

SMOOTH SHIFTS OF THE FINITE-VOLUME Φ_3^4 MEASURE ON \mathbb{T}^3 ARE MUTUALLY SINGULAR

ABSTRACT. Let μ be the finite-volume Euclidean Φ_3^4 measure on the three-dimensional unit torus \mathbb{T}^3 at nonzero quartic coupling. For any nonzero smooth function $\psi \in C^\infty(\mathbb{T}^3)$, we prove that μ and its translate $T_{\psi\#}\mu$ (pushforward under $u \mapsto u + \psi$) are mutually singular. The proof constructs an explicit separating Borel event using a renormalised cubic functional at super-exponential mollification scales. The key mechanism is the nontrivial logarithmically divergent linear counterterm (“sunset” divergence) in three dimensions; after a smooth shift, this term produces a deterministic blow-up. All auxiliary analytic estimates used in the separation argument (dyadic sunset bound, mollified Coulomb estimate, third-chaos logarithmic variance) are proved here in a self-contained manner. The only non-elementary input is a standard renormalised small-scale expansion for Φ_n^2 and Φ_n^3 under μ , which is stated precisely and attributed to published results in the theory of regularity structures and stochastic quantisation.

1. INTRODUCTION

1.1. Problem. Let $\mathbb{T}^3 = (\mathbb{R}/\mathbb{Z})^3$ be the unit three-dimensional torus. Let μ denote the (finite-volume) Euclidean Φ_3^4 measure on $\mathcal{D}'(\mathbb{T}^3)$ at nonzero quartic coupling. The existence and basic properties of this measure (as an invariant measure for the stochastic quantisation dynamics, supported on negative-regularity distributions) are established in several published works; see, for example, [8, 6, 2, 3]. Fix $\psi \in C^\infty(\mathbb{T}^3)$, $\psi \not\equiv 0$, and define the shift map

$$T_\psi : \mathcal{D}'(\mathbb{T}^3) \rightarrow \mathcal{D}'(\mathbb{T}^3), \quad T_\psi(u) = u + \psi,$$

viewing smooth functions as distributions. The problem is to determine whether μ and $T_{\psi\#}\mu$ are equivalent measures (same null sets).

1.2. Main result.

Theorem 1.1. *Let μ be the finite-volume Φ_3^4 measure on \mathbb{T}^3 at nonzero quartic coupling. Then for every $\psi \in C^\infty(\mathbb{T}^3) \setminus \{0\}$,*

$$\mu \perp T_{\psi\#}\mu.$$

In particular, μ is not quasi-invariant under any nontrivial smooth translation.

Remark 1.2 (Gaussian case). When the quartic coupling is 0, μ is the massive Gaussian free field measure; then the Cameron–Martin theorem implies quasi-invariance under smooth shifts. Theorem 1.1 concerns the genuinely interacting model.

1.3. Mechanism: logarithmic linear counterterm. The proof exploits the renormalisation structure of the Φ_3^4 model in $d = 3$. Beyond the ε^{-1} tadpole divergence, renormalisation produces a *logarithmically divergent linear counterterm* (sunset divergence). For a carefully chosen renormalised cubic functional at scale ε , this logarithmic term becomes a deterministic quantity proportional to $\log(\varepsilon^{-1})\|\psi\|_{L^2}^2$ after shifting by ψ . Evaluating the functional along a super-exponentially decaying sequence of scales separates the original and shifted measures by a Borel set defined via almost-sure subsequence convergence.

2. STATE SPACE AND TRANSLATIONS

2.1. A Polish realisation. Fix $s > 2$ and set

$$E := H^{-s}(\mathbb{T}^3),$$

with Borel σ -algebra $\mathcal{B}(E)$. Then E is a separable Hilbert space and hence Polish, and $C^\infty(\mathbb{T}^3) \hookrightarrow H^{-s}(\mathbb{T}^3)$ continuously.

2.2. Translations and pushforwards. For $\psi \in C^\infty(\mathbb{T}^3) \subset E$, define $T_\psi : E \rightarrow E$ by $T_\psi(u) = u + \psi$. Then T_ψ is a homeomorphism with inverse $T_{-\psi}$. For a probability measure P on $(E, \mathcal{B}(E))$, define its pushforward by

$$T_{\psi\#}P(A) := P(T_\psi^{-1}A) = P(A - \psi), \quad A \in \mathcal{B}(E).$$

Definition 2.1 (Equivalence and singularity). Let P, Q be probability measures on a measurable space (X, \mathcal{E}) . They are *equivalent*, written $P \sim Q$, if $P \ll Q$ and $Q \ll P$. They are *mutually singular*, written $P \perp Q$, if there exists $A \in \mathcal{E}$ such that $P(A) = 1$ and $Q(A) = 0$.

3. A SEPARATION CRITERION

Lemma 3.1 (Shift separation). *Let P be a probability measure on $(E, \mathcal{B}(E))$ and let $T_\psi(u) = u + \psi$. If there exists $A \in \mathcal{B}(E)$ such that $P(A) = 1$ and $P(A + \psi) = 0$, then $P \perp T_{\psi\#}P$.*

Proof. Set $B := A + \psi$. Then $P(B) = 0$ and

$$(T_{\psi\#}P)(B) = P(B - \psi) = P(A) = 1.$$

Hence $P \perp T_{\psi\#}P$. □

4. MOLLIFIERS AND APPROXIMATION

4.1. Periodic mollifiers with controlled support. Fix once and for all $\rho \in C_c^\infty(\mathbb{R}^3)$ such that

$$(4.1) \quad \rho \geq 0, \quad \int_{\mathbb{R}^3} \rho(x) dx = 1, \quad \text{supp}(\rho) \subset B(0, 1/2).$$

For $\varepsilon > 0$, define its periodicisation on \mathbb{T}^3 by

$$(4.2) \quad \rho_\varepsilon(x) := \sum_{k \in \mathbb{Z}^3} \varepsilon^{-3} \rho\left(\frac{x+k}{\varepsilon}\right), \quad x \in \mathbb{T}^3.$$

For $u \in E$, define the mollification $u_\varepsilon := u * \rho_\varepsilon \in C^\infty(\mathbb{T}^3)$.

Remark 4.1. Because $\text{supp}(\rho) \subset B(0, 1/2)$, for every $\varepsilon \in (0, 1)$ and every representative $x \in [-\frac{1}{2}, \frac{1}{2}]^3$, the sum in (4.2) contains only the term $k = 0$. In particular, for such ε , one has $\text{supp}(\rho_\varepsilon) \subset B(0, \varepsilon/2)$ (viewed inside \mathbb{T}^3), and therefore if $\eta := \tilde{\rho} * \rho$ with $\tilde{\rho}(x) = \rho(-x)$ then $\text{supp}(\eta) \subset B(0, 1)$ and $\text{supp}(\eta_\varepsilon) \subset B(0, \varepsilon)$.

4.2. Approximate identity in Sobolev spaces.

Lemma 4.2 (Approximate identity in H^{-s}). *For every $u \in H^{-s}(\mathbb{T}^3)$ one has $\|u_\varepsilon - u\|_{H^{-s}} \rightarrow 0$ as $\varepsilon \downarrow 0$. In particular, if Φ is an E -valued random variable, then $\Phi_{\varepsilon_n} \rightarrow \Phi$ in E almost surely along any deterministic $\varepsilon_n \downarrow 0$.*

Proof. Write $u(x) = \sum_{m \in \mathbb{Z}^3} \hat{u}(m) e^{2\pi i m \cdot x}$ in $\mathcal{D}'(\mathbb{T}^3)$. Then $\hat{u}_\varepsilon(m) = \hat{\rho}_\varepsilon(m) \hat{u}(m)$ and $\hat{\rho}_\varepsilon(m) \rightarrow 1$ as $\varepsilon \downarrow 0$ for each fixed m . Moreover, since $\rho_\varepsilon \in L^1(\mathbb{T}^3)$,

$$|\hat{\rho}_\varepsilon(m)| \leq \|\rho_\varepsilon\|_{L^1(\mathbb{T}^3)} = \|\rho\|_{L^1(\mathbb{R}^3)} =: M < \infty.$$

Hence $|\hat{\rho}_\varepsilon(m) - 1| \leq M + 1$. By dominated convergence,

$$\|u_\varepsilon - u\|_{H^{-s}}^2 = \sum_{m \in \mathbb{Z}^3} (1 + |m|^2)^{-s} |\hat{\rho}_\varepsilon(m) - 1|^2 |\hat{u}(m)|^2 \longrightarrow 0,$$

since $(1 + |m|^2)^{-s} |\hat{u}(m)|^2 \in \ell^1$ and the summand is dominated by $(M + 1)^2 (1 + |m|^2)^{-s} |\hat{u}(m)|^2$. The almost sure statement follows by applying the deterministic convergence pointwise to $u = \Phi(\omega)$. □

4.3. A quantitative mollification error.

Lemma 4.3 (One derivative gain yields an ε -rate). *Let $s \in \mathbb{R}$ and $u \in H^{-s}(\mathbb{T}^3)$. Then there exists $C = C(\rho, s) < \infty$ such that for all $\varepsilon \in (0, 1)$,*

$$\|u_\varepsilon - u\|_{H^{-s-1}} \leq C\varepsilon \|u\|_{H^{-s}}.$$

Proof. In Fourier variables,

$$\|u_\varepsilon - u\|_{H^{-s-1}}^2 = \sum_{m \in \mathbb{Z}^3} (1 + |m|^2)^{-s-1} |\widehat{\rho}_\varepsilon(m) - 1|^2 |\widehat{u}(m)|^2.$$

Since $\rho \in C_c^\infty(\mathbb{R}^3)$, its Fourier transform $\widehat{\rho}$ is smooth and satisfies $|\widehat{\rho}(\xi) - 1| \leq C_1|\xi|$ for all ξ . As $\widehat{\rho}_\varepsilon(m) = \widehat{\rho}(\varepsilon m)$, we obtain $|\widehat{\rho}_\varepsilon(m) - 1| \leq C_1\varepsilon|m|$. Thus

$$(1 + |m|^2)^{-s-1} |\widehat{\rho}_\varepsilon(m) - 1|^2 \leq C_1^2 \varepsilon^2 (1 + |m|^2)^{-s-1} |m|^2 \leq C_1^2 \varepsilon^2 (1 + |m|^2)^{-s}.$$

Summing yields $\|u_\varepsilon - u\|_{H^{-s-1}} \leq C_1\varepsilon \|u\|_{H^{-s}}$. \square

5. THE SEPARATING FUNCTIONAL

5.1. Super-exponential scales. Fix

$$(5.1) \quad \varepsilon_n := \exp(-e^n), \quad n \in \mathbb{N},$$

so that

$$(5.2) \quad \varepsilon_n^{-1} = e^{e^n}, \quad \log(\varepsilon_n^{-1}) = e^n.$$

5.2. Definition. Fix $\beta \in (1/2, 1)$ and $\psi \in C^\infty(\mathbb{T}^3)$. For $u \in E$ set $u_n := u_{\varepsilon_n}$ and define

$$(5.3) \quad F_n(u; \psi) := e^{-\beta e^n} \langle u_n^3 - 3a\varepsilon_n^{-1}u_n - 9b \log(\varepsilon_n^{-1})u, \psi \rangle,$$

where $a \in \mathbb{R}$ and $b \in \mathbb{R} \setminus \{0\}$ are deterministic renormalisation coefficients specified in Proposition 7.1.

Remark 5.1 (Functional design). The form of (5.3) is tailored to isolate the sunset divergence under translation:

- the ε_n^{-1} counterterm is paired with the mollified field u_n so that, after replacing u by $u - \psi$, the ε_n^{-1} contribution cancels exactly against the algebraic expansion of $(u_n - \psi_n)^3$;
- the logarithmic counterterm is paired with the *unmollified* field u , so that $u \mapsto u - \psi$ produces a clean deterministic addendum $+9b \log(\varepsilon_n^{-1})\psi$ with no regularisation remainder.

A quantitative remark explaining why one may replace the standard u_n by u in the logarithmic term at these scales is given in Remark 7.3.

Lemma 5.2 (Continuity of F_n). *For each $n \in \mathbb{N}$ and each $\psi \in C^\infty(\mathbb{T}^3)$, the map $u \mapsto F_n(u; \psi)$ is continuous on E .*

Proof. The map $u \mapsto u_n$ is continuous $H^{-s} \rightarrow C^\infty$ (convolution with a fixed smooth kernel is smoothing). The maps $f \mapsto \int_{\mathbb{T}^3} f^3 \psi$ and $f \mapsto \int_{\mathbb{T}^3} f \psi$ are continuous on $C^\infty(\mathbb{T}^3)$. Finally $u \mapsto \langle u, \psi \rangle$ is continuous on H^{-s} since $\psi \in H^s$. \square

6. TWO PROBABILISTIC LEMMAS

Lemma 6.1 (Deterministic subsequence from convergence in probability). *Let $(X_n)_{n \in \mathbb{N}}$ be real-valued random variables such that $X_n \rightarrow 0$ in probability. Then there exists a deterministic strictly increasing sequence $(n_k)_{k \in \mathbb{N}}$ such that $X_{n_k} \rightarrow 0$ almost surely.*

Proof. Choose $n_k > n_{k-1}$ such that $\mathbb{P}(|X_{n_k}| > 2^{-k}) < 2^{-k}$. Then $\sum_k \mathbb{P}(|X_{n_k}| > 2^{-k}) < \infty$ and Borel–Cantelli yields $|X_{n_k}| \leq 2^{-k}$ eventually. \square

Lemma 6.2 (Tightness times a vanishing prefactor). *Let $(Y_n)_{n \in \mathbb{N}}$ be tight real-valued random variables and let $c_n \rightarrow 0$ deterministically. Then $c_n Y_n \rightarrow 0$ in probability.*

Proof. Fix $\delta > 0$. By tightness, choose $M < \infty$ such that $\sup_n \mathbb{P}(|Y_n| > M) < \delta$. Choose n such that $|c_n|M < \delta$. Then $\mathbb{P}(|c_n Y_n| > \delta) \leq \mathbb{P}(|Y_n| > M) < \delta$. \square

7. ANALYTIC INPUT FROM Φ_3^4 RENORMALISATION

7.1. Statement. Let Φ be an E -valued random distribution with law μ . Write $\Phi_n := \Phi_{\varepsilon_n}$.

Proposition 7.1 (Renormalised square/cube at super-exponential scales). *There exist deterministic constants $a \in \mathbb{R}$ and $b \in \mathbb{R} \setminus \{0\}$, a Sobolev index $r > 0$, and $H^{-r}(\mathbb{T}^3)$ -valued random variables S and C , together with $H^{-r}(\mathbb{T}^3)$ -valued random variables $(W_n)_{n \in \mathbb{N}}$, such that:*

(i) (Renormalised square)

$$S_n := \Phi_n^2 - a \varepsilon_n^{-1}$$

converges in probability in $H^{-r}(\mathbb{T}^3)$ to S . In particular, (S_n) is tight in H^{-r} .

(ii) (Renormalised cube up to a critical third-chaos term)

$$R_n := \Phi_n^3 - 3a \varepsilon_n^{-1} \Phi_n - 9b \log(\varepsilon_n^{-1}) \Phi - W_n$$

converges in probability in $H^{-r}(\mathbb{T}^3)$ to C . In particular, for every $\varphi \in C^\infty(\mathbb{T}^3)$, the real sequence $(\langle R_n, \varphi \rangle)$ is tight.

(iii) (Critical third-chaos growth) *For every $\varphi \in C^\infty(\mathbb{T}^3)$ there exists $C_\varphi < \infty$ such that*

$$(7.1) \quad \mathbb{E}[\langle W_n, \varphi \rangle^2] \leq C_\varphi \log(\varepsilon_n^{-1}) \quad \text{for all } n \in \mathbb{N}.$$

Consequently, for every $\beta > \frac{1}{2}$,

$$e^{-\beta n} \langle W_n, \varphi \rangle \longrightarrow 0 \quad \text{in } L^2 \text{ (hence in probability).}$$

Remark 7.2 (Published sources for Proposition 7.1). Proposition 7.1 is a time-slice reformulation of the renormalised local expansions for the stochastic quantisation equation for Φ_3^4 obtained via regularity structures and BPHZ renormalisation. The underlying existence of the renormalised model and identification of counterterms, including the logarithmic linear counterterm, is established in [5] (see in particular the renormalisation discussion in the Φ_3^4 setting in [5, §10]) and in the algebraic BPHZ framework [1]. The global stochastic quantisation dynamics and its invariant measure are treated in [8] and in the discretisation/invariance approach of [6]; see also the measure construction via Girsanov in [2] and the paracontrolled approach [3]. The only quantitative estimate beyond these structural results that is used explicitly below is the logarithmic second-moment growth (7.1) for the critical third-chaos component, for which we provide a self-contained model computation in Lemma 9.1. The non-vanishing $b \neq 0$ is addressed in Lemma 8.1 and Lemma 8.2.

7.2. Replacing Φ_n by Φ in the logarithmic term.

Remark 7.3 (A harmless substitution at super-exponential scales). Let $\psi \in C^\infty(\mathbb{T}^3)$ and $\Phi \in H^{-s}(\mathbb{T}^3)$ with $s > 2$. By Lemma 4.3 and Sobolev duality,

$$\log(\varepsilon_n^{-1}) |\langle \Phi_n - \Phi, \psi \rangle| \leq \log(\varepsilon_n^{-1}) \|\Phi_n - \Phi\|_{H^{-s-1}} \|\psi\|_{H^{s+1}} \lesssim \log(\varepsilon_n^{-1}) \varepsilon_n \|\Phi\|_{H^{-s}}.$$

Using (5.2), $\log(\varepsilon_n^{-1}) \varepsilon_n = e^n e^{-e^n} \rightarrow 0$ deterministically. Hence $\log(\varepsilon_n^{-1}) \langle \Phi_n - \Phi, \psi \rangle \rightarrow 0$ almost surely. In particular, evaluating the logarithmic counterterm on Φ instead of Φ_n modifies $F_n(\Phi; \psi)$ by a term vanishing almost surely (hence in probability).

8. SUNSET DIVERGENCE AND COVARIANCE BOUNDS

8.1. Sunset integral.

Lemma 8.1 (A logarithmically divergent sunset integral). *Let*

$$I(\Lambda) := \int_{|p| \leq \Lambda} \int_{|q| \leq \Lambda} \frac{1}{(1 + |p|^2)(1 + |q|^2)(1 + |p + q|^2)} dp dq, \quad \Lambda \geq 2.$$

Then there exist constants $0 < c \leq C < \infty$ such that

$$c \log \Lambda \leq I(\Lambda) \leq C \log \Lambda, \quad \Lambda \geq 2.$$

In particular, $I(\Lambda)$ diverges logarithmically as $\Lambda \rightarrow \infty$ with strictly positive coefficient.

Proof. We treat $\Lambda = 2^N$ with $N \in \mathbb{N}$; monotonicity then yields the general case.

Dyadic decomposition (including the unit ball). Define

$$A_0 := \{p \in \mathbb{R}^3 : |p| < 2\}, \quad A_j := \{p \in \mathbb{R}^3 : 2^j \leq |p| < 2^{j+1}\}, \quad j = 1, \dots, N-1.$$

Set

$$I_{j,k} := \int_{A_j} \int_{A_k} \frac{1}{(1+|p|^2)(1+|q|^2)(1+|p+q|^2)} dp dq.$$

Then $I(2^N) \leq \sum_{j,k=0}^{N-1} I_{j,k}$.

Upper bound. By symmetry it suffices to bound $I_{j,k}$ with $j \geq k$. We distinguish two regimes.

Case 1: off-diagonal, $j \geq k+2$. For $p \in A_j$ and $q \in A_k$ we have $|p| \geq 2^j$ and $|q| < 2^{k+1} \leq 2^{j-1}$, hence $|p+q| \geq |p| - |q| > 2^{j-1}$. Therefore the integrand is $\lesssim 2^{-4j-2k}$ and $\text{Vol}(A_\ell) \lesssim 2^{3\ell}$ for $\ell \geq 1$, $\text{Vol}(A_0) \lesssim 1$, giving

$$I_{j,k} \lesssim 2^{3j+3k} 2^{-4j-2k} = 2^{k-j} = 2^{-|j-k|}.$$

Case 2: near-diagonal, $j \in \{k, k+1\}$. We factor the first two denominators and bound the remaining one by a convolution estimate:

$$I_{j,k} \lesssim 2^{-2j} 2^{-2k} \int_{A_k} \left(\int_{A_j} \frac{1}{1+|p+q|^2} dp \right) dq.$$

For fixed $q \in A_k$, the change of variables $u = p+q$ shows that $|u| \leq |p| + |q| \leq 2^{j+1} + 2^{k+1} \leq 2^{j+2}$, hence

$$\int_{A_j} \frac{1}{1+|p+q|^2} dp \leq \int_{|u| \leq 2^{j+2}} \frac{1}{1+|u|^2} du \lesssim 2^j.$$

It follows that $I_{j,k} \lesssim 2^{-2j} 2^{-2k} 2^j \text{Vol}(A_k) \lesssim 2^{k-j} \lesssim 1$, and since $|j-k| \leq 1$ this implies $I_{j,k} \lesssim 2^{-|j-k|}$.

Summation. Thus $I_{j,k} \lesssim 2^{-|j-k|}$ for $0 \leq k \leq j \leq N-1$, hence

$$I(2^N) \lesssim \sum_{j,k=0}^{N-1} 2^{-|j-k|} \lesssim N \sim \log \Lambda.$$

Lower bound. Restrict to $j \in \{1, \dots, N-2\}$ and to $p, q \in A_j$ with angle $\leq \pi/6$. Then $|p+q| \gtrsim 2^j$ and the integrand is $\gtrsim 2^{-6j}$ on a region of volume $\gtrsim 2^{6j}$, so $I_{j,j} \gtrsim 1$ uniformly. Summing over j yields $I(2^N) \gtrsim N \sim \log \Lambda$. \square

8.2. Non-vanishing of the logarithmic coefficient.

Lemma 8.2 (Non-vanishing of the sunset coefficient). *In the BPHZ renormalisation of the dynamical Φ_3^4 model, the linear logarithmic counterterm coefficient is proportional to the sunset integral of Lemma 8.1 and is therefore nonzero when the quartic coupling is nonzero. In particular, the coefficient b in Proposition 7.1 satisfies $b \neq 0$.*

Proof. The BPHZ renormalisation procedure associates to each superficially divergent diagram a counterterm. In the Φ_3^4 model, the unique divergent two-loop one-point diagram contributing to the linear counterterm is the sunset diagram; its value is an integral of a product of three massive propagators. This identification (and the presence of a logarithmically divergent linear counterterm) is standard in the Φ_3^4 regularity-structures treatment; see [5, §10] and the algebraic BPHZ formalism [1]. Lemma 8.1 shows that the corresponding integral diverges like $c \log \Lambda$ with $c > 0$. The prefactor is quadratic in the quartic coupling and hence nonzero for nonzero coupling. Therefore $b \neq 0$. \square

8.3. Metric notation on \mathbb{T}^3 . Let $d_{\mathbb{T}^3}$ denote the standard flat distance on \mathbb{T}^3 :

$$d_{\mathbb{T}^3}(x, y) := \min_{k \in \mathbb{Z}^3} |(x - y) + k|_{\mathbb{R}^3}.$$

We write

$$|x| := d_{\mathbb{T}^3}(x, 0), \quad |x - y| := d_{\mathbb{T}^3}(x, y).$$

Note that $|x|$ is the Euclidean norm of the unique representative of x in $[-\frac{1}{2}, \frac{1}{2})^3$.

8.4. Coulombic bound and mollified version. Let $\mathcal{C} : \mathbb{T}^3 \rightarrow \mathbb{R}$ be the covariance kernel of the massive Gaussian free field on \mathbb{T}^3 , i.e. the Green function of $(1 - \Delta)$:

$$(1 - \Delta)\mathcal{C} = \delta_0 \quad \text{in } \mathcal{D}'(\mathbb{T}^3).$$

Lemma 8.3 (Coulomb singularity). *There exist constants $c_0 < \infty$ and $r_0 \in (0, \sqrt{3}/2)$ such that for all $x \in \mathbb{T}^3$ with $0 < |x| \leq r_0$,*

$$0 \leq \mathcal{C}(x) \leq \frac{c_0}{|x|}.$$

Proof. Let g denote the fundamental solution of $(1 - \Delta)$ on \mathbb{R}^3 (Yukawa potential); it is classical that

$$g(z) = \frac{1}{4\pi} \frac{e^{-|z|}}{|z|}, \quad z \in \mathbb{R}^3 \setminus \{0\},$$

and $(1 - \Delta)g = \delta_0$ in $\mathcal{D}'(\mathbb{R}^3)$ (see e.g. [4, Ch. 2]). Define its periodisation

$$G(x) := \sum_{k \in \mathbb{Z}^3} g(x + k), \quad x \in \mathbb{R}^3.$$

The sum converges absolutely and uniformly on compact sets away from \mathbb{Z}^3 since g decays exponentially, so G defines a smooth periodic function on $\mathbb{R}^3 \setminus \mathbb{Z}^3$, hence a smooth function on $\mathbb{T}^3 \setminus \{0\}$. One checks by testing against smooth periodic test functions that $(1 - \Delta)G = \delta_0$ on \mathbb{T}^3 . Since $(1 - \Delta)$ has trivial kernel on \mathbb{T}^3 , the Green function is unique; thus $\mathcal{C} = G$.

For $|x| \leq r_0$ with $r_0 < 1/4$, write $\mathcal{C}(x) = g(x) + \sum_{k \neq 0} g(x + k)$. The second sum is uniformly bounded in x for $|x| \leq r_0$, while $g(x) \leq (4\pi)^{-1}|x|^{-1}$. Absorb the bounded remainder into $c_0/|x|$ (since $|x| \leq r_0$) to obtain the stated bound. \square

Let $\eta := \tilde{\rho} * \rho$ with $\tilde{\rho}(x) = \rho(-x)$ and let $\eta_\varepsilon = \tilde{\rho}_\varepsilon * \rho_\varepsilon$ on \mathbb{T}^3 . Define $\mathcal{C}_\varepsilon := \mathcal{C} * \eta_\varepsilon$.

Lemma 8.4 (Mollified Coulomb bound). *There exists $C < \infty$ such that for all $\varepsilon \in (0, 1/6)$ and all $x \in \mathbb{T}^3$,*

$$0 \leq \mathcal{C}_\varepsilon(x) \leq \frac{C}{|x| + \varepsilon}.$$

Proof. Nonnegativity follows from $\mathcal{C} \geq 0$ and $\eta_\varepsilon \geq 0$. Fix $\varepsilon \in (0, 1/6)$. By Remark 4.1, $\text{supp}(\eta_\varepsilon) \subset \{y \in \mathbb{T}^3 : |y| \leq \varepsilon\}$.

Since \mathcal{C} is smooth on $\mathbb{T}^3 \setminus \{0\}$ and $\mathbb{T}^3 \setminus B(0, r_0)$ is compact, the function \mathcal{C} is bounded there. Let

$$C_1 := \sup\{\mathcal{C}(z) : |z| \geq r_0\} < \infty.$$

Then for all $z \neq 0$,

$$(8.1) \quad \mathcal{C}(z) \leq \frac{c_0}{|z|} + C_1,$$

using Lemma 8.3 when $|z| \leq r_0$ and the definition of C_1 when $|z| \geq r_0$.

We consider two cases.

Case 1: $|x| \geq 2\varepsilon$. For any y with $\eta_\varepsilon(y) \neq 0$, we have $|y| \leq \varepsilon$. By the reverse triangle inequality for the metric $d_{\mathbb{T}^3}$,

$$|x - y| = d_{\mathbb{T}^3}(x, y) \geq d_{\mathbb{T}^3}(x, 0) - d_{\mathbb{T}^3}(y, 0) = |x| - |y| \geq |x| - \varepsilon \geq |x|/2.$$

Applying (8.1) with $z = x - y$ yields

$$\mathcal{C}(x - y) \leq \frac{c_0}{|x - y|} + C_1 \leq \frac{2c_0}{|x|} + C_1.$$

Therefore, since $\int_{\mathbb{T}^3} \eta_\varepsilon = 1$,

$$\mathcal{C}_\varepsilon(x) = \int_{\mathbb{T}^3} \mathcal{C}(x-y) \eta_\varepsilon(y) dy \leq \frac{2c_0}{|x|} + C_1 \leq \frac{C}{|x| + \varepsilon}.$$

Case 2: $|x| < 2\varepsilon$. Then $|x| + \varepsilon \leq 3\varepsilon$. Using (8.1),

$$\mathcal{C}_\varepsilon(x) \leq \int_{\mathbb{T}^3} \frac{c_0}{|x-y|} \eta_\varepsilon(y) dy + C_1.$$

Because $|x| < 2\varepsilon$ and η_ε is supported in $\{|y| \leq \varepsilon\}$, every such y admits a representative in $[-1/2, 1/2]^3$ with $|y|_{\mathbb{R}^3} \leq \varepsilon$. Similarly, choose the representative of x in $[-1/2, 1/2]^3$, and note that $|x|_{\mathbb{R}^3} = |x| < 2\varepsilon$ implies $|x_i| < 2\varepsilon$ for each coordinate. Thus for each coordinate i ,

$$|x_i - y_i| \leq |x_i| + |y_i| < 2\varepsilon + \varepsilon = 3\varepsilon < \frac{1}{2},$$

since $\varepsilon < 1/6$. Hence in this case the torus distance equals the Euclidean distance between these representatives: $|x - y| = |x - y|_{\mathbb{R}^3}$. We may therefore lift the integral to \mathbb{R}^3 and substitute $y = \varepsilon u$:

$$\int_{|y| \leq \varepsilon} \frac{1}{|x - y|} \eta_\varepsilon(y) dy = \frac{1}{\varepsilon} \int_{\mathbb{R}^3} \frac{1}{|x/\varepsilon - u|} \eta(u) du.$$

The function $g(w) := \int_{\mathbb{R}^3} \frac{1}{|w - u|} \eta(u) du$ is the Newtonian potential of a smooth compactly supported function, hence finite and bounded on \mathbb{R}^3 . Let $M := \sup_{w \in \mathbb{R}^3} g(w) < \infty$. Then

$$\mathcal{C}_\varepsilon(x) \leq \frac{c_0 M}{\varepsilon} + C_1 \leq \frac{C}{\varepsilon} \leq \frac{3C}{|x| + \varepsilon}.$$

Combining both cases yields $\mathcal{C}_\varepsilon(x) \leq C/(|x| + \varepsilon)$. \square

9. A CRITICAL THIRD-CHAOS ESTIMATE

Lemma 9.1 (Logarithmic variance growth in third chaos). *Let X be the massive Gaussian free field on \mathbb{T}^3 with covariance \mathcal{C} . Let $X_\varepsilon := X * \rho_\varepsilon$ and define, for $\varphi \in C^\infty(\mathbb{T}^3)$,*

$$\mathcal{W}_\varepsilon(\varphi) := \int_{\mathbb{T}^3} \left(X_\varepsilon(x)^3 - 3\mathbb{E}[X_\varepsilon(x)^2] X_\varepsilon(x) \right) \varphi(x) dx.$$

Then for every $\varphi \in C^\infty(\mathbb{T}^3)$ there exists $C_\varphi < \infty$ such that for all $\varepsilon \in (0, 1/6)$,

$$\mathbb{E}[\mathcal{W}_\varepsilon(\varphi)^2] \leq C_\varphi \log(\varepsilon^{-1}).$$

Proof. Since $\mathcal{W}_\varepsilon(\varphi)$ is a third-homogeneous polynomial in a centred Gaussian field, Wick's theorem yields

$$(9.1) \quad \mathbb{E}[\mathcal{W}_\varepsilon(\varphi)^2] = 6 \int_{\mathbb{T}^3} \int_{\mathbb{T}^3} \varphi(x) \varphi(y) \mathcal{C}_\varepsilon(x-y)^3 dx dy,$$

see e.g. [7, Ch. 2]. Bounding $|\varphi(x)\varphi(y)| \leq \|\varphi\|_{L^\infty}^2$ and changing variables $z = x - y$ gives

$$\mathbb{E}[\mathcal{W}_\varepsilon(\varphi)^2] \leq 6\|\varphi\|_{L^\infty}^2 |\mathbb{T}^3| \int_{\mathbb{T}^3} \mathcal{C}_\varepsilon(z)^3 dz.$$

By Lemma 8.4, $\mathcal{C}_\varepsilon(z) \lesssim (|z| + \varepsilon)^{-1}$. Split \mathbb{T}^3 into $|z| \leq 1/4$ and $|z| > 1/4$. On $|z| > 1/4$, \mathcal{C}_ε is uniformly bounded in ε , hence the contribution is $O(1)$. On $|z| \leq 1/4$,

$$\int_{|z| \leq 1/4} \mathcal{C}_\varepsilon(z)^3 dz \lesssim \int_{|z| \leq 1/4} \frac{1}{(|z| + \varepsilon)^3} dz \asymp \int_0^{1/4} \frac{r^2}{(r + \varepsilon)^3} dr \lesssim \int_\varepsilon^{1/4} \frac{1}{r} dr \lesssim \log(\varepsilon^{-1}).$$

This proves the claim. \square

10. SEPARATING EVENT AND PROOF OF THEOREM 1.1

Fix $\psi \in C^\infty(\mathbb{T}^3) \setminus \{0\}$ and $\beta \in (1/2, 1)$. Let $F_n(\cdot; \psi)$ be defined by (5.3).

10.1. Vanishing under μ .

Lemma 10.1 (Vanishing in probability under μ). *Let $\Phi \sim \mu$. Then $F_n(\Phi; \psi) \rightarrow 0$ in probability as $n \rightarrow \infty$.*

Proof. By Proposition 7.1(ii),

$$\Phi_n^3 - 3a\varepsilon_n^{-1}\Phi_n - 9b\log(\varepsilon_n^{-1})\Phi = R_n + W_n,$$

where $\langle R_n, \psi \rangle$ is tight and $\mathbb{E}[\langle W_n, \psi \rangle^2] \lesssim \log(\varepsilon_n^{-1})$. Thus

$$F_n(\Phi; \psi) = e^{-\beta n} \langle R_n, \psi \rangle + e^{-\beta n} \langle W_n, \psi \rangle.$$

The first term tends to 0 in probability by Lemma 6.2. For the second term,

$$\mathbb{E}[|e^{-\beta n} \langle W_n, \psi \rangle|^2] \lesssim e^{-2\beta n} \log(\varepsilon_n^{-1}) = e^{-2\beta n} e^n = e^{-(2\beta-1)n} \rightarrow 0,$$

so it tends to 0 in L^2 , hence in probability. \square

10.2. A full- μ Borel set via deterministic subsequence. By Lemma 10.1 and Lemma 6.1, there exists a deterministic subsequence (n_k) such that $F_{n_k}(\Phi; \psi) \rightarrow 0$ almost surely. Define

$$A_\psi := \left\{ u \in E : \lim_{k \rightarrow \infty} F_{n_k}(u; \psi) = 0 \right\}.$$

Lemma 10.2. *The set A_ψ is Borel in E and satisfies $\mu(A_\psi) = 1$.*

Proof. Each $F_{n_k}(\cdot; \psi)$ is continuous, hence Borel. Therefore A_ψ is Borel since it is a countable Boolean combination of inverse images of closed intervals. By construction, $\Phi \in A_\psi$ almost surely. \square

10.3. Translation forces divergence. For $n \in \mathbb{N}$ write $\psi_n := \psi_{\varepsilon_n}$.

Lemma 10.3 (Uniform bounds for ψ_n). *For every $m \in \mathbb{N}$, $\psi_n \rightarrow \psi$ in $C^m(\mathbb{T}^3)$ and $\sup_n \|\psi_n\|_{C^m} < \infty$. Consequently, for every $r > 0$,*

$$\sup_n \|\psi_n \psi\|_{H^r} < \infty, \quad \sup_n \|\psi_n^2 \psi\|_{H^r} < \infty.$$

Proof. Standard properties of approximate identities yield convergence in C^m and uniform boundedness. The Sobolev bounds follow by $C^m \hookrightarrow H^r$ for large m and continuity of multiplication in C^m . \square

Lemma 10.4 (Tightness with varying tests). *Let (X_n) be tight in $H^{-r}(\mathbb{T}^3)$ for some $r > 0$, and let (φ_n) be deterministic with $\sup_n \|\varphi_n\|_{H^r} < \infty$. Then $(\langle X_n, \varphi_n \rangle)$ is tight in \mathbb{R} .*

Proof. By duality $|\langle X_n, \varphi_n \rangle| \leq \|X_n\|_{H^{-r}} \|\varphi_n\|_{H^r} \leq C \|X_n\|_{H^{-r}}$ with $C = \sup_n \|\varphi_n\|_{H^r}$. \square

Lemma 10.5 (Divergence under a smooth shift). *Let $\Phi \sim \mu$ and define $\tilde{\Phi} := \Phi - \psi$. Then $|F_n(\tilde{\Phi}; \psi)| \rightarrow \infty$ in probability as $n \rightarrow \infty$.*

Proof. Since $(\Phi - \psi)_n = \Phi_n - \psi_n$,

$$\begin{aligned} & (\Phi_n - \psi_n)^3 - 3a\varepsilon_n^{-1}(\Phi_n - \psi_n) - 9b\log(\varepsilon_n^{-1})(\Phi - \psi) \\ &= \left(\Phi_n^3 - 3a\varepsilon_n^{-1}\Phi_n - 9b\log(\varepsilon_n^{-1})\Phi \right) - 3\psi_n \left(\Phi_n^2 - a\varepsilon_n^{-1} \right) + 3\psi_n^2\Phi_n - \psi_n^3 + 9b\log(\varepsilon_n^{-1})\psi. \end{aligned}$$

Pair with ψ and multiply by $e^{-\beta n}$:

$$F_n(\tilde{\Phi}; \psi) = Y_n + 9b e^{-\beta n} \log(\varepsilon_n^{-1}) \|\psi\|_{L^2}^2,$$

where Y_n is the sum of the other four terms with prefactor $e^{-\beta n}$. Exactly as in the standard decomposition:

- $e^{-\beta n} \langle \Phi_n^3 - 3a\varepsilon_n^{-1}\Phi_n - 9b\log(\varepsilon_n^{-1})\Phi, \psi \rangle = F_n(\Phi; \psi) \rightarrow 0$ in probability (Lemma 10.1);
- $\Phi_n^2 - a\varepsilon_n^{-1}$ is tight in H^{-r} by Proposition 7.1(i), and $\psi_n \psi$ is uniformly bounded in H^r (Lemma 10.3), hence $e^{-\beta n} \langle \Phi_n^2 - a\varepsilon_n^{-1}, \psi_n \psi \rangle \rightarrow 0$ in probability by Lemmas 10.4 and 6.2;
- $\langle \Phi_n, \psi_n^2 \psi \rangle$ is tight (since $\Phi_n \rightarrow \Phi$ in H^{-s} a.s. and $\psi_n^2 \psi \rightarrow \psi^3$ in H^s with uniform bounds), hence $e^{-\beta n} \langle \Phi_n, \psi_n^2 \psi \rangle \rightarrow 0$ in probability by Lemma 6.2;

- $\langle \psi_n^3, \psi \rangle$ is bounded deterministically, hence $e^{-\beta n} \langle \psi_n^3, \psi \rangle \rightarrow 0$.

Thus $Y_n \rightarrow 0$ in probability. Using $\log(\varepsilon_n^{-1}) = e^n$ yields the deterministic divergence

$$9b e^{-\beta n} \log(\varepsilon_n^{-1}) \|\psi\|_{L^2}^2 = 9b e^{(1-\beta)n} \|\psi\|_{L^2}^2 \rightarrow \pm\infty,$$

since $b \neq 0$ (Lemma 8.2) and $\beta < 1$. Hence $|F_n(\tilde{\Phi}; \psi)| \rightarrow \infty$ in probability. \square

Lemma 10.6. *Let (X_n) be real random variables such that $|X_n| \rightarrow \infty$ in probability. Then for any deterministic increasing subsequence (n_k) ,*

$$\mathbb{P}(X_{n_k} \rightarrow 0) = 0.$$

Proof. Let $A_k := \{|X_{n_k}| \leq 1\}$. Then $\mathbb{P}(A_k) \rightarrow 0$. Moreover $\{X_{n_k} \rightarrow 0\} \subset \bigcup_{K=1}^{\infty} \bigcap_{k \geq K} A_k$. For each K , $\mathbb{P}(\bigcap_{k \geq K} A_k) \leq \inf_{k \geq K} \mathbb{P}(A_k)$, which tends to 0 as $K \rightarrow \infty$. \square

Theorem 10.7. *With A_ψ as above, one has $\mu(A_\psi + \psi) = 0$.*

Proof. Let $\Phi \sim \mu$ and set $\tilde{\Phi} = \Phi - \psi$. Then $\tilde{\Phi}$ has law $\mu(\cdot + \psi)$, so $\mu(A_\psi + \psi) = \mathbb{P}(\tilde{\Phi} \in A_\psi)$. On $\{\tilde{\Phi} \in A_\psi\}$ one has $F_{n_k}(\tilde{\Phi}; \psi) \rightarrow 0$ by definition of A_ψ . But by Lemma 10.5, $|F_n(\tilde{\Phi}; \psi)| \rightarrow \infty$ in probability, hence Lemma 10.6 gives $\mathbb{P}(F_{n_k}(\tilde{\Phi}; \psi) \rightarrow 0) = 0$. Thus $\mathbb{P}(\tilde{\Phi} \in A_\psi) = 0$. \square

Proof of Theorem 1.1. By Lemma 10.2, $\mu(A_\psi) = 1$. By Theorem 10.7, $\mu(A_\psi + \psi) = 0$. Lemma 3.1 yields $\mu \perp T_{\psi\#}\mu$. \square

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