{\phi}_N$ under γ is exactly as described there. Furthermore, $C_N = \mathbb{E}^\gamma [\hat{\phi}_N(x)^2]$ for each $x \in \mathbb{T}^2$ so that the term $w_N(y)$, defined in (4.7), can be rewritten as the integral over the fourth Wick power of $\hat{\phi}$ with respect to this Gaussian measure, that is,

$$w_N = \frac{1}{4} \int_{\mathbb{T}^2} H_4(\hat{\phi}_N, C_N) dx, \quad (4.10)$$

(where $H_4(X, C) = X^4 - 6CX^2 + 3C^2$, see (A.1) for the definition of the Hermite polynomial H_n). In particular w_N has zero mean under the Gaussian measure $d\gamma$ and according to (A.13) all its stochastic moments are uniformly bounded in N .

We now derive a uniform-in- N bound on $\mathbb{E}^\gamma [e^{-w_N}]$ following a classical argument due to Nelson (see e.g. [16, Sec. 8.6] or [10, Sec. 4]).

Proposition 4.1. *There exists a constant K , independent of N , such that*

$$\mathbb{E}^\gamma [e^{-w_N}] \leq K. \quad (4.11)$$

Proof. First note that the definition (A.2) of Hermite polynomials implies for any $M \in \mathbb{N}$

$$H_4(\hat{\phi}_M, C_M) = (\hat{\phi}_M^2(x) - 3C_M)^2 - 6C_M^2,$$

so that

$$w_M \geq -\frac{3}{2}L^2C_M^2 =: -D_M.$$

Since $\mathbb{E}^\gamma [e^{-w_N} 1_{\{w_N \geq 0\}}] \leq \mathbb{P}^\gamma \{w_N \geq 0\} \leq 1$, it is sufficient to bound

$$\begin{aligned} \mathbb{E}^\gamma [e^{-w_N} 1_{\{w_N < 0\}}] &= 1 + \int_0^\infty e^t \mathbb{P}^\gamma \{-w_N > t\} dt \\ &\leq e + \int_1^\infty e^t \mathbb{P}^\gamma \{-w_N > t\} dt. \end{aligned}$$

If $t \geq D_N$, then $\mathbb{P}^\gamma \{-w_N > t\} = 0$, otherwise we have for any M

$$\begin{aligned} \mathbb{P}^\gamma \{-w_N > t\} &\leq \mathbb{P}^\gamma \{w_M - w_N > t - D_M\} \\ &\leq \mathbb{P}^\gamma \{|w_M - w_N|^{p(t)} > |t - D_M|^{p(t)}\}, \end{aligned}$$

for any choice of $p(t) \in 2\mathbb{N}$. We apply this inequality for $M = M(t)$ satisfying

$$t - D_{M(t)} \geq 1, \quad (4.12)$$

which implies that $M(t) < N$.

Then we get by Markov's inequality and Lemma A.4 combined with (A.11) (cf. (A.14))

$$\begin{aligned} \mathbb{P}^\gamma \{-w_N > t\} &\leq \mathbb{E}^\gamma [|w_{M(t)} - w_N|^{p(t)}] \\ &\leq \text{const} (p(t) - 1)^{p(t)} \mathbb{E}^\gamma [|w_{M(t)} - w_N|^2]^{p(t)/2} \\ &\leq \text{const} \frac{(p(t) - 1)^{p(t)}}{M(t)^{(1-\eta)p(t)/2}} \end{aligned}$$

for any $\eta > 0$. The condition (4.12) on $M(t)$ imposes that $\log M(t)$ grows at most as $t^{1/2}$. Choosing for instance $p(t) \sim t^\beta$ for $\beta > \frac{1}{2}$, since

$$\log(e^t \mathbb{P}^\gamma\{-w_N > t\}) \leq t + \beta t^\beta \log t - c(1 - \eta)t^{\beta+1/2}$$

for a constant $c > 0$ depending only on L , we obtain a convergent integral. □

This a priori estimate can now quite easily be turned into a sharper estimate. Indeed, we have the following improvement of (4.11).

Proposition 4.2. *We have*

$$\mathbb{E}^\gamma[e^{-\varepsilon w_N}] = 1 + \mathcal{O}(\varepsilon), \tag{4.13}$$

where the remainder is bounded uniformly in N .

Proof. Introduce the sets

$$\Omega_+ = \{\hat{\phi}_N : w_N > 0\}$$

and $\Omega_- = \Omega_+^c$. Then we have

$$\mathbb{E}^\gamma[e^{-\varepsilon w_N} 1_{\Omega_+}] = \mathbb{P}^\gamma(\Omega_+) + \mathbb{E}^\gamma[(e^{-\varepsilon w_N} - 1)1_{\Omega_+}].$$

Note that on Ω_+ , we have $-\varepsilon|w_N| \leq e^{-\varepsilon w_N} - 1 \leq 0$, so that

$$\mathbb{P}^\gamma(\Omega_+) - \varepsilon \mathbb{E}^\gamma[|w_N|1_{\Omega_+}] \leq \mathbb{E}^\gamma[e^{-\varepsilon w_N} 1_{\Omega_+}] \leq \mathbb{P}^\gamma(\Omega_+).$$

Since w_N has finite variance bounded uniformly in N , we know by Cauchy–Schwarz that $\mathbb{E}^\gamma[|w_N|]$ is bounded uniformly in N . Similarly, we have

$$\mathbb{E}^\gamma[e^{-\varepsilon w_N} 1_{\Omega_-}] = \mathbb{P}^\gamma(\Omega_-) + \mathbb{E}^\gamma[(e^{-\varepsilon w_N} - 1)1_{\Omega_-}].$$

This time, we use that on Ω_- , one has $0 \leq e^{-\varepsilon w_N} - 1 \leq \varepsilon|w_N|e^{-\varepsilon w_N}$. Thus by Cauchy–Schwarz,

$$\begin{aligned} 0 \leq \mathbb{E}^\gamma[(e^{-\varepsilon w_N} - 1)1_{\Omega_-}] &\leq \varepsilon \mathbb{E}^\gamma[|w_N|e^{-\varepsilon w_N} 1_{\Omega_-}] \\ &\leq \varepsilon \left(\mathbb{E}^\gamma[e^{-2\varepsilon w_N} 1_{\Omega_-}] \mathbb{E}^\gamma[|w_N|^2 1_{\Omega_-}] \right)^{1/2}. \end{aligned}$$

The term $\mathbb{E}^\gamma[|w_N|^2]$ is bounded uniformly in N as before, while the term $\mathbb{E}^\gamma[e^{-2\varepsilon w_N}]$ is bounded uniformly in N for $\varepsilon \leq 2$ by Proposition 4.1. Summing the two estimates, we get the result. □

Substituting this estimate in (4.9), we immediately get the following upper bound on the capacity.

Corollary 4.3. *There exists a constant c_+ , uniform in ε and N , such that the capacity satisfies the upper bound*

$$\text{cap}_A(B) \leq \sqrt{\frac{|\lambda_0|\varepsilon}{2\pi}} \prod_{k \neq 0, |k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k}} [1 + c_+\varepsilon]. \tag{4.14}$$

4.2 Lower bound on the partition function

By symmetry, cf. (3.12), the partition function can be computed using the relation

$$\frac{1}{2} \mathcal{Z}_N(\varepsilon) = \int_{\Omega'_+} e^{-V_N(z)/\varepsilon} dz, \quad \Omega'_+ = \{z_0 > 0\}. \tag{4.15}$$

A lower bound on $\mathcal{Z}_N(\varepsilon)$ can be obtained quite directly from Jensen’s inequality. It will be convenient to shift coordinates to the positive stable stationary solution of the deterministic equation (without the normalisation). That is, we set

$$\phi_N(x) = 1 + \sqrt{\varepsilon} \hat{\phi}_{N,+}(x), \tag{4.16}$$

with the Fourier decomposition

$$\hat{\phi}_{N,+}(x) = \sum_{|k| \leq N} y_k e_k(x). \tag{4.17}$$

Let $\hat{\Omega}'_+ = \{y_0 > -1/\sqrt{\varepsilon}\}$ denote the image of Ω'_+ under this transformation. Substituting in (2.4) and using the relation (A.3) yields the following expression for the potential:

$$\begin{aligned} V_N^+(y) &:= \frac{1}{\varepsilon} V_N[1 + \sqrt{\varepsilon} \hat{\phi}_{N,+}(x)] \\ &= -\frac{L^2}{4\varepsilon} + \frac{1}{2} \int_{\mathbb{T}^2} \left(\|\nabla \hat{\phi}_{N,+}(x)\|^2 - \hat{\phi}_{N,+}^2(x) + 3H_2(\hat{\phi}_{N,+}, C_N) \right) dx \\ &\quad + \frac{1}{4} \int_{\mathbb{T}^2} \left(4\sqrt{\varepsilon} H_3(\hat{\phi}_{N,+}(x), C_N) + \varepsilon H_4(\hat{\phi}_{N,+}(x), C_N) \right) dx. \end{aligned} \tag{4.18}$$

Now the relevant Gaussian measure γ_+ is defined by the quadratic form

$$\begin{aligned} g_{N,+}(y) &= \frac{1}{2} \int_{\mathbb{T}^2} \left(\|\nabla \hat{\phi}_{N,+}(x)\|^2 - \hat{\phi}_{N,+}^2(x) + 3\hat{\phi}_{N,+}^2(x) \right) dx \\ &= \frac{1}{2} \sum_{0 < |k| \leq N} (\lambda_k + 3) |y_k|^2. \end{aligned} \tag{4.19}$$

Observe that a term $-\frac{3}{2} C_N L^2$ appears owing to the Hermite polynomial $3H_2(\hat{\phi}_{N,+}, C_N)$. It is precisely this term which is ultimately responsible for the renormalisation of the pre-factor. To bound expectations of the terms appearing in the last line of (4.18) it is convenient to rewrite them as Wick powers with respect to the Gaussian measure defined by $g_{N,+}$. The associated renormalisation constant is

$$C_{N,+} = \frac{1}{L^2} \sum_{0 < |k| \leq N} \frac{1}{\lambda_k + 3}. \tag{4.20}$$

Observe in particular that

$$C_N - C_{N,+} = \frac{1}{L^2} \sum_{0 < |k| \leq N} \frac{3}{|\lambda_k|(\lambda_k + 3)} \tag{4.21}$$

is bounded uniformly in N . Using the relation (A.5) that allows to transform Hermite polynomials with respect to different constants we get

$$\begin{aligned} \sqrt{\varepsilon} H_3(\hat{\phi}_{N,+}, C_N) &= \sqrt{\varepsilon} H_3(\hat{\phi}_{N,+}, C_{N,+}) - 3\sqrt{\varepsilon} (C_N - C_{N,+}) H_1(\hat{\phi}_{N,+}, C_{N,+}) \\ \frac{\varepsilon}{4} H_4(\hat{\phi}_{N,+}, C_N) &= \frac{\varepsilon}{4} H_4(\hat{\phi}_{N,+}, C_{N,+}) - \frac{3}{2} \varepsilon (C_N - C_{N,+}) H_2(\hat{\phi}_{N,+}, C_{N,+}) \\ &\quad + \frac{3}{4} \varepsilon (C_N - C_{N,+})^2. \end{aligned} \tag{4.22}$$

Now we define the random variables

$$U_{n,N}^+ = \int_{\mathbb{T}^2} : \hat{\phi}_{N,+}^n(x) : dx = \int_{\mathbb{T}^2} H_n(\hat{\phi}_{N,+}(x), C_{N,+}) dx \quad (4.23)$$

which have zero mean under γ_+ as well as a variance bounded uniformly in N . Substituting (4.22) in (4.18), we get

$$V_N^+(y) = q + g_{N,+}(y) + w_{N,+}(y), \quad (4.24)$$

where

$$q = -\frac{L^2}{4\varepsilon} - \frac{3}{2}L^2C_N + \frac{3}{4}L^2\varepsilon(C_N - C_{N,+})^2$$

$$w_{N,+}(y) = \sqrt{\varepsilon}U_{3,N}^+ + \frac{1}{4}\varepsilon U_{4,N}^+ - 3(C_N - C_{N,+})\left(\frac{\varepsilon}{2}U_{2,N}^+ + \sqrt{\varepsilon}U_{1,N}^+\right). \quad (4.25)$$

It follows by a similar argument as in the previous section that

$$\frac{1}{2}\mathcal{Z}_N(\varepsilon) = \prod_{|k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k + 3}} e^{-q} \mathbb{E}^{\gamma_+} [e^{-w_{N,+}} 1_{\hat{\Omega}'_+}]. \quad (4.26)$$

Proposition 4.4. *There exists a constant c_- , independent of N and ε , such that*

$$\mathbb{E}^{\gamma_+} [e^{-w_{N,+}} 1_{\hat{\Omega}'_+}] \geq 1 - e^{-c_-/\varepsilon}. \quad (4.27)$$

Proof. Recall that $w_{N,+}$ has zero expectation under γ_+ . Jensen’s inequality yields

$$\begin{aligned} \mathbb{E}^{\gamma_+} [e^{-w_{N,+}} 1_{\hat{\Omega}'_+}] &= \mathbb{P}^{\gamma_+}(\hat{\Omega}'_+) \mathbb{E}^{\gamma_+} [e^{-w_{N,+}} \mid \hat{\Omega}'_+] \\ &\geq \mathbb{P}^{\gamma_+}(\hat{\Omega}'_+) e^{-\mathbb{E}^{\gamma_+}[w_{N,+} \mid \hat{\Omega}'_+]} \\ &= \mathbb{P}^{\gamma_+}(\hat{\Omega}'_+) e^{-\mathbb{E}^{\gamma_+}[w_{N,+} 1_{\hat{\Omega}'_+}] / \mathbb{P}^{\gamma_+}(\hat{\Omega}'_+)}. \end{aligned}$$

Note that the event $\hat{\Omega}'_+$ simply says that the first marginal of the Gaussian distribution γ_+ is larger than a constant. Since this marginal is a one-dimensional Gaussian distribution, centred at the positive stationary solution, standard tail estimates show that there is a constant $c_0 > 0$ such that

$$\mathbb{P}^{\gamma_+}(\hat{\Omega}'_+) \geq 1 - e^{-c_0/\varepsilon}.$$

Furthermore, there is a constant K such that uniformly in N and for $n = 1, 2, 3, 4$,

$$\begin{aligned} |\mathbb{E}^{\gamma_+} [U_{n,N}^+ 1_{\hat{\Omega}'_+}]| &= |\mathbb{E}^{\gamma_+} [U_{n,N}^+ 1_{(\hat{\Omega}'_+)^c}]| \leq \mathbb{E}^{\gamma_+} [|U_{n,N}^+| 1_{(\hat{\Omega}'_+)^c}] \\ &\leq \mathbb{E}^{\gamma_+} [(U_{n,N}^+)^2]^{1/2} \mathbb{P}^{\gamma_+}((\hat{\Omega}'_+)^c)^{1/2} \leq K e^{-c_0/2\varepsilon}. \end{aligned}$$

It thus follows that there exists a constant $c_1 > 0$ such that

$$\mathbb{E}^{\gamma_+} [w_{N,+} 1_{\hat{\Omega}'_+}] \geq -c_1 e^{-c_0/2\varepsilon},$$

which yields the required estimate $\mathbb{E}^{\gamma_+} [e^{-w_{N,+}} 1_{\hat{\Omega}'_+}] \geq 1 - e^{-c_-/\varepsilon}$. □

Combining this result with (4.26) and Corollary 4.3, we finally obtain the following lower bound on the expected transition times.

Proposition 4.5. *There exists a constant C_- , uniform in N and ε , such that*

$$\mathbb{E}^{\mu_{A,B}} [\tau_B] \geq 2\pi \left(\frac{e^{3/|\lambda_0|}}{|\lambda_0|(\lambda_0 + 3)} \prod_{0 < |k| \leq N} \left[\frac{e^{3/\lambda_k}}{1 + 3/\lambda_k} \right] \right)^{1/2} e^{L^2/4\varepsilon} [1 - C_- \varepsilon] \quad (4.28)$$

holds for all $N \geq 1$.

Proof. Plugging (4.27) into (4.26), using the upper bound (4.14) on the capacity and substituting in (3.13), we obtain

$$\mathbb{E}^{\mu_{A,B}}[\tau_B] \geq \sqrt{\frac{2\pi}{|\lambda_0|\varepsilon}} \sqrt{\frac{2\pi\varepsilon}{\lambda_0 + 3}} \prod_{0 < |k| \leq N} \sqrt{\frac{\lambda_k}{\lambda_k + 3}} e^{L^2/4\varepsilon} e^{3L^2 C_N/2} [1 + \mathcal{O}(\varepsilon)].$$

Using the fact that

$$\frac{3}{2}L^2 C_N = \frac{3}{2} \left(\frac{1}{|\lambda_0|} + \sum_{0 < |k| \leq N} \frac{1}{\lambda_k} \right)$$

yields the result. □

5 Upper bound on the expected transition time

In this section we derive the upper bound on the expected transition time, i.e., we will prove an upper bound on the partition function given by (3.16) with $r(\varepsilon) \geq -c_- \sqrt{\varepsilon}$ (in Section 5.2) and a lower bound on the capacity given by (3.15) with the same bound on $r(\varepsilon)$ (in Section 5.3).

Inspired by [5], we decompose the field ϕ into its mean (given by the zeroth Fourier coefficient z_0) and its (rescaled) transversal fluctuations, see (5.1). In fact, in Proposition 3.2 in [5] it was observed that in a similar system the potential V_N could be bounded from below by a function which only depends on the mean and a uniformly convex function in the transversal direction. In Lemma 5.3 we obtain a similar bound in our setting (but we only state it as a z_0 -dependent lower bound, which is all we need). The point here is that although this lower bound diverges logarithmically as $N \rightarrow \infty$, it does not become worse as $\varepsilon \rightarrow 0$ and thus permits to mimic the Nelson argument in Proposition 5.4 to obtain a bound on the integral over the transversal directions in the partition function, which does not depend too badly on z_0 and ε . Once this a priori bound is established we rewrite this transversal integral once more, this time with respect to a Gaussian reference measure $g_{N,\perp,z_0}(y_\perp)$ (in the terminology of Quantum Field Theory this amounts to a z_0 -dependent change of “mass”) to obtain a sharp upper bound on the integral over the transversal directions (in Proposition 5.7).

The argument for the lower bound on the capacity is similar to [8]. Using the characterisation (3.8) of $\text{cap}_A(B)$ the lower bound can be obtained by solving a one-dimensional variational problem in the z_0 direction (see (5.26) below). It is here where the assumption that the sets A and B are not too small enters.

Finally, in Section 5.4 it remains to treat the integral in the z_0 direction. Combining the bounds of the previous two sections one obtains an upper bound on the ratio $\frac{\mathcal{Z}_N(\varepsilon)}{2 \text{cap}_A(B)}$ in terms of an integral in z_0 over a function which depends on z_0 but not on N . This then permits to apply standard one-dimensional Laplace asymptotics to conclude the argument.

5.1 Longitudinal-transversal decomposition of the potential

We denote the rescaled fluctuating part of the Fourier expansion (3.1) by

$$\hat{\phi}_{N,\perp}(x) = \hat{\phi}_N(x) - \frac{z_0}{\sqrt{\varepsilon}L} = \sum_{0 < |k| \leq N} y_k e_k(x). \tag{5.1}$$

Note in particular the Parseval identity

$$\int_{\mathbb{T}^2} \hat{\phi}_{N,\perp}^2(x) dx = \sum_{0 < |k| \leq N} |y_k|^2. \tag{5.2}$$

Similarly to (4.18) the potential can be written in the form

$$\begin{aligned} \frac{1}{\varepsilon} V_N(y_\perp) &= \frac{1}{\varepsilon} q(z_0) + g_{N,\perp}(y_\perp) + \frac{1}{4} \int_{\mathbb{T}^2} \frac{6z_0^2}{L^2} H_2(\hat{\phi}_{N,\perp}(x), C_N) dx \\ &\quad + \frac{1}{4} \int_{\mathbb{T}^2} \left(\frac{4z_0}{L} \sqrt{\varepsilon} H_3(\hat{\phi}_{N,\perp}(x), C_N) + \varepsilon H_4(\hat{\phi}_{N,\perp}(x), C_N) \right) dx, \end{aligned} \quad (5.3)$$

where this time

$$\begin{aligned} q(z_0) &= \frac{1}{4L^2} z_0^4 - \frac{1}{2} |\lambda_0| z_0^2, \\ g_{N,\perp}(y_\perp) &= \frac{1}{2} \sum_{0 < |k| \leq N} \lambda_k |y_k|^2. \end{aligned} \quad (5.4)$$

Here we have used the fact that by assumption $\int_{\mathbb{T}^2} \hat{\phi}_{N,\perp}(x) dx = 0$, so that the corresponding term drops. The quadratic form $g_{N,\perp}$ defines a Gaussian measure γ_0^\perp with normalisation

$$\mathcal{N}_0^\perp = \prod_{0 < |k| \leq N} \sqrt{\frac{2\pi}{\lambda_k}}. \quad (5.5)$$

The associated renormalisation constant is given by

$$\mathbb{E}^{\gamma_0^\perp} [\hat{\phi}_{N,\perp}^2(x)] = \frac{1}{L^2} \sum_{0 < |k| \leq N} \frac{1}{\lambda_k} =: C_N^\perp = C_N - \frac{1}{L^2}. \quad (5.6)$$

As before, we are interested in the Wick powers $H_n(\hat{\phi}_{N,\perp}, C_N^\perp)$ with respect to this measure and set

$$U_{n,N}^\perp = \int_{\mathbb{T}^2} H_n(\hat{\phi}_{N,\perp}, C_N^\perp) dx, \quad U_n^\perp = \lim_{N \rightarrow \infty} U_{n,N}^\perp. \quad (5.7)$$

By construction, these random variables have (under the measure γ_0^\perp) zero mean and a variance bounded uniformly in N . Furthermore we see that

$$\int_{\mathbb{T}^2} H_3(\hat{\phi}_{N,\perp}(x), C_N) dx = U_{3,N}^\perp, \quad (5.8)$$

owing to the fact that $\hat{\phi}_{N,\perp}$ has zero mean, and

$$\int_{\mathbb{T}^2} H_4(\hat{\phi}_{N,\perp}(x), C_N) dx = U_{4,N}^\perp - \frac{6}{L^2} U_{2,N}^\perp + \frac{3}{L^2}. \quad (5.9)$$

The following expression for the potential then follows immediately from (5.3) and (A.5).

Proposition 5.1. *The potential can be decomposed as*

$$\frac{1}{\varepsilon} V_N(z_0, y_\perp) = \frac{1}{\varepsilon} q(z_0) + q_1(z_0, \varepsilon) + g_{N,\perp}(y_\perp) + w_{N,\perp}(z_0, y_\perp), \quad (5.10)$$

where $q(z_0)$ and $g_{N,\perp}(y_\perp)$ are given in (5.4), and

$$\begin{aligned} q_1(z_0, \varepsilon) &= -\frac{3z_0^2}{2L^2} + \frac{3\varepsilon}{4L^2}, \\ w_{N,\perp}(z_0, y_\perp) &= \frac{3(z_0^2 - \varepsilon)}{2L^2} U_{2,N}^\perp + \frac{z_0}{L} \sqrt{\varepsilon} U_{3,N}^\perp + \frac{1}{4} \varepsilon U_{4,N}^\perp. \end{aligned} \quad (5.11)$$

5.2 Upper bound on the partition function

In order to obtain an upper bound on $\mathcal{Z}_N(\varepsilon)$, we will first perform the integration over the fluctuating modes y_\perp , and then the integration over the mean value z_0 . The basic observation is the following rewriting of $\mathcal{Z}_N(\varepsilon)$.

Proposition 5.2. *The partition function is given by the integral*

$$\mathcal{Z}_N(\varepsilon) = \int_{-\infty}^{\infty} e^{-q(z_0)/\varepsilon} g(z_0, \varepsilon) dz_0, \tag{5.12}$$

where

$$g(z_0, \varepsilon) = e^{-q_1(z_0, \varepsilon)} \prod_{0 < |k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k}} \mathbb{E} \gamma_0^\perp [e^{-w_{N,\perp}(z_0, \cdot)}]. \tag{5.13}$$

By standard, one-dimensional Laplace asymptotics, we expect the integral (5.12) to be close to $2\sqrt{\pi\varepsilon} e^{L^2/4\varepsilon} g(L, \varepsilon)$. As g depends strongly on N , one has to use some care when performing the Laplace asymptotics. The solution to this difficulty is to not carry out the Laplace asymptotics directly for $\mathcal{Z}_N(\varepsilon)$, but instead for the ratio $\mathcal{Z}_N(\varepsilon)/\text{cap}_A(B)$, see Section 5.4 below. Our aim is thus to bound the expectation in (5.13). In order to apply a Nelson estimate, we will need a lower bound on $w_{N,\perp}(z_0, y_\perp)$. In fact, for later use (see Proposition 5.7), we will derive a lower bound for the slightly more general quantity

$$w_{N,\perp}^{(\mu)}(z_0, y_\perp) = \frac{3z_0^2}{2L^2} U_{2,N}^\perp + \mu \left(-\frac{3\varepsilon}{2L^2} U_{2,N}^\perp + \frac{z_0}{L} \sqrt{\varepsilon} U_{3,N}^\perp + \frac{1}{4} \varepsilon U_{4,N}^\perp \right), \tag{5.14}$$

where μ is a real parameter. Note in particular that $w_{N,\perp}^{(1)}(z_0, y_\perp) = w_{N,\perp}(z_0, y_\perp)$. The proof of the following simple but useful lower bound is inspired by Proposition 3.2 in [5].

Lemma 5.3. *For any $N \in \mathbb{N}$, $z_0 \in \mathbb{R}$ and $\mu \in (0, \frac{3}{2})$,*

$$w_{N,\perp}^{(\mu)}(z_0, y_\perp) \geq -D_N(z_0, \mu, \varepsilon), \tag{5.15}$$

where

$$D_N(z_0, \mu, \varepsilon) = \frac{3}{2} z_0^2 C_N + \frac{3}{4} \mu \varepsilon C_N^2 L^2 \left(\frac{3}{1 - 2\mu/3} - 1 \right). \tag{5.16}$$

Proof. Using the definition (5.7) of Wick powers, we see that

$$\begin{aligned} w_{N,\perp}^{(\mu)}(z_0, y_\perp) &= \frac{1}{4} \int_{\mathbb{T}^2} \hat{\phi}_{N,\perp}^2(x) \left[\mu \varepsilon \hat{\phi}_{N,\perp}^2(x) + 4\mu \frac{z_0}{L} \sqrt{\varepsilon} \hat{\phi}_{N,\perp}(x) + 6 \frac{z_0^2}{L^2} - 6\mu \varepsilon C_N \right] dx \\ &\quad - \frac{3}{2} (z_0^2 - \mu \varepsilon) C_N^\perp + \frac{3}{4} \mu \varepsilon (C_N^\perp)^2 L^2, \end{aligned} \tag{5.17}$$

where we have used the fact that $C_N^\perp + \frac{1}{L^2} = C_N$ as well as $\int_{\mathbb{T}^2} \hat{\phi}_{N,\perp}(x) dx = 0$. A completion-of-squares argument shows that the term in square brackets in (5.17) is bounded below by

$$\mu \varepsilon \left(1 - \frac{2}{3} \mu \right) \hat{\phi}_{N,\perp}^2(x) - 6\mu \varepsilon C_N.$$

Performing a second completion of squares shows that the integral in (5.17) is bounded below by

$$\frac{\mu \varepsilon}{4} \int_{\mathbb{T}^2} \left[\hat{\phi}_{N,\perp}^4(x) \left(1 - \frac{2}{3} \mu \right) - 6C_N \hat{\phi}_{N,\perp}^2(x) \right] dx \geq -\frac{9\mu \varepsilon C_N^2 L^2}{4(1 - 2\mu/3)}.$$

The result follows, bounding C_N^\perp above by C_N . □

We are now in a position to imitate the proof of Proposition 4.1, to show the following upper bound.

Proposition 5.4. *There exist constants $M(\mu)$ and $\varepsilon_0(\mu)$, uniform in N , ε and z_0 , such that*

$$\mathbb{E}^{\gamma_0^\perp} \left[e^{-w_{N,\perp}^{(\mu)}(z_0, \cdot)} \right] \leq M(\mu) \left[1 + \sqrt{\varepsilon} e^{M(\mu)z_0^2 \log(1+z_0^2)/\sqrt{\varepsilon}} \right] \quad (5.18)$$

holds for any $\mu \in (0, \frac{3}{2})$ and all $\varepsilon < \varepsilon_0(\mu)$.

Proof. We will give the proof for $w_\perp^{(\mu)} = \lim_{N \rightarrow \infty} w_{N,\perp}^{(\mu)}(z_0, \cdot)$, since the same proof applies for any finite N . To be able to apply the integration-by-parts formula

$$\begin{aligned} \mathbb{E}^{\gamma_0^\perp} \left[e^{-w_\perp^{(\mu)}} \right] &\leq e + \int_1^\infty e^t \mathbb{P}^{\gamma_0^\perp} \{ -w_\perp^{(\mu)} > t \} dt \\ &= e \left[1 + \int_0^\infty e^t \mathbb{P}^{\gamma_0^\perp} \{ -w_\perp^{(\mu)} > 1 + t \} dt \right], \end{aligned}$$

we have to estimate $\mathbb{P}^{\gamma_0^\perp} \{ -w_\perp^{(\mu)} > t \}$ when $t \geq 1$. For any such t , we pick an $N(t) \in \mathbb{N}$ such that

$$t - D_{N(t)}(z_0, \mu, \varepsilon) \geq 1.$$

Note that by (5.16), there exists a constant $M_0(\mu)$, uniform in N , ε and z_0 , such that

$$D_N(z_0, \mu, \varepsilon) \leq M_0(\mu) \left[\varepsilon (\log N)^2 + z_0^2 \log N \right].$$

The condition on $N(t)$ is thus satisfied if we impose the condition

$$\varepsilon (\log N(t))^2 + z_0^2 \log N(t) \leq \frac{t-1}{M_0(\mu)}. \quad (5.19)$$

By Lemma 5.3 and the above condition, we have

$$\mathbb{P}^{\gamma_0^\perp} \{ -w_\perp^{(\mu)} > t \} \leq \mathbb{P}^{\gamma_0^\perp} \{ |w_{N(t),\perp}^{(\mu)} - w_\perp^{(\mu)}| > 1 \}.$$

Now observe (c.f. (5.14)) that

$$w_{N(t),\perp}^{(\mu)} - w_\perp^{(\mu)} = \sum_{j=2}^4 a_j(z_0, \varepsilon) (U_{j,N(t)}^\perp - U_j^\perp),$$

with

$$a_2 = \frac{3}{2L^2} (z_0^2 - \mu\varepsilon), \quad a_3 = \frac{z_0}{L} \sqrt{\varepsilon}, \quad a_4 = \frac{1}{4} \varepsilon.$$

It follows that

$$\mathbb{P}^{\gamma_0^\perp} \{ |w_{N(t),\perp}^{(\mu)} - w_\perp^{(\mu)}| > 1 \} \leq \sum_{j=2}^4 P_j, \quad P_j = \mathbb{P}^{\gamma_0^\perp} \left\{ |a_j| |U_{j,N(t)}^\perp - U_j^\perp| > \frac{1}{3} \right\}.$$

For any choice of $p_j(t) \in 2\mathbb{N}$, we have by Markov's inequality

$$P_j \leq |3a_j(z_0, \varepsilon)|^{p_j(t)} E_j, \quad E_j = \mathbb{E}^{\gamma_0^\perp} \left[|U_{j,N(t)}^\perp - U_j^\perp|^{p_j(t)} \right]$$

where by Nelson's estimate (A.12)

$$E_j \leq \text{const} (p_j(t) - 1)^{jp_j(t)/2} \mathbb{E}^{\gamma_0^\perp} \left[|U_{j,N(t)}^\perp - U_j^\perp|^2 \right]^{p_j(t)/2} \leq \text{const} \frac{(p_j(t) - 1)^{jp_j(t)/2}}{N(t)^{p_j(t)}}.$$

A possible choice is to take (where $a \wedge b := \min\{a, b\}$ and $[a]$ is the integer part of a)

$$p_j(t) = 2 \left[(t-1)^{1/2} \wedge \frac{z_0^2}{\sqrt{\varepsilon}} \right]$$

$$\log N(t) = \frac{1}{M_1} \left[\left(\frac{t-1}{\varepsilon} \right)^{1/2} \wedge \frac{t-1}{z_0^2} \right],$$

with M_1 large enough to satisfy (5.19). Indeed, this yields $\log(N(t)^{p_j(t)}) \simeq c(t-1)/\sqrt{\varepsilon}$ for some $c = c(\mu) > 0$, and thus

$$\mathbb{P}^{\gamma_0^\perp} \{-w_\perp^{(\mu)} > t\} \leq \text{const } e^{c' \log(1+z_0^2)z_0^2/\sqrt{\varepsilon}} e^{-c(t-1)/\sqrt{\varepsilon}}$$

for some $c' > 0$, where the first exponential is due to the term $|3a_2|^{p_2(t)}$. This shows that

$$\int_0^\infty e^t \mathbb{P}^{\gamma_0^\perp} \{-w_\perp^{(\mu)} > 1+t\} dt \leq \text{const } e^{c' \log(1+z_0^2)z_0^2/\sqrt{\varepsilon}} \left(\frac{c}{\sqrt{\varepsilon}} - 1 \right)^{-1}$$

if $\varepsilon < c^2$. Substituting in the integration-by-parts formula proves the claim. □

Our aim is now to sharpen this bound by applying a similar trick as in the proof of Proposition 4.2. To this end, it will be convenient to work with Gaussian measures $\gamma_{z_0}^\perp$, defined by the quadratic form

$$g_{N,\perp,z_0}(y_\perp) = \sum_{0 < |k| \leq N} \left[\lambda_k + \frac{3z_0^2}{L^2} \right] |y_k|^2. \tag{5.20}$$

The following result allows converting between expectations with respect to γ_0^\perp and $\gamma_{z_0}^\perp$.

Lemma 5.5. *For any random variable $X = X(y_\perp)$ integrable with respect to γ_0^\perp ,*

$$\mathbb{E}^{\gamma_0^\perp} [X] = K(z_0) \mathbb{E}^{\gamma_{z_0}^\perp} \left[X e^{3z_0^2 U_{2,N}^\perp / 2L^2} \right], \tag{5.21}$$

where

$$K(z_0) = \left[\prod_{0 < |k| \leq N} \frac{e^{3z_0^2/L^2 \lambda_k}}{1 + 3z_0^2/L^2 \lambda_k} \right]^{1/2}.$$

Proof. This follows from a short computation, writing out explicitly the density of γ_0^\perp and expressing $\sum_k |y_k|^2$ in terms of $U_{2,N}^\perp$. □

Remark 5.6. Writing $\zeta_k = 3z_0^2/L^2 \lambda_k$ and using the fact that the Taylor series of $\log(1+\zeta_k)$ is alternating, we obtain

$$2 \log K(z_0) = \sum_{0 < |k| \leq N} [\zeta_k - \log(1 + \zeta_k)] \leq \frac{1}{2} \sum_{0 < |k| \leq N} \zeta_k^2 \leq \text{const } z_0^4. \tag{5.22}$$

This shows that $K(z_0) \leq e^{M_1 z_0^4}$ for some constant M_1 , independent of N and z_0 .

We can now state the sharper bound on the expectation of $e^{-w_{N,\perp}}$.

Proposition 5.7. *There exists a constant $M > 0$, uniform in N, ε and z_0 , such that*

$$\mathbb{E}^{\gamma_0^\perp} \left[e^{-w_{N,\perp}(z_0, \cdot)} \right] \leq K(z_0) \left[1 + M\sqrt{\varepsilon}(1 + |z_0|) \left(1 + \sqrt{\varepsilon} e^{Mz_0^2 \log(1+z_0^2)/\sqrt{\varepsilon}} \right) \right]. \tag{5.23}$$

Proof. To lighten the notation, we drop the argument (z_0, \cdot) of $w_{N,\perp}$. By Lemma 5.5, we have

$$\mathbb{E}^{\gamma_0^\perp} [e^{-w_{N,\perp}}] = K(z_0) \mathbb{E}^{\gamma_{z_0}^\perp} [e^{-\hat{w}_{N,\perp}}]$$

where

$$\hat{w}_{N,\perp} = -\frac{3\varepsilon}{2L^2} U_{2,N}^\perp + \frac{z_0}{L} \sqrt{\varepsilon} U_{3,N}^\perp + \frac{1}{4} \varepsilon U_{4,N}^\perp.$$

As in the proof of Proposition 4.2, we write

$$\begin{aligned} \mathbb{E}^{\gamma_{z_0}^\perp} [e^{-\hat{w}_{N,\perp}}] &\leq 1 + \mathbb{E}^{\gamma_{z_0}^\perp} [(e^{-\hat{w}_{N,\perp}} - 1) \mathbf{1}_{\{\hat{w}_{N,\perp} < 0\}}] \\ &\leq 1 + \mathbb{E}^{\gamma_{z_0}^\perp} [|\hat{w}_{N,\perp}| e^{-\hat{w}_{N,\perp}} \mathbf{1}_{\{\hat{w}_{N,\perp} < 0\}}] \\ &\leq 1 + \mathbb{E}^{\gamma_{z_0}^\perp} [(\hat{w}_{N,\perp})^p]^{1/p} \mathbb{E}^{\gamma_{z_0}^\perp} [e^{-q\hat{w}_{N,\perp}}]^{1/q}. \end{aligned}$$

In the last line, we have used Hölder’s inequality, and $p, q \geq 1$ are Hölder conjugates. It follows from (A.13) that

$$\mathbb{E}^{\gamma_{z_0}^\perp} [(\hat{w}_{N,\perp})^p]^{1/p} \leq \text{const} \sqrt{\varepsilon} (1 + |z_0|).$$

Furthermore, another application of Lemma 5.5 yields

$$\begin{aligned} \mathbb{E}^{\gamma_{z_0}^\perp} [e^{-q\hat{w}_{N,\perp}}] &= \frac{1}{K(z_0)} \mathbb{E}^{\gamma_0^\perp} [e^{-q\hat{w}_{N,\perp} - 3z_0^2 U_{2,N}^\perp / 2L^2}] \\ &= \frac{1}{K(z_0)} \mathbb{E}^{\gamma_0^\perp} [e^{-w_{N,\perp}^{(q)}}]. \end{aligned}$$

Applying Proposition 5.4 for some $q \in (1, \frac{3}{2})$ and combining the different estimates yields the result. \square

5.3 Lower bound on the capacity

Recall that $A = -I \times A_\perp$ and $B = I \times A_\perp$ where $\pm I = \pm[1 - \delta, 1 + \delta]$ (with $0 < \delta < 1$) are small intervals around the two stationary solutions. Let $D = J \times D_\perp$ where $J = [-\rho, \rho]$ is an interval joining $-I$ and I , with $\rho = 1 - \delta$, and $D_\perp \subset A_\perp$ (see Figure 1 – we set $\hat{\rho} = \rho L$ since in Fourier variables, coordinates are scaled by a factor L). Then we have

$$\begin{aligned} \text{cap}_A(B) &\geq \varepsilon \int_D \left| \frac{\partial h_{A,B}(z)}{\partial z_0} \right|^2 e^{-V_N(z)/\varepsilon} dz \\ &\geq \varepsilon \int_{D_\perp} \left[\inf_{f: f(-\hat{\rho})=1, f(\hat{\rho})=0} \int_{-\hat{\rho}}^{\hat{\rho}} e^{-V_N(z_0, z_\perp)/\varepsilon} f'(z_0)^2 dz_0 \right] dz_\perp. \end{aligned} \tag{5.24}$$

Writing the Euler–Lagrange equations, it is easy to see that the minimiser for the term in brackets is such that

$$f'(z_0) = \frac{e^{V_N(z_0, z_\perp)/\varepsilon}}{\int_{-\hat{\rho}}^{\hat{\rho}} e^{V_N(y, z_\perp)/\varepsilon} dy}. \tag{5.25}$$

This yields the lower bound

$$\text{cap}_A(B) \geq \varepsilon \int_{D_\perp} \frac{1}{\int_{-\hat{\rho}}^{\hat{\rho}} e^{V_N(z_0, z_\perp)/\varepsilon} dz_0} dz_\perp. \tag{5.26}$$

Proposition 5.8. *There exists a constant $c_- > 0$, uniform in ε and N , such that*

$$\text{cap}_A(B) \geq \sqrt{\frac{\varepsilon |\lambda_0|}{2\pi}} \prod_{0 < |k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k}} [\mathbb{P}^{\gamma_0^\perp}(\hat{D}_\perp) - c_- \sqrt{\varepsilon}], \tag{5.27}$$

where $\hat{D}_\perp = D_\perp / \sqrt{\varepsilon}$.

Proof. We start by obtaining a lower bound on V_N in which z_0 is decoupled from the transverse coordinates. Using the expression (5.10) obtained in Proposition 5.1 and the elementary inequality $2|ab| \leq (a^2/c + cb^2)$ for $c > 0$, we obtain

$$\frac{1}{\varepsilon} V_N(z_0, y_\perp) \leq \frac{1}{\varepsilon} q(z_0) + q_1(z_0, \varepsilon) + \frac{z_0^2}{2L^2} + \frac{3z_0^4}{4L^2\sqrt{\varepsilon}} + g_{N,\perp}(y_\perp) + \sqrt{\varepsilon} R(y_\perp, \varepsilon),$$

where

$$R(y_\perp, \varepsilon) = \frac{3}{4L^2} (U_{2,N}^\perp)^2 + \frac{1}{2} \sqrt{\varepsilon} (U_{3,N}^\perp)^2 - \frac{3\sqrt{\varepsilon}}{2L^2} U_{2,N}^\perp + \frac{1}{4} \sqrt{\varepsilon} U_{4,N}^\perp.$$

Substituting in (5.26) (and taking into account the scaling $z_\perp = \sqrt{\varepsilon} y_\perp$) yields

$$\text{cap}_A(B) \geq \frac{\varepsilon}{\mathcal{J}} \prod_{0 < |k| \leq N} \sqrt{\frac{2\pi\varepsilon}{\lambda_k}} \mathbb{E}^{\gamma_0^\perp} [e^{-\sqrt{\varepsilon}R} 1_{\hat{D}_\perp}],$$

where

$$\mathcal{J} = \int_{-\hat{\rho}}^{\hat{\rho}} \exp\left\{ \frac{1}{\varepsilon} q(z_0) + q_1(z_0, \varepsilon) + \frac{z_0^2}{2L^2} + \frac{3z_0^4}{4L^2\sqrt{\varepsilon}} \right\} dz_0.$$

Since $q(z_0)$ has a quadratic maximum on $[-\hat{\rho}, \hat{\rho}]$ at 0, standard one-dimensional Laplace asymptotics (see for instance [27, Chapter 3, Thms. 7.1 and 8.1]) show that

$$\mathcal{J} = \sqrt{\frac{2\pi\varepsilon}{|\lambda_0|}} [1 + \mathcal{O}(\sqrt{\varepsilon})].$$

Furthermore, Jensen’s inequality implies that

$$\begin{aligned} \mathbb{E}^{\gamma_0^\perp} [e^{-\sqrt{\varepsilon}R} 1_{\hat{D}_\perp}] &= \mathbb{E}^{\gamma_0^\perp} [e^{-\sqrt{\varepsilon}R} | \hat{D}_\perp] \mathbb{P}^{\gamma_0^\perp}(\hat{D}_\perp^c) \\ &\geq e^{-\sqrt{\varepsilon} \mathbb{E}^{\gamma_0^\perp} [R | \hat{D}_\perp]} \mathbb{P}^{\gamma_0^\perp}(\hat{D}_\perp^c) \\ &\geq \left(1 - \sqrt{\varepsilon} \frac{\mathbb{E}^{\gamma_0^\perp} [|R|]}{\mathbb{P}^{\gamma_0^\perp}(\hat{D}_\perp^c)} \right) \mathbb{P}^{\gamma_0^\perp}(\hat{D}_\perp^c). \end{aligned}$$

Since $\mathbb{E}^{\gamma_0^\perp} [|R|]$ is bounded uniformly by (A.13), the result follows. \square

The lower bound on the capacity is thus complete, provided we take D_\perp large enough to capture almost all the mass of γ_0^\perp . A possible choice is as follows.

Lemma 5.9. *Assume*

$$D_\perp \supset \prod_{0 < |k| \leq N} [-a_k, a_k] \quad \text{with } a_k = \sqrt{\frac{4\varepsilon \log(\varepsilon^{-1}) [1 + \log \lambda_k]}{\lambda_k}}. \quad (5.28)$$

Then for sufficiently small ε , one has

$$\mathbb{P}^{\gamma_0^\perp}(\hat{D}_\perp^c) = \mathcal{O}(\varepsilon). \quad (5.29)$$

Proof. Standard Gaussian tail estimates show that

$$\begin{aligned} \mathbb{P}^{\gamma_0^\perp}(\hat{D}_\perp^c) &\leq \sum_{0 < |k| \leq N} 2 e^{-a_k^2 \lambda_k / 2} \\ &= 2 \sum_{0 < |k| \leq N} \exp\{-2 \log(\varepsilon^{-1}) [1 + \log \lambda_k]\} \\ &\leq 2\varepsilon^2 \sum_{0 < |k| \leq N} \lambda_k^{-2 \log(\varepsilon^{-1})}. \end{aligned}$$

The last sum is bounded uniformly in N if $\varepsilon \leq e^{-1}$. \square

Note that if D_\perp is a ball in a Sobolev space as specified in Definition 2.1, then it indeed satisfies (5.28).

5.4 Laplace asymptotics and transition times

Combining the results from the last two subsections, we finally obtain the following upper bound on the expected transition time.

Proposition 5.10. *There exists a constant C_+ , uniform in N and ε , such that*

$$\mathbb{E}^{\mu_{A,B}}[\tau_B] \leq 2\pi \left(\frac{e^{3/|\lambda_0|}}{|\lambda_0|(\lambda_0 + 3)} \prod_{0 < |k| \leq N} \left[\frac{e^{3/\lambda_k}}{1 + 3/\lambda_k} \right] \right)^{1/2} e^{L^2/4\varepsilon} [1 + C_+ \sqrt{\varepsilon}] \tag{5.30}$$

holds for all $N \geq 1$.

Proof. It follows from Proposition 5.2, Proposition 5.7 and Proposition 5.8 that

$$\frac{\mathcal{Z}_N(\varepsilon)}{2 \operatorname{cap}_A(B)} \leq \sqrt{\frac{2\pi}{\varepsilon|\lambda_0|}} \int_0^\infty e^{-q(z_0)/\varepsilon} \hat{g}(z_0, \varepsilon) dz_0,$$

where

$$\hat{g}(z_0, \varepsilon) \leq e^{-q_1(z_0, \varepsilon)} K(z_0) \left[1 + M' \sqrt{\varepsilon} (1 + |z_0|) (1 + \sqrt{\varepsilon} e^{Mz_0^2 \log(1+z_0^2)/\sqrt{\varepsilon}}) \right]$$

for a constant $M' \geq M$. In particular, we have

$$\hat{g}(L, 0) \leq e^{3/2} K(L) = \left(e^{3/|\lambda_0|} \prod_{0 < |k| \leq N} \left[\frac{e^{3/\lambda_k}}{1 + 3/\lambda_k} \right] \right)^{1/2}.$$

Since q reaches its minimum $-L^2/4$ on \mathbb{R}_+ in $z_0 = L$, writing

$$\frac{1}{\sqrt{\varepsilon}} \int_0^\infty e^{-[q(z_0) - q(L)]/\varepsilon} \hat{g}(z_0, \varepsilon) dz_0 \leq \mathcal{I}_0 + \sqrt{\varepsilon} \mathcal{I}_1 + \varepsilon \mathcal{I}_2$$

and applying one-dimensional Laplace asymptotics, we obtain

$$\mathcal{I}_0 \leq \sqrt{\pi} \hat{g}(L, 0) (1 + C_+ \sqrt{\varepsilon}),$$

for the leading term, while \mathcal{I}_1 and \mathcal{I}_2 are bounded. □

A Hermite polynomials and Wick powers

In this appendix we recall some well-known facts about Hermite polynomials $H_n = H_n(X, C)$ we use throughout the article. Recall that they are defined recursively by setting

$$\begin{cases} H_0 = 1, \\ H_n = XH_{n-1} - C \partial_X H_{n-1} \end{cases} \quad n \in \mathbb{N}. \tag{A.1}$$

In particular, we have

$$\begin{aligned} H_1(X, C) &= X & H_2(X, C) &= X^2 - C \\ H_3(X, C) &= X^3 - 3CX & H_4(X, C) &= X^4 - 6CX^2 + 3C^2. \end{aligned} \tag{A.2}$$

The following binomial identity for Hermite polynomials is well-known.

Lemma A.1 (Binomial formula for Hermite polynomials, [9, Lem. 3.1]). *We have for any $n \in \mathbb{N}$ and $X, v, C \in \mathbb{R}$*

$$H_n(X + v, C) = \sum_{k=0}^n \binom{n}{k} H_{n-k}(X, C) v^k. \tag{A.3}$$

We will mostly be interested in Hermite polynomials of centred Gaussian random variables X and we will typically choose $C = \mathbb{E}[X^2]$. In this case the random variable $H_n(X, \mathbb{E}[X^2])$ is sometimes referred to as the n -th Wick power of X and denoted by $: X^n :$. The following identity is one of the key properties of Wick powers.

Lemma A.2 ([26, Lem. 1.1.1]). *Let X, Y be centred jointly Gaussian random variables. Then*

$$\mathbb{E}[: X^n :](: Y^m :) = \begin{cases} \mathbb{E}[XY]^n & \text{if } n = m \\ 0 & \text{else .} \end{cases} \tag{A.4}$$

Note that this implies in particular, that for $n \geq 1$ we have $\mathbb{E}[: X^n :] = 0$. In some computations it is convenient for us to change the value of the constant C appearing in H_n . This will be relevant when changing the Gaussian reference measure. The following transformation rule, valid for any real X, C, \bar{C} is easy to check:

$$\begin{aligned} H_1(X, C) &= H_1(X, \bar{C}) \\ H_2(X, C) &= H_2(X, \bar{C}) - (C - \bar{C}) \\ H_3(X, C) &= H_3(X, \bar{C}) - 3(C - \bar{C})H_1(X, \bar{C}) \\ H_4(X, C) &= H_4(X, \bar{C}) - 6(C - \bar{C})H_2(X, \bar{C}) + 3(C - \bar{C})^2 . \end{aligned} \tag{A.5}$$

We now use (A.4) to derive some classical facts about (spectral Galerkin approximations of) the two-dimensional massive Gaussian free field and its Wick powers. For any N and for $x \in \mathbb{T}^2 = \mathbb{R}^2 / (L\mathbb{Z})^2$ we consider the random field

$$\phi_N(x) = \sum_{|k| \leq N} \frac{z_k}{\sqrt{|\lambda_k + m^2|}} e_k(x) , \tag{A.6}$$

where the Fourier basis functions $e_k(x)$ and the eigenvalues λ_k of $(-\Delta + 1)$ are defined as in Section 2. The z_k are complex-valued Gaussian random variables which are independent up to the constraint $z_k = \bar{z}_{-k}$ which makes ϕ_N a real-valued field and which satisfy

$$\mathbb{E}[z_k z_{-\ell}] = \begin{cases} 1 & \text{if } k = \ell \\ 0 & \text{else .} \end{cases} \tag{A.7}$$

Note that due to the constraint $z_k = \bar{z}_{-k}$ these $(2N + 1)^2$ dependent complex-valued Gaussian random variables can be represented in terms of $(2N + 1)^2$ independent real-valued random variables. The mass $m^2 \geq 0$ is a parameter of the model, which in our case only takes either the value 0 or the value 3.

For fixed x we get

$$\mathbb{E}[\phi_N(x)^2] = \sum_{|k| \leq N} \frac{1}{|\lambda_k + m^2|} =: C_N . \tag{A.8}$$

Note that C_N diverges logarithmically, which suggests that the random variables $\phi_N(x)$ for a fixed x do not converge to a meaningful limit as N goes to ∞ . However, it is well-known that for any test-function ψ the random variables $\int \phi_N(x) \psi(x) dx$ converge in L^2 (with respect to probability) to a Gaussian limiting random variable. We will not make use of this general fact, but only use that the integrals of $\phi_N(x)$ as well as its Wick powers $: \phi_N^n(x) : = H_n(\phi_N(x), C_N)$ have a uniformly-in- N bounded variance. To see this

we write for $M > N$

$$\begin{aligned} & \mathbb{E} \left[\int_{\mathbb{T}^2} : \phi_M^n(x) : dx \int_{\mathbb{T}^2} : \phi_N^n(y) : dy \right] \\ &= \int_{\mathbb{T}^2} \int_{\mathbb{T}^2} \mathbb{E} [(: \phi_M^n(x) :) (: \phi_N^n(y) :)] dx dy = \int_{\mathbb{T}^2} \int_{\mathbb{T}^2} \mathbb{E} [\phi_M(x) \phi_N(y)]^n dx dy \\ &= \int_{\mathbb{T}^2} \int_{\mathbb{T}^2} \left(\frac{1}{L} \sum_{|k| \leq N} \frac{e_k(x-y)}{|\lambda_k + m^2|} \right)^n dx dy = L^{2-2n} \int_{\mathbb{T}^2} \left(\sum_{|k| \leq N} \frac{e_k(x)}{|\lambda_k + m^2|} \right)^n dx \\ &= L^{2-2n} \sum_{\substack{k_1+k_2+\dots+k_n=0 \\ |k_i| \leq N}} \frac{1}{|\lambda_{k_1} + m^2|} \cdots \frac{1}{|\lambda_{k_n} + m^2|}. \end{aligned} \tag{A.9}$$

This calculation has several immediate consequences. First of all we can conclude that as announced above the variances of $\int_{\mathbb{T}^2} : \phi_N^n(x) : dx$ are uniformly bounded as N goes to ∞ :

$$\sup_N \mathbb{E} \left[\left(\int_{\mathbb{T}^2} : \phi_N^n(x) : dx \right)^2 \right] = L^{4-2n} \sum_{\substack{k_1+k_2+\dots+k_n=0 \\ k_i \in \mathbb{Z}^2}} \frac{1}{|\lambda_{k_1} + m^2|} \cdots \frac{1}{|\lambda_{k_n} + m^2|} < \infty. \tag{A.10}$$

Indeed, the convergence of this sum can be checked easily (e.g. as in [32, Lem. 3.10]). Furthermore, we get for $M > N$

$$\begin{aligned} & \mathbb{E} \left[\left(\int_{\mathbb{T}^2} : \phi_M^n(x) : dx - \int_{\mathbb{T}^2} : \phi_N^n(x) : dx \right)^2 \right] \\ &= L^{4-2n} \sum_{\substack{k_1+k_2+\dots+k_n=0 \\ |k_i| \leq M}} \frac{1}{|\lambda_{k_1} + m^2|} \cdots \frac{1}{|\lambda_{k_n} + m^2|} \\ &\quad - L^{4-2n} \sum_{\substack{k_1+k_2+\dots+k_n=0 \\ |k_i| \leq N}} \frac{1}{|\lambda_{k_1} + m^2|} \cdots \frac{1}{|\lambda_{k_n} + m^2|} \\ &\leq C_{n,L} \frac{(\log N)^{n-2}}{N^2}, \end{aligned} \tag{A.11}$$

for a constant $C_{n,L}$ which depends on n, L but not on N, M .

Finally, we recall the definition of Wiener chaos which in this finite dimensional context is the following:

Definition A.3. For $n \in \mathbb{N}_0$ the n -th (inhomogeneous) Wiener chaos generated by the random variables $(z_k)_{|k| \leq N}$ is the vector space of real-valued random variables X which can be written as polynomials of degree at most n in the finitely many random variables z_k .

As stated this definition depends on the number of independent Gaussians used to define the Wiener chaos. However, the following classical and important estimate holds true uniformly in that number. See e.g. [10, Thm. 4.1] for a direct proof. This Theorem can also be deduced immediately from the hyper-contractivity of the Ornstein–Uhlenbeck semigroup [26, Thm. 1.4.1].

Lemma A.4 (Equivalence of moments). *Let X be a random variable, belonging to the n -th inhomogeneous Wiener chaos. Then for any $p \geq 1$ one has*

$$\mathbb{E} [X^{2p}]^{\frac{1}{2p}} \leq C_n (2p - 1)^{\frac{n}{2}} \mathbb{E} [X^2]^{\frac{1}{2}} \tag{A.12}$$

where C_n only depends on n .

Remark A.5. The n -th *homogeneous* Wiener chaos is defined as the orthogonal complement (with respect to the L^2 scalar product) of the $n - 1$ -st inhomogeneous Wiener chaos in the n -th inhomogeneous Wiener chaos. If in the previous Lemma, X takes values in the homogeneous Wiener chaos, then the estimate holds true with constant $C_n = 1$.

Now combining Lemma A.4 with (A.10) we obtain for $p \geq 1$ that

$$\sup_N \mathbb{E} \left[\left(\int_{\mathbb{T}^2} : \phi_N^n(x) : dx \right)^{2p} \right] < \infty \quad (\text{A.13})$$

and combining Lemma A.4 with (A.11) we see that for $M > N$ and any $p \geq 1$

$$\mathbb{E} \left[\left(\int_{\mathbb{T}^2} : \phi_M^n(x) : dx - \int_{\mathbb{T}^2} : \phi_N^n(x) : dx \right)^{2p} \right]^{\frac{1}{p}} \leq C_{n,L} (2p - 1) \frac{(\log N)^{n-2}}{N} \quad (\text{A.14})$$

for a constant $C_{n,L}$ which depends on n, L but not on p, M, N .

References

- [1] Shigeki Aida, *Semi-classical limit of the lowest eigenvalue of a Schrödinger operator on a Wiener space. II. $P(\phi)_2$ -model on a finite volume*, J. Funct. Anal. **256** (2009), no. 10, 3342–3367. MR-2504528
- [2] Florent Barret, *Sharp asymptotics of metastable transition times for one dimensional SPDEs*, Ann. Inst. Henri Poincaré Probab. Stat. **51** (2015), no. 1, 129–166. MR-3300966
- [3] Florent Barret, Anton Bovier, and Sylvie Méléard, *Uniform estimates for metastable transition times in a coupled bistable system*, Electron. J. Probab. **15** (2010), no. 12, 323–345. MR-2609590
- [4] Nils Berglund, *Kramers’ law: validity, derivations and generalisations*, Markov Process. Related Fields **19** (2013), no. 3, 459–490. MR-3156961, arXiv:1106.5799.
- [5] Nils Berglund, Bastien Fernandez, and Barbara Gentz, *Metastability in interacting nonlinear stochastic differential equations. I. From weak coupling to synchronization*, Nonlinearity **20** (2007), no. 11, 2551–2581. MR-2361246
- [6] Nils Berglund and Barbara Gentz, *Sharp estimates for metastable lifetimes in parabolic SPDEs: Kramers’ law and beyond*, Electron. J. Probab. **18** (2013), no. 24, 58. MR-3035752
- [7] Anton Bovier and Frank den Hollander, *Metastability. a potential-theoretic approach*, Grundlehren der Mathematischen Wissenschaften [Fundamental Principles of Mathematical Sciences], vol. 351, Springer, Cham, 2015. MR-3445787
- [8] Anton Bovier, Michael Eckhoff, Véronique Gayraud, and Markus Klein, *Metastability in reversible diffusion processes. I. Sharp asymptotics for capacities and exit times*, J. Eur. Math. Soc. (JEMS) **6** (2004), no. 4, 399–424. MR-2094397
- [9] Giuseppe Da Prato and Arnaud Debussche, *Strong solutions to the stochastic quantization equations*, Ann. Probab. **31** (2003), no. 4, 1900–1916. MR-2016604
- [10] Giuseppe Da Prato and Luciano Tubaro, *Wick powers in stochastic PDEs: an introduction*, Tech. Report UTM 711, University of Trento, 2007.
- [11] Giacomo Di Gesù and Dorian Le Peutrec, *Small noise spectral gap asymptotics for a large system of nonlinear diffusions*, Journal of Spectral Theory, to appear (2017), 1–46, arXiv:1506.04434.
- [12] Henry Eyring, *The activated complex in chemical reactions*, The Journal of Chemical Physics **3** (1935), no. 2, 107–115.
- [13] William G. Faris and Giovanni Jona-Lasinio, *Large fluctuations for a nonlinear heat equation with noise*, J. Phys. A **15** (1982), no. 10, 3025–3055. MR-684578
- [14] Robin Forman, *Functional determinants and geometry*, Invent. Math. **88** (1987), no. 3, 447–493. MR-884797

- [15] M. I. Freidlin and A. D. Wentzell, *Random perturbations of dynamical systems*, second ed., Grundlehren der Mathematischen Wissenschaften [Fundamental Principles of Mathematical Sciences], vol. 260, Springer-Verlag, New York, 1998, Translated from the 1979 Russian original by Joseph Szücs. MR-1652127
- [16] James Glimm and Arthur Jaffe, *Quantum physics*, Springer-Verlag, New York-Berlin, 1981, A functional integral point of view. MR-628000
- [17] Martin Hairer, *A theory of regularity structures*, *Invent. Math.* **198** (2014), no. 2, 269–504. MR-3274562
- [18] Martin Hairer, Marc D. Ryser, and Hendrik Weber, *Triviality of the 2D stochastic Allen-Cahn equation*, *Electron. J. Probab.* **17** (2012), no. 39, 14. MR-2928722
- [19] Martin Hairer and Hendrik Weber, *Large deviations for white-noise driven, nonlinear stochastic PDEs in two and three dimensions*, *Ann. Fac. Sci. Toulouse Math. (6)* **24** (2015), no. 1, 55–92. MR-3325951
- [20] Bernard Helffer, Markus Klein, and Francis Nier, *Quantitative analysis of metastability in reversible diffusion processes via a Witten complex approach*, *Mat. Contemp.* **26** (2004), 41–85. MR-2111815
- [21] G. Jona-Lasinio and P. K. Mitter, *Large deviation estimates in the stochastic quantization of ϕ_2^4* , *Comm. Math. Phys.* **130** (1990), no. 1, 111–121. MR-1055688
- [22] H. A. Kramers, *Brownian motion in a field of force and the diffusion model of chemical reactions*, *Physica* **7** (1940), 284–304. MR-0002962
- [23] Robert S. Maier and Daniel L. Stein, *Droplet nucleation and domain wall motion in a bounded interval*, *Phys. Rev. Lett.* **87** (2001), 270601–1.
- [24] Fabio Martinelli, Enzo Olivieri, and Elisabetta Scoppola, *Small random perturbations of finite- and infinite-dimensional dynamical systems: unpredictability of exit times*, *J. Statist. Phys.* **55** (1989), no. 3-4, 477–504. MR-1003525
- [25] Edward Nelson, *The free Markoff field*, *J. Functional Analysis* **12** (1973), 211–227. MR-0343816
- [26] David Nualart, *The Malliavin calculus and related topics*, second ed., Probability and its Applications (New York), Springer-Verlag, Berlin, 2006. MR-2200233
- [27] F. W. J. Olver, *Asymptotics and special functions*, Academic Press [A subsidiary of Harcourt Brace Jovanovich, Publishers], New York-London, 1974, Computer Science and Applied Mathematics. MR-0435697
- [28] Joran Rolland, Freddy Bouchet, and Eric Simonnet, *Computing Transition Rates for the 1-D Stochastic Ginzburg–Landau–Allen–Cahn Equation for Finite-Amplitude Noise with a Rare Event Algorithm*, *J. Stat. Phys.* **162** (2016), no. 2, 277–311. MR-3441361
- [29] Barry Simon, *Trace ideals and their applications*, second ed., Mathematical Surveys and Monographs, vol. 120, American Mathematical Society, Providence, RI, 2005. MR-2154153
- [30] Makoto Sugiura, *Metastable behaviors of diffusion processes with small parameter*, *J. Math. Soc. Japan* **47** (1995), no. 4, 755–788. MR-1348758
- [31] Pavlos Tsatsoulis and Hendrik Weber, *Spectral Gap for the Stochastic Quantization Equation on the 2-dimensional Torus*, *Annales de l’Institut Henri Poincaré*, to appear (2017), 1–46, arXiv:1609.08447.
- [32] Rongchan Zhu and Xiangchan Zhu, *Three-dimensional Navier-Stokes equations driven by space-time white noise*, *J. Differential Equations* **259** (2015), no. 9, 4443–4508. MR-3373412

Electronic Journal of Probability

Electronic Communications in Probability

Advantages of publishing in EJP-ECP

- Very high standards
- Free for authors, free for readers
- Quick publication (no backlog)
- Secure publication (LOCKSS¹)
- Easy interface (EJMS²)

Economical model of EJP-ECP

- Non profit, sponsored by IMS³, BS⁴, ProjectEuclid⁵
- Purely electronic

Help keep the journal free and vigorous

- Donate to the IMS open access fund⁶ (click here to donate!)
- Submit your best articles to EJP-ECP
- Choose EJP-ECP over for-profit journals

¹LOCKSS: Lots of Copies Keep Stuff Safe <http://www.lockss.org/>

²EJMS: Electronic Journal Management System <http://www.vtex.lt/en/ejms.html>

³IMS: Institute of Mathematical Statistics <http://www.imstat.org/>

⁴BS: Bernoulli Society <http://www.bernoulli-society.org/>

⁵Project Euclid: <https://projecteuclid.org/>

⁶IMS Open Access Fund: <http://www.imstat.org/publications/open.htm>