

# 2D Coulomb Gases and the Renormalized Energy

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## Abstract

We study the statistical mechanics of classical two-dimensional “Coulomb gases” with general potential and arbitrary  $\beta$ , the inverse of the temperature. Such ensembles also correspond to random matrix models in some particular cases. The formal limit case  $\beta = \infty$  corresponds to “weighted Fekete sets” and also falls within our analysis.

It is known that in such a system points should be asymptotically distributed according to a macroscopic “equilibrium measure,” and that a large deviations principle holds for this, as proven by Ben Arous and Zeitouni [BZ].

By a suitable splitting of the Hamiltonian, we connect the problem to the “renormalized energy”  $W$ , a Coulombian interaction for points in the plane introduced in [SS1], which is expected to be a good way of measuring the disorder of an infinite configuration of points in the plane. By so doing, we are able to examine the situation at the microscopic scale, and obtain several new results: a next order asymptotic expansion of the partition function, estimates on the probability of fluctuation from the equilibrium measure at microscale, and a large deviations type result, which states that configurations above a certain threshold of  $W$  have exponentially small probability. When  $\beta \rightarrow \infty$ , the estimate becomes sharp, showing that the system has to “crystallize” to a minimizer of  $W$ . In the case of weighted Fekete sets, this corresponds to saying that these sets should microscopically look almost everywhere like minimizers of  $W$ , which are conjectured to be “Abrikosov” triangular lattices.

**keywords:** Coulomb gas, one-component plasma, random matrices, Ginibre ensemble, Fekete sets, Abrikosov lattice, triangular lattice, renormalized energy, large deviations, crystallization.

**MSC classification:** 82B05, 82D10, 82D99, 15B52

## 1 Introduction

We are interested in studying the probability law

$$(1.1) \quad d\mathbb{P}_n^\beta(x_1, \dots, x_n) = \frac{1}{Z_n^\beta} e^{-\frac{\beta}{2} w_n(x_1, \dots, x_n)} dx_1 \dots dx_n$$

where  $Z_n^\beta$  is the associated partition function (the normalizing factor such that  $\mathbb{P}_n^\beta$  is a probability measure) and

$$(1.2) \quad w_n(x_1, \dots, x_n) = - \sum_{i \neq j} \log |x_i - x_j| + n \sum_{i=1}^n V(x_i),$$

is the Hamiltonian. Here the  $x_i$ 's belong to  $\mathbb{R}^2$  (identified with the complex plane  $\mathbb{C}$ ),  $\beta > 0$  is a parameter corresponding to (the inverse of) the temperature and  $V$  is a potential satisfying some growth and regularity assumptions, which we will detail below.

The probability law  $\mathbb{P}_n^\beta$  is the Gibbs measure of what is called either a classical “two-dimensional Coulomb system” or “Coulomb gas” or “two-dimensional one-component plasma”, or sometimes “Gaussian  $\beta$ -ensemble” or Dyson gas. It was first pointed out by Wigner [Wi] and later exploited by Dyson [Dy], that Coulomb gases are naturally related to random matrices. This is somehow due to the fact that  $\exp(\sum_{i \neq j} \log |x_i - x_j|)$  is the square of the Vandermonde determinant  $\prod_{i < j} |x_i - x_j|$  and thus the law  $\mathbb{P}_n^\beta$ , in the particular case when  $V(x) = |x|^2$  and  $\beta = 2$  corresponds to the law of eigenvalues for the *Ginibre ensemble* (as shown in [Gin], see also [Me], Chap. 15), which is the set of matrices with independent normal (complex) Gaussian entries. For the general background and references to the literature, we refer to the book by Forrester [Fo]. These Coulomb gases and the Ginibre ensemble have been much studied, particularly from the random matrix point of view, but also from the statistical mechanics point of view, particularly relevant references in the physics literature are [AJ, SM, JLM].

The Gibbs measure  $\mathbb{P}_n^\beta$  can also be studied for  $x_i$ 's belonging to the real line (which we call the one-dimensional case). In the context of statistical mechanics (general  $\beta$ ), this corresponds to “log gases,” and in the context of random matrices ( $\beta = 1, 2, 4$ ), to Hermitian or symmetric random matrices (whose eigenvalues are always real). Even when  $\beta \notin \{1, 2, 4\}$  there is a corresponding random matrix model [DE], although it is more complicated. This one-dimensional case has been significantly more studied than the two-dimensional situation, and more can be achieved as well, in particular local statistics and “universal” spacings of eigenvalues have been established [VV, BEY1, BEY2]. This was only very recently extended to the two-dimensional case [BYY].

We also apply our methods and extend them to the one-dimensional setting, this is the object of our companion paper [SS4]. Let us finally also point out that studying  $\mathbb{P}_n^\beta$  with the  $x_i$ 's restricted to the unit circle and with  $\beta = 1, 2, 4$  also has a random matrix interpretation: it corresponds to the so-called circular ensembles, e.g. in the  $\beta = 2$  case, eigenvalues of the unitary matrices distributed according to the Haar measure, also a well-studied model. We also plan on examining this case in the future.

The current research on the random matrix aspect in the complex case focuses on studying the more general case of random matrices with entries that are not necessarily Gaussian and showing that the average behavior is the same as for the Ginibre ensemble, see [Ba, TV1, TV2]. We are instead limited to exact Vandermonde factors but we emphasize that our results are valid for all  $\beta$ , hence they are not limited to random matrices or the determinantal case and thus for the proof we cannot rely on any explicit random matrix model. Our results have some universality feature in the sense that they are valid for a large class of potentials  $V$  (only a few growth and regularity assumptions are made).

The function  $w_n$  can also be studied for its own sake: it can be seen as the interaction energy between similarly charged particles confined by the potential  $V$ . The case where  $V(x)$  is quadratic arises for instance as the interaction energy for superconducting vortices in the Ginzburg-Landau theory, in the regime where their number is fixed, bounded (see [SS2], Chapter 11). In the case (not treated here) without potential  $V$  but where the points are constrained to be on a manifold (such as the sphere) or a compact set, the minimizers of  $w_n$ , or maximizers of  $\prod_{i < j} |x_i - x_j|$ , are known as Fekete points or Fekete sets, cf. the book of Saff

and Totik [SaTo] for general reference. These are interesting in their own right – they arise mainly in polynomial interpolation – and the literature on the question of their distribution in various situations is vast. When instead  $V$  is a general smooth enough function (the situation we treat here), the minimizers of  $w_n$  are called *weighted Fekete points* or weighted Fekete sets, and are also of interest, cf. again [SaTo].

We will pursue the analysis of these weighted Fekete sets, which can be seen as the formal limit  $\beta \rightarrow +\infty$  of (1.1), in parallel with the analysis of (1.1) for general  $\beta$ , and obtain new results in both cases.

In the case of the Ginibre ensemble, i.e. when  $V(x) = |x|^2$  and  $\beta = 1$ , it is known that the “spectral measure”  $\nu_n := \frac{1}{n} \sum_{i=1}^n \delta_{x_i}$  converges to the uniform measure on the unit disc  $\frac{1}{\pi} \mathbf{1}_{B_1} dx$ . More precisely  $\mathbb{P}_n^\beta$ , seen as a probability on the space of probability measures on  $\mathbb{C}$  (the spectral measures) converges to a Dirac mass at  $\mu_0 = \frac{1}{\pi} \mathbf{1}_{B_1} dx$ . This is the celebrated “circular law”, attributed in this case to Ginibre, Mehta, an unpublished paper of Silverstein in 1984, and then Girko [Gir]. The large deviations from this law was established by Ben Arous and Zeitouni [BZ] (see Theorem 4 below): they showed that a large deviations principle holds with speed  $n^2$  and rate function

$$(1.3) \quad I(\mu) = \int_{\mathbb{R}^2 \times \mathbb{R}^2} -\log|x-y| d\mu(x) d\mu(y) + \int_{\mathbb{R}^2} |x|^2 d\mu(x)$$

whose unique minimizer among probabilities is of course the “circle law” distribution  $\frac{1}{\pi} \mathbf{1}_{B_1} dx$ .

For the case of a general  $V$  and a general  $\beta$ , the same large deviations principle holds with the rate function, analogue of (1.3), being

$$(1.4) \quad I(\mu) = \int_{\mathbb{R}^2 \times \mathbb{R}^2} -\log|x-y| d\mu(x) d\mu(y) + \int_{\mathbb{R}^2} V(x) d\mu(x).$$

This can be readapted from the proof of [BZ], otherwise it is proven in possibly higher complex dimensions in [Ber]. Again the spectral measure  $\nu_n = \frac{1}{n} \sum_{i=1}^n \delta_{x_i}$  converges to the minimizer among probability measures of  $I$ , called the *equilibrium measure*, which we will denote  $\mu_0$ . In the case of weighted Fekete sets, the analogue to the circular law has been known to be true for a much longer time: it was proved by Fekete, Polya and Szëgo that, for minimizers of (1.2),  $\frac{1}{n} \sum_{i=1}^n \delta_{x_i}$  converges to the same equilibrium measure minimizing  $I$  (then also referred to as the electrostatic interaction energy), whose description goes back to Gauss, and was carried out with modern mathematical rigor by Frostman [Fro].

Fekete sets on the sphere are probably the most studied among Fekete sets (cf. [SK]). By stereographic projection of the sphere onto the plane, they can be treated as (1.2) but with a weakly confining potential  $V(x) = \log(1 + |x|^2)$ , which barely fails to be in the class treated in this paper. The corresponding large deviations result (for the case with temperature) for such potentials was proven in [Ha].

We are interested in examining the “next order” behavior, or that of fluctuations around the limiting distribution  $\mu_0$ . Let us mention that such questions have already been addressed, often with the point of view of deriving explicit scaling limits (e.g. [BoSi, Gin]) or laws for certain statistics of fluctuations. One can see for example [AHM1, AHM2] where the authors essentially prove that the law of the linear statistics of the fluctuations is a Gaussian with specific variance and mean, or also [Rid] for related results. These results are however valid only for the “determinantal case”  $\beta = 2$ . Our approach and results are in some sense orthogonal, and again valid for all  $\beta$  and general  $V$ 's.

Recalling  $I$  and  $\mu_0$  are found through the large deviations at speed  $n^2$ , we look into the speed  $n$  and, while we do not prove a complete large deviations principle at this speed, we show there is still a sort of rate function for which a “threshold phenomenon” holds. This analysis is based on an expansion, through a crucial but simple “splitting formula” (which we present in Section 1.1 below) of  $w_n(x_1, \dots, x_n)$ , as equal to  $n^2 I(\mu_0) - \frac{n}{2} \log n$  plus a term of order  $n$ , whose prefactor tends as  $n \rightarrow +\infty$  to the “renormalized energy”  $W$ , a Coulombian interaction of points in the plane with a uniform neutralizing background, that we introduced in [SS1] and whose definition we will recall below in Section 1.2. To be more precise the limit term is the average value of  $W$  on the set of blow-up limits of the configuration of points  $x_1, \dots, x_n$  at the scale  $1/\sqrt{n}$ . It is this average that partially plays the role of a rate function at speed  $n$ . For a precise statement, see Theorem 5.

Another way of saying this is in the language of  $\Gamma$ -convergence (for a definition we refer to [Br, DM]), suffice it to know that this is the right notion of convergence to ensure that minimizers of  $w_n$  converge – via their empirical measures – to minimizers of  $I$ , i.e. to  $\mu_0$ ): it is not very difficult to show (for a short proof, see [SS2, Prop. 11.1] – the proof there is for  $V$  quadratic but works with no change for general  $V$ ) that  $\frac{w_n}{n^2}$   $\Gamma$ -converges as  $n \rightarrow \infty$  to  $I$ , defined in (1.4). Here we examine the next order in the “ $\Gamma$ -expansion” of  $w_n$ , i.e. we study the  $\Gamma$ -convergence of  $\frac{1}{n}(w_n - n^2 I(\mu_0) + \frac{n}{2} \log n)$ , and show that the  $\Gamma$ -limit is (the average of)  $W$ . Consequently, after blow-ups at scale  $\sqrt{n}$ , minimizers of  $w_n$  (i.e. weighted Fekete sets) should minimize the (average of the) renormalized energy  $W$ . For a precise statement, see Theorem 2.

Before yet giving a precise definition, let us mention that we introduced the renormalized energy  $W$  in [SS1] for the study of the interaction of vortices in the context of the Ginzburg-Landau energy of superconductivity (for general reference on the topic, cf. [SS2]). Configurations that minimize the Ginzburg-Landau energy with applied magnetic field, exhibit in certain regimes “point vortices” that are densely packed (there are  $n \gg 1$  of them) and are expected to arrange themselves in perfect triangular lattices (i.e. with  $60^\circ$  angles), named *Abrikosov lattices* after the physicist who predicted them [Ab]. The Abrikosov lattices are indeed observed in experiments on superconductors<sup>1</sup>. In [SS1] we made this partly rigorous by showing that minimizers of the Ginzburg-Landau energy have vortices that minimize the renormalized energy  $W$  after blow-up at the scale  $\sqrt{n}$ . The conjecture made in [SS1], also supported by some mathematical evidence (see Section 1.2), is then that the minimal value of  $W$  is achieved by the triangular lattice ; if proven true this would completely justify why vortices form these patterns. Combining this conjecture with the above conclusion that weighted Fekete sets should (after blow-up) minimize  $W$ , we thus obtain the conjecture that they also should locally form Abrikosov (triangular) lattices.

Does the same hold with finite temperature  $\beta$ , i.e. for Coulomb gases? Let us phrase the question more precisely. The law  $\mathbb{P}_n^\beta$  induces a probability measure on the family of blow-ups of  $(x_1, \dots, x_n)$  around a given origin point in  $\Sigma := \text{Supp}(\mu_0)$  — the parameter of the family — at the scale  $\sqrt{n}$ , a blow-up scale after which the resulting points are typically separated by order 1 distances. In the limit  $n \rightarrow \infty$  this yields a probability measure on the set of configurations of points in the plane and we may ask if, almost surely, the blow-up configurations minimize  $W$ . Our results indicate that this is not the case, however we are able to prove that there is a threshold phenomenon, in the sense that except with exponentially small probability, the average of  $W$  is below a certain constant, itself converging to the minimum of  $W$  as  $\beta \rightarrow \infty$ ,

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<sup>1</sup>For photos one can see <http://www.fys.uio.no/super/vortex/>

which indicates *crystallisation*, i.e. if the above conjecture is true, we should see Abrikosov lattices as  $\beta \rightarrow \infty$ .

To our knowledge, this is the first time Coulomb gases or Fekete sets are rigorously connected to triangular lattices, in agreement with predictions in the physics literature (see [AJ] and references therein).

A corollary of our way of expanding  $w_n$  is that we obtain a next order estimate of the partition function  $Z_n^\beta$ , a result we can already state:

**Theorem 1.** *Let  $V$  satisfy assumptions (1.11) – (1.14) below. There exist functions  $f_1, f_2$  depending only on  $V$ , such that for any  $\beta_0 > 0$  and any  $\beta \geq \beta_0$ , and for  $n$  larger than some  $n_0$  depending on  $\beta_0$ , we have*

$$(1.5) \quad n\beta f_1(\beta) \leq \log Z_n^\beta - \left( -\frac{\beta}{2}n^2 I(\mu_0) + \frac{\beta n}{4} \log n \right) \leq n\beta f_2(\beta),$$

with  $f_1, f_2$  bounded in  $[\beta_0, +\infty)$  and such that

$$(1.6) \quad \lim_{\beta \rightarrow \infty} f_1(\beta) = \lim_{\beta \rightarrow \infty} f_2(\beta) = -\frac{\alpha}{2}$$

where  $\alpha$  is some constant related to  $W$ , and explicited in (1.40) below.

This improves on the known results, which only gave the expansion  $\log Z_n^\beta \sim \frac{\beta}{2}n^2 I(\mu_0)$ . It also seems to contradict the result of the calculations of [ZW]. Let us recall that an exact value for  $Z_n^\beta$  is only known for the Ginibre ensemble case of  $\beta = 2$  and  $V(x) = |x|^2$ : it is  $Z_n^2 = n^{-\frac{1}{2}n(n+1)} \pi^n \prod_{k=1}^n k!$  (see [Me, Chap. 15]). Known asymptotics allow to deduce (cf. [Fo, eq. (4.184)]):

$$(1.7) \quad \log Z_n^2 = -\frac{3n^2}{4} + \frac{n}{2} \log n + n(-1 + \frac{1}{2} \log 2 + \frac{3}{2} \log \pi) + O(\log n) \quad \text{as } n \rightarrow \infty,$$

where we note the value of  $I(\mu_0)$  is indeed  $\frac{3}{4}$  for this potential. On the other hand, no exact formula exists for general potentials<sup>2</sup>, nor for quadratic potentials if  $\beta \neq 2$ . This is in contrast with the one-dimensional situation for which, at least in the case of quadratic  $V$ ,  $Z_n^\beta$  has an explicit expression for every  $\beta$ , given by the famous Selberg integral formulas (see e.g. [AGZ]).

In statistical mechanics language, the existence of an exact asymptotic expansion up to order  $n$  for  $\log Z_n^\beta$  is essentially the existence of a thermodynamic limit. This is established in [LN] for a three-dimensional Coulomb system, and in a nonrigorous way in [SM] in two dimensions. The existence of the thermodynamic limit here remains to be completed by getting upper and lower bounds which match up to  $o(n)$  in (1.5).

As suggested by the strong analogy between Coulomb gases and interacting vortices in the Ginzburg-Landau model, we will draw heavily on methods we introduced in [SS1], such as the splitting, blow-up, the use of the ergodic theorem, and the properties of the renormalized energy.

The rest of the introduction is organized as follows: first we give some more notation, give the assumptions we need to make on  $V$  and state the splitting formula, then we present the definition of the renormalized energy and the main results from [SS1] that we will use, and finally we state our main results and comment on them.

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<sup>2</sup>an exception is the result of [DGIL] for a quadrupole potential

## 1.1 The equilibrium measure and the splitting formula

We need to introduce some notation, and for this we need to describe the equilibrium measure  $\mu_0$  minimizing (1.4) among probability measures.

This description, which is now classical in potential theory (see [SaTo, Chap. 1]) says that, provided  $\lim_{|x| \rightarrow +\infty} \frac{V(x)}{2} - \log |x| = +\infty$  and  $\log V$  is lower semicontinuous, this equilibrium measure exists, is unique, and is characterized by the fact that there exists a constant  $c$  such that, quasi-everywhere,

$$(1.8) \quad U^{\mu_0} + \frac{V}{2} \geq c \quad \text{and} \quad U^{\mu_0} + \frac{V}{2} = c \quad \text{on } \text{Supp}(\mu_0),$$

where for any measure  $\mu$ ,  $U^\mu$  denotes the potential generated by  $\mu$ , defined by

$$(1.9) \quad U^\mu(x) = -2\pi\Delta^{-1}\mu := - \int_{\mathbb{R}^2} \log |x - y| d\mu(y).$$

Here and in all the paper, we denote by  $\Delta^{-1}$  the operator of convolution by  $\frac{1}{2\pi} \log |\cdot|$ . It is such that  $\Delta \circ \Delta^{-1} = Id$ , where  $\Delta$  is the usual Laplacian. We denote the support of  $\mu_0$  by  $\Sigma$ .

Another way to characterize  $U^{\mu_0}$  is as the solution of the following *obstacle problem*<sup>3</sup> : It is a superharmonic function bounded below by  $c - V/2$  and harmonic outside the so-called *coincidence set*

$$(1.10) \quad \Sigma = \{U^{\mu_0} = c - V/2\}.$$

This implies in particular that  $U^{\mu_0}$  is  $C^{1,1}$  if  $V$  is (see [Ca]).

It is now a good time to state the assumptions on  $V$  that we assume are satisfied in the sequel.

$$(1.11) \quad \lim_{|x| \rightarrow +\infty} \frac{V(x)}{2} - \log |x| = +\infty,$$

$$(1.12) \quad V \text{ is } C^3 \text{ and there exists } \underline{m}, \overline{m} > 0 \text{ s.t. } \underline{m} \leq \frac{\Delta V}{4\pi} \leq \overline{m},$$

$$(1.13) \quad V \text{ is such that } \partial\Sigma \text{ is } C^1,$$

$$(1.14) \quad \text{there exists } \beta_1 > 0 \text{ such that } \int_{\mathbb{R}^2 \setminus B_1} e^{-\beta_1(V/2(x) - \log |x|)} dx < +\infty.$$

The assumption (1.11) on the growth of  $V$  is what is needed to apply the results from [SaTo] and to guarantee that (1.4) has a minimizer. The other conditions are technical and certainly not optimal, they are meant to ensure that  $\mu_0$  and its support are regular enough and that  $\mu_0$  never degenerates, which we will need for example when making explicit constructions.

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<sup>3</sup>The obstacle problem is a free-boundary problem and a much-studied classical problem in the calculus of variations, for general reference see [Fri, KS].

Indeed, assumptions (1.12)–(1.13), together with (1.8), (1.9) and the regularity of  $V$ , ensure that

$$(1.15) \quad d\mu_0 = m_0(x) dx, \quad \text{where} \quad m_0(x) = \frac{\Delta V(x)}{4\pi} \mathbf{1}_\Sigma(x).$$

hence for the  $\underline{m}, \bar{m} > 0$  of (1.12) we have

$$(1.16) \quad \underline{m} \leq m_0 \leq \bar{m}.$$

Assumption (1.14) is a supplementary assumption on the growth of  $V$  at infinity, needed for the case with temperature, to show that the partition function is well-defined. It only requires slightly more than (1.11), for example assuming  $\lim_{|x| \rightarrow \infty} \frac{V}{2}(x) - (1+\varepsilon) \log |x| = +\infty$  for some  $\varepsilon > 0$  suffices.

Next, we set  $\zeta = U^{\mu_0} + \frac{V}{2} - c$  where  $c$  is the constant in (1.8) and (1.10). This function satisfies

$$(1.17) \quad \begin{cases} \Delta \zeta = \frac{1}{2} \Delta V \mathbf{1}_{\mathbb{R}^2 \setminus \Sigma} \\ \zeta = 0 \\ \zeta > 0 \end{cases} \quad \begin{array}{l} \text{quasi-everywhere in } \Sigma \\ \text{quasi-everywhere in } \mathbb{R}^2 \setminus \Sigma \end{array}$$

From our assumption (1.12), it follows from [Ca, Lemma 5] that there exists  $\kappa > 0$  such that for every  $x \in \mathbb{R}^2$ ,

$$(1.18) \quad \zeta(x) \geq \kappa \operatorname{dist}(x, \Sigma)^2,$$

and such a rate is in fact optimal [Ca, Lemma 2]. All the quantities introduced so far:  $\mu_0$ ,  $\Sigma$ ,  $\zeta$ , depend only on the data of  $V$ .

The function  $\zeta$  arises in the splitting formula for  $w_n$  which we now present. As mentioned above, expanding the probability density to the next order goes along with blowing-up the point configuration by a factor  $\sqrt{n}$ . We then denote the blown-up quantities by primes. For example the *blown-up coordinates*  $x'_i = \sqrt{n}x_i$ ,  $m_0'(x') = m_0(x)$ ,  $d\mu'_0(x') = m_0'(x')dx'$  etc . . . .

The splitting formula, proven in Section 2, is the observation that, for any  $x_1, \dots, x_n \in \mathbb{R}^2$ , we have

$$(1.19) \quad w_n(x_1, \dots, x_n) = n^2 I(\mu_0) - \frac{n}{2} \log n + \frac{1}{\pi} W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2}) + 2n \sum_{i=1}^n \zeta(x_i),$$

where

$$(1.20) \quad H'_n := -2\pi \Delta^{-1} \left( \sum_{i=1}^n \delta_{x'_i} - \mu'_0 \right),$$

and where, in agreement with formula (1.29) below (the existence of the limit as  $\eta \rightarrow 0$  will also be discussed there),

$$(1.21) \quad W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2}) := \lim_{\eta \rightarrow 0} \left( \frac{1}{2} \int_{\mathbb{R}^2 \setminus \cup_{i=1}^n B(x'_i, \eta)} |\nabla H'_n|^2 + \pi n \log \eta \right).$$

The function  $H'_n$  physically corresponds to the electrostatic potential generated by the positive point charges  $\sum_i \delta_{x'_i}$  and the diffuse negative charge  $-\mu'_0$ . Its opposite gradient, that we will

denote by  $E_n$  physically corresponds to the electric field generated by the charges (hence the notation).

Letting, for a measure  $\nu$

$$(1.22) \quad F_n(\nu) := \begin{cases} \frac{1}{n} \left( \frac{1}{\pi} W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2}) + 2n \int \zeta d\nu \right) & \text{if } \nu \text{ is of the form } \sum_{i=1}^n \delta_{x_i} \\ +\infty & \text{otherwise,} \end{cases}$$

the relation (1.19) can be rewritten

$$(1.23) \quad w_n(x_1, \dots, x_n) = n^2 I(\mu_0) - \frac{n}{2} \log n + n F_n \left( \sum_{i=1}^n \delta_{x_i} \right).$$

This allows to separate orders as announced since we will see that  $F_n(\sum_{i=1}^n \delta_{x_i})$  is typically of order 1.

We may next cancel out leading order terms and rewrite the probability law (1.1) as

$$(1.24) \quad d\mathbb{P}_n^\beta(x_1, \dots, x_n) = \frac{1}{K_n^\beta} e^{-n \frac{\beta}{2} F_n(\sum_i \delta_{x_i})} dx_1 \dots dx_n$$

where

$$(1.25) \quad K_n^\beta := Z_n^\beta e^{\frac{\beta}{2}(n^2 I(\mu_0) - \frac{n}{2} \log n)}.$$

As we will see below  $\log K_n^\beta$  is of order  $n\beta$ , which leads to Theorem 1.

We will also denote

$$(1.26) \quad \widehat{F}_n(\nu) = F_n(\nu) - 2 \int \zeta d\nu = \frac{1}{n\pi} W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2}).$$

In view of (1.19) the main task in our proof is to pass to the limit  $n \rightarrow \infty$  in  $W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2})$  and obtain a limiting energy, which will be (the average of) our Coulomb renormalized energy  $W$ . Passing to the limit in (1.20) will lead to solutions of  $-\Delta H = 2\pi \left( \sum_p \delta_p - cste \right)$  where the sum is now infinite. The limit energy thus has to be defined on objects of this form, or equivalently (by taking  $E = -\nabla H$ ) solutions of  $\operatorname{div} E = 2\pi \left( \sum_p \delta_p - cste \right)$ ,  $\operatorname{curl} E = 0$ . The definition will be given just below. The passage to the limit is not obvious, for several reasons. The first is the lack of local charge neutrality, and the fact that the energy density associated to  $W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2})$  is not pointwise bounded below. The second is the need of the ‘‘averaged formulation’’ alluded to above, this will be provided by an abstract method relying on the ergodic theorem, and inspired by Varadhan.

## 1.2 The renormalized energy

We now define precisely the ‘‘renormalized energy’’  $W$  introduced in [SS1], which is a way of computing the Coulomb interaction between an infinite number of point charges in the plane with a uniform neutralizing background of density  $m$ . We point out that, to our knowledge, each of the analogous Coulomb systems studied in the physics literature (e.g. [SM, AJ]) comprise a finite number of point charges, and hence implicitly extend only to a



bounded domain on which there is charge neutrality. Here we do not assume any local charge neutrality.

We denote by  $B(x, R)$  or  $B_R(x)$  the ball centered at  $x$  with radius  $R$  and let  $B_R = B(0, R)$ . In all the paper, when  $U$  is a measurable set,  $|U|$  will denote its Lebesgue measure, and when  $U$  is a finite set,  $\#U$  will denote its cardinal.  $f$  will denote an integral average.

The point of the definition of  $W$  below is that we would like to define  $W(\nabla H)$  for  $H$  solving  $-\Delta H = 2\pi(\sum_p \delta_p - m)$  as  $\limsup_{R \rightarrow \infty} \int_{B_R} |\nabla H|^2$ , however these integrals diverge because of the logarithmic divergence of  $H$  near each point. Instead, we compute  $\int |\nabla H|^2$  in a “renormalized” way or in “finite parts”, by cutting out holes around each  $p$  and subtracting off the corresponding divergence, in the manner of [BBH], from which the name “renormalized energy” is borrowed.

**Definition 1.1.** *Let  $m$  be a nonnegative number. For any continuous function  $\chi$  and any vector-field  $E$  in  $\mathbb{R}^2$  such that*

$$(1.27) \quad \operatorname{div} E = 2\pi(\nu - m), \quad \operatorname{curl} E = 0$$

where  $\nu$  has the form

$$(1.28) \quad \nu = \sum_{p \in \Lambda} \delta_p \quad \text{for some discrete set } \Lambda \subset \mathbb{R}^2,$$

we let

$$(1.29) \quad W(E, \chi) = \lim_{\eta \rightarrow 0} \left( \frac{1}{2} \int_{\mathbb{R}^2 \setminus \cup_{p \in \Lambda} B(p, \eta)} \chi |E|^2 + \pi \log \eta \sum_{p \in \Lambda} \chi(p) \right).$$

To see that the limit  $\eta \rightarrow 0$  exists, it suffices to observe that in view of (1.27)–(1.28),  $E$  is a gradient and near each  $p \in \Lambda$  we may write  $E = \nabla \log |\cdot - p| + \nabla f(\cdot)$  where  $f$  is  $C^1$  by elliptic regularity. The limit follows easily. It also follows that  $E$  belongs to  $L^q_{\text{loc}}$  for any  $q < 2$ .

**Definition 1.2.** *Let  $m$  be a nonnegative number. Let  $E$  be a vector field in  $\mathbb{R}^2$ . We say  $E$  belongs to the admissible class  $\mathcal{A}_m$  if (1.27), (1.28) hold and*

$$(1.30) \quad \frac{\nu(B_R)}{|B_R|} \quad \text{is bounded by a constant independent of } R > 1.$$

In the sequel  $K_R$  will denote the two-dimensional squares  $[-R, R]^2$ . We also use the notation  $\chi_{K_R}$  for positive cutoff functions satisfying, for some constant  $C$  independent of  $R$ ,

$$(1.31) \quad |\nabla \chi_{K_R}| \leq C, \quad \operatorname{Supp}(\chi_{K_R}) \subset K_R, \quad \chi_{K_R}(x) = 1 \text{ if } d(x, K_R^c) \geq 1.$$

**Definition 1.3.** *The renormalized energy  $W$  is defined, for  $E \in \mathcal{A}_m$ , by*

$$(1.32) \quad W(E) = \limsup_{R \rightarrow \infty} \frac{W(E, \chi_{K_R})}{|K_R|},$$

with  $\{\chi_{K_R}\}_R$  satisfying (1.31).

We note that we have taken a slightly different definition from [SS1]: first the vector-fields in (1.27) have been rotated by  $\pi/2$ , second  $\mathcal{A}_m$  here corresponds to  $\mathcal{A}_{2\pi m}$  in [SS1]. Finally, in [SS1] we presented the definition with averages over general sets, here we have chosen for simplicity to introduce it only with square averages.

In theory, many different  $E$ 's could correspond to a given  $\nu$  (one can always add the gradient of a harmonic function). But as it turns out, they only differ by a constant:

**Lemma 1.4.** *Let  $m \geq 0$  and  $\nu = \sum_{p \in \Lambda} \delta_p$ , where  $\Lambda \subset \mathbb{R}^2$  is discrete, and assume there exists  $E$  such that*

$$(1.33) \quad \operatorname{div} E = 2\pi(\nu - m), \quad \operatorname{curl} E = 0, \quad \text{and} \quad W(E) < +\infty.$$

*Then any other  $E'$  satisfying (1.33) is such that  $E - E'$  is constant.*

*If there exists  $E$  such that (1.33) holds and such that*

$$(1.34) \quad \lim_{R \rightarrow \infty} \int_{K_R} E = 0,$$

*then any other  $E'$  satisfying (1.33) is such that  $W(E') > W(E)$ .*

*Proof.* Let  $E, E'$  be as above. We may view them as complex functions of a complex variable. From (1.33) we have  $\operatorname{div} (E - E') = \operatorname{curl} (E - E') = 0$  and thus  $E - E'$  is holomorphic. We can write it as a power series  $\sum_{n=0}^{\infty} a_n z^n$  with infinite radius of convergence. On the other hand, from the finiteness of  $W(E)$  and  $W(E')$  we deduce easily that there exists  $C > 0$  such that

$$(1.35) \quad \forall R > 1, \quad \int_{K_R} |E - E'|^2 \leq CR^2.$$

But by Cauchy's formula we have, for any  $R > 0$  and  $t \in [R, R+1]$

$$a_n = \frac{1}{2i\pi} \int_{\partial B(0,t)} \frac{(E - E')(z)}{z^{n+1}} dz = \frac{1}{2i\pi} \int_R^{R+1} \int_{\partial B(0,t)} \frac{(E - E')(z)}{z^{n+1}} dz dt.$$

It follows that

$$|a_n| \leq \frac{1}{2\pi R^{n+1}} \int_{B(0,R+1) \setminus B(0,R)} |E - E'| \leq \frac{C}{R^{n+1}} R^{3/2}$$

where we have used the Cauchy-Schwarz inequality and (1.35). Letting  $R \rightarrow \infty$  we find that  $a_n = 0$  for any  $n \geq 1$  and thus  $E - E'$  is constant. For the second statement, we deduce from the first statement that  $E' = E + \vec{C}$  for some constant vector  $\vec{C} \neq 0$ , and then

$$W(E', \chi_{K_R}) = W(E, \chi_{K_R}) + \vec{C} \cdot \int E \chi_{K_R} + \frac{|\vec{C}|^2}{2} \int \chi_{K_R},$$

so that dividing by  $|K_R|$ , passing to the limit as  $R \rightarrow +\infty$  and in view of (1.34), we find  $W(E') = W(E) + \frac{1}{2}|\vec{C}|^2$ .  $\square$

Note that given  $\nu$ , the above lemma shows that either for all  $E$ 's satisfying (1.33) the limit  $\lim_{R \rightarrow \infty} \int_{K_R} E$  exists, or it exists for none of them. Both cases may occur.

The following additional facts and remarks about  $W$  are mostly from [SS1]:

- In [SS1], we introduced  $W$  as being computed with averages over general shapes (say balls, squares etc). We showed that the minimum of  $W$  over  $\mathcal{A}_m$  does not depend on the shape used. Since squares are the most useful ones, we restricted to them here for the sake of simplicity.
- It was shown in [SS1, Theorem 1] that the value of  $W$  does not depend on the choice of  $\{\chi_{K_R}\}_R$  as long as it satisfies (1.31).
- $W$  is bounded below and admits a minimizer over  $\mathcal{A}_1$ , cf. [SS1, Theorem 1].
- It is easy to check that if  $E$  belongs to  $\mathcal{A}_m$ ,  $m > 0$ , then  $E' = \frac{1}{\sqrt{m}}E(\cdot/\sqrt{m})$  belongs to  $\mathcal{A}_1$  and

$$(1.36) \quad W(E) = m \left( W(E') - \frac{\pi}{2} \log m \right).$$

Consequently if  $E$  is a minimizer of  $W$  over  $\mathcal{A}_m$ , then  $E'$  minimizes  $W$  over  $\mathcal{A}_1$ . In particular

$$(1.37) \quad \min_{\mathcal{A}_m} W = m \left( \min_{\mathcal{A}_1} W - \frac{\pi}{2} \log m \right).$$

- Because the number of points is in general infinite, the interaction over large balls needs to be normalized by the volume, as in a thermodynamic limit. Thus  $W$  does not feel compact perturbations of the configuration of points. Even though the interactions are long-range, this is not difficult to justify rigorously.
- In [GS] some necessary and some sufficient conditions on the configuration of points for which  $W(E) < \infty$  are given.
- We may define  $W$  as a function of the point measure  $\nu$  only, by setting for every  $\nu$  satisfying (1.28)

$$(1.38) \quad \mathbb{W}(\nu) = \inf_{E \text{ such that (1.27) holds}} W(E),$$

and  $\mathbb{W}(\nu) = +\infty$  if  $\nu$  is not of the form  $\sum_{p \in \Lambda} \delta_p$ . This definition is somehow “relaxed” since  $\mathbb{W}(\nu) \leq W(E)$  for any  $E$  satisfying (1.27). The main point to check is the measurability of  $\mathbb{W}$ , which we will discuss below in Section 6.6.

- In the case  $m = 1$  and when the set of points  $\Lambda$  is periodic with respect to some lattice  $\mathbb{Z}\vec{u} + \mathbb{Z}\vec{v}$  then it can be viewed as a set of  $n$  points  $a_1, \dots, a_n$  over the torus  $\mathbb{T}_{(\vec{u}, \vec{v})} := \mathbb{R}^2 / (\mathbb{Z}\vec{u} + \mathbb{Z}\vec{v})$  with  $|\mathbb{T}_{(\vec{u}, \vec{v})}| = n$ . In this case, the infimum of  $W(E)$  among  $E$ 's which satisfy (1.33) is achieved by  $E_{\{a_i\}} = -\nabla h$ , where  $h$  is the periodic solution to  $-\Delta h = 2\pi(\sum_i \delta_{a_i} - 1)$ , and

$$(1.39) \quad W(E_{\{a_i\}}) = \frac{\pi}{|\mathbb{T}_{(\vec{u}, \vec{v})}|} \sum_{i \neq j} G(a_i - a_j) + \pi \lim_{x \rightarrow 0} (G(x) + \log |x|)$$

where  $G$  is the Green function of the torus with respect to its volume form, i.e. the solution to

$$-\Delta G(x) = 2\pi \left( \delta_0 - \frac{1}{|\mathbb{T}_{(\vec{u}, \vec{v})}|} \right) \quad \text{in } \mathbb{T}_{(\vec{u}, \vec{v})}.$$

An explicit expression for  $G$  can be found via Fourier series and this leads to an explicit expression for  $W$  of the form  $\sum_{i \neq j} E(a_i - a_j)$  where  $E$  is an Eisenstein series (for more details see [SS1, Lemma 1.3] and also [BoSe]). In this periodic setting, the expression of  $W$  is thus much simpler than (1.32) and reduces to the computation of a sum of explicit pairwise interaction.

- When the set of points  $\Lambda$  is itself exactly a lattice  $\mathbb{Z}\vec{u} + \mathbb{Z}\vec{v}$  then  $W$  can be expressed explicitly through the Epstein Zeta function of the lattice. Moreover, using results from number theory, it is proved in [SS1, Theorem 2], that the unique minimizer of  $W$  over lattice configurations of fixed volume is the triangular lattice. This supports the conjecture that the Abrikosov triangular lattice is a global minimizer of  $W$ , with a slight abuse of language since  $W$  is here not a function of the points, but of their associated “electric fields”  $E_{\{a_i\}}$ .

This last fact allows us to think of  $W$  as a way of measuring the disorder and lack of homogeneity of a configuration of points in the plane (this point of view is pursued in [BoSe] with explicit computations for random point processes). Another way to see it is to view  $W$  as measuring the distance between  $\sum_{p \in \Lambda} \delta_p$  and the constant  $m$  in  $H^{-1}$ , the dual space to the Sobolev space  $H_0^1$  (with  $\|f\|_{H_0^1} = \|\nabla f\|_{L^2}$ ) which only makes sense modulo the “renormalization” as  $\eta \rightarrow 0$  and modulo normalizing by the volume.

We may now define the constant  $\alpha$  which appears in Theorem 1 and in Theorem 2 below:

$$(1.40) \quad \alpha := \frac{1}{\pi} \int_{\Sigma} \min_{\mathcal{A}_{m_0(x)}} W \, dx = \frac{1}{\pi} \min_{\mathcal{A}_1} W - \frac{1}{2} \int_{\Sigma} m_0(x) \log m_0(x) \, dx,$$

where we have used (1.37) and the fact that, from (1.15),  $\int_{\Sigma} m_0 = 1$ . Note that  $\alpha$  only depends on  $V$ , via the integral term, and on the (so far) unknown constant  $\min_{\mathcal{A}_1} W$ .

### 1.3 Statement of main results

Our first result identifies the  $\Gamma$ -limit of  $\{F_n\}_n$ , defined in (1.22) or (1.23). This in particular allows a description at the microscopic level of the weighted Fekete sets minimizing  $w_n$ . Below we abuse notation by writing  $\nu_n = \sum_{i=1}^n \delta_{x_i}$  when it should be  $\nu_n = \sum_{i=1}^n \delta_{x_{i,n}}$ . For such a  $\nu_n$ , we let  $\nu'_n = \sum_{i=1}^n \delta_{x'_i}$  be the measure in blown-up coordinates and  $E_n = -\nabla H'_n$  be the associated electric field, where  $H'_n$  is defined by (1.20) — equivalently  $E_n$  is the solution of  $\text{curl } E_n = 0$ ,  $\text{div } E_n = 2\pi(\nu'_n - m'_0 dx')$  in  $\mathbb{R}^2$  which tends to 0 at infinity. (To avoid confusion, we emphasize here that  $\nu_n$  lives at the original scale while  $E_n$  lives at the blown-up scale and that  $m'_0$  is the blown-up density of the equilibrium measure  $\mu_0$ .) We also let

$$(1.41) \quad P_{\nu_n} = \int_{\Sigma} \delta_{(x, E_n(\sqrt{n}x + \cdot))} \, dx,$$

i.e. the push-forward of the normalized Lebesgue measure on  $\Sigma$  by  $x \mapsto (x, E_n(\sqrt{n}x + \cdot))$ . It is a probability measure on  $X := \Sigma \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  (couples of (blow-up centers, blown-up current around this center)). We emphasize that  $P_{\nu_n}$  is a probability measure which has nothing to do with  $\mathbb{P}_n^\beta$ . Each realization or configuration  $(x_1, \dots, x_n)$  gives rise in a deterministic fashion to its  $P_{\nu_n}$ , which encodes all the blown-up profiles of associated electric fields. We denote by  $i_n$  this mapping (or embedding)

$$(1.42) \quad \begin{aligned} i_n : \mathbb{C}^n &\rightarrow \mathcal{P}(X) \\ (x_1, \dots, x_n) &\mapsto P_{\nu_n} \end{aligned}$$

where  $\mathcal{P}(X)$  denotes the space of probability measures on  $X = \Sigma \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$ . We view  $\mathcal{P}(X)$  as endowed with the topology of weak convergence of probabilities.

The limiting object as  $n \rightarrow +\infty$  in the  $\Gamma$ -limit of  $w_n$  was the limit of  $\frac{\nu_n}{n}$ . In taking the  $\Gamma$ -limit of  $F_n$ , the limiting object is more complex, it is the limit  $P$  of  $P_{\nu_n}$ . This is a probability measure on all blown-up electric fields obtained from a given  $(x_1, \dots, x_n)$ . Thus it is like a Young measure akin to the Young measures on micropatterns introduced in [AM].

We will here and below use the notation

$$(1.43) \quad D(x', R) = \nu_n \left( B \left( x, \frac{R}{\sqrt{n}} \right) \right) - n\mu_0 \left( B \left( x, \frac{R}{\sqrt{n}} \right) \right),$$

where  $x' = \sqrt{n}x$  as usual, to denote the fluctuations of the number of points in a microscopic ball of radius  $R$ . Note that  $\widehat{F}_n$  was defined in (1.26) and the result below is slightly stronger than the  $\Gamma$ -convergence of  $F_n$  since  $F_n \geq \widehat{F}_n$ .

**Theorem 2** (Microscopic behavior of weighted Fekete sets). *Let the potential  $V$  satisfy assumptions (1.11)–(1.13). Let  $m_0$  be the density of the equilibrium measure  $\mu_0$ . Fix from now on  $1 < p < 2$  and let  $X = \Sigma \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$ .*

**A. Lower bound.** *Let  $\nu_n = \sum_{i=1}^n \delta_{x_i}$  be a sequence such that  $\widehat{F}_n(\nu_n) \leq C$ . Then  $P_{\nu_n}$  defined by (1.41) is a probability measure on  $X$  and*

1. *Any subsequence of  $\{P_{\nu_n}\}_n$  has a convergent subsequence converging to some  $P \in \mathcal{P}(X)$  as  $n \rightarrow \infty$ .*
2. *The first marginal of  $P$  is the normalized Lebesgue measure on  $\Sigma$ .  $P$  is invariant by  $(x, E) \mapsto (x, E(\lambda(x) + \cdot))$ , for any  $\lambda(x)$  of class  $C^1$  from  $\Sigma$  to  $\mathbb{R}^2$  (we will say  $T_{\lambda(x)}$ -invariant).*
3. *For  $P$  almost every  $(x, E)$  we have  $E \in \mathcal{A}_{m_0(x)}$ .*
4. *Defining  $\alpha$  as in (1.40), it holds that*

$$(1.44) \quad \liminf_{n \rightarrow \infty} \widehat{F}_n(\nu_n) \geq \frac{|\Sigma|}{\pi} \int W(E) dP(x, E) \geq \alpha.$$

**B. Upper bound construction.** *Conversely, assume  $P$  is a  $T_{\lambda(x)}$ -invariant probability measure on  $X$  whose first marginal is  $\frac{1}{|\Sigma|} dx|_{\Sigma}$  and such that for  $P$ -almost every  $(x, E)$  we have  $E \in \mathcal{A}_{m_0(x)}$ . Then there exists a sequence  $\{\nu_n = \sum_{i=1}^n \delta_{x_i}\}_n$  of empirical measures on  $\Sigma$  and a sequence  $\{E_n\}_n$  in  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  such that  $\text{div } E_n = 2\pi(\nu'_n - m_0')$  and such that defining  $P_n$  as in (1.41), we have  $P_n \rightarrow P$  as  $n \rightarrow \infty$  and*

$$(1.45) \quad \limsup_{n \rightarrow \infty} F_n(\nu_n) \leq \frac{|\Sigma|}{\pi} \int W(E) dP(x, E).$$

**C. Consequences for minimizers.** *If  $(x_1, \dots, x_n)$  minimizes  $w_n$  for every  $n$  and  $\nu_n = \sum_{i=1}^n \delta_{x_i}$ , then the limit  $P$  of  $P_{\nu_n}$  as defined in (1.41) satisfies the following.*

1. *For  $P$ -almost every  $(x, E)$ ,  $E$  minimizes  $W$  over  $\mathcal{A}_{m_0(x)}$ .*

2. We have

$$\lim_{n \rightarrow \infty} F_n(\nu_n) = \lim_{n \rightarrow \infty} \widehat{F}_n(\nu_n) = \frac{|\Sigma|}{\pi} \int W(E) dP(x, E) = \alpha, \quad \lim_{n \rightarrow \infty} \sum_{i=1}^n \text{dist}^2(x_i, \Sigma) = 0.$$

3. There exists  $C > 0$  such that for every  $x' \in \mathbb{R}^2$ ,  $R > 1$  and using the notation (1.43) we have

$$(1.46) \quad D(x', R)^2 \min \left( 1, \frac{|D(x', R)|}{R^2} \right) \leq Cn.$$

Note that part B of the theorem is only a partial converse to part A because the constructed  $E_n$  need not be curl free.

**Remark 1.5.** Defining  $Q_{\nu_n} = \int_{\Sigma} \delta_{x, \nu'_n(\sqrt{n}x + \cdot)} dx$ , or equivalently as the push-forward of  $P_{\nu_n}$  by the map  $(x, E) \mapsto \frac{1}{2\pi} \text{div } E + m'_0(\sqrt{n}x + \cdot) dx'$ , we can also express this limiting result in terms of the limit  $Q$  to  $Q_{\nu_n}$ , which is the push-forward of  $P$  by  $(x, E) \mapsto \frac{1}{2\pi} \text{div } E + m_0(x)$ . The limiting energy for both the upper bound and the lower bound is then

$$\frac{|\Sigma|}{\pi} \int \mathbb{W}(\nu) dQ(x, \nu).$$

Of course such a statement is a bit weaker than Theorem 2 since some information is lost: namely we do not keep the information of which  $E$  corresponded to  $\nu$ .

This theorem is the analogue of the main result of [SS1] but for  $w_n$  rather than the Ginzburg-Landau energy. It is technically simpler to prove, except for the possibility of a nonconstant weight  $m_0(x)$  which was absent from [SS1]. It can be stated as the fact that  $\frac{|E|}{\pi} \int W dP$ , which can be understood as the average of  $W$  with respect to all possible blow-up centers in  $\Sigma$  (chosen uniformly at random), is the  $\Gamma$ -limit of  $w_n$  at next order. Its minimum over all admissible probabilities is  $\alpha$ .

The estimate (1.46) gives a control on the “discrepancy”  $D$  (between the effective number of points and the expected one) at the scale  $R/\sqrt{n}$ . Note that in a recent paper [AOC], the authors also study the fine behavior of weighted Fekete sets. Using completely different methods, based on Beurling-Landau densities and techniques going back to [La], they are able to show the very strong result that

$$\limsup_{R \rightarrow \infty} \limsup_{n \rightarrow \infty} \frac{D(x_n, R)}{R^2} = 0,$$

as long as  $\text{dist}(x_n, \partial\Sigma') \geq \log^2 n$ . This shows that the density of points follows  $\mu_0$  at the microscopic scale  $1/\sqrt{n}$  and thus the configurations are very rigid. This still leaves however some uncertainty about the patterns they should follow. On the contrary, our result is less precise about  $D(x, R)$  since we only recover the optimal estimate when  $R$  grows faster than  $n^{1/4}$ , but it connects the pattern formed by the points to the Abrikosov triangular lattice via the minimization of  $W$ .

We now turn to Coulomb gases, i.e. to the case with temperature. It is straightforward from the form (1.24) and the estimate (provided by Theorem 1)  $\log K_n^\beta = O(n\beta)$  where  $K_n^\beta$  is defined in (1.25), to deduce that  $F_n \leq C$  except on a set of small probability, because  $F_n$  controls the deviation between  $\nu_n$  and  $n\mu_0$  and controls  $W$ . This fact allows to derive various consequences, the first being estimates on the probability of certain rare events.

**Theorem 3.** *Let  $V$  satisfy assumptions (1.11)–(1.14).*

*There exists a universal constant  $R_0 > 0$  and  $c, C > 0$  depending only on  $V$  such that: For any  $\beta_0 > 0$ , any  $n$  large enough depending on  $\beta_0$ , and any  $\beta > \beta_0$ , for any  $x_1, \dots, x_n \in \mathbb{R}^2$ , any  $R > R_0$ , any  $x'_0 = \sqrt{n}x_0 \in \mathbb{R}^2$  and any  $\eta > 0$ , letting  $\nu_n = \sum_{i=1}^n \delta_{x_i}$ , we have the following:*

$$(1.47) \quad \log \mathbb{P}_n^\beta (|D(x'_0, R)| \geq \eta R^2) \leq -c\beta \min(\eta^2, \eta^3)R^4 + C\beta(R^2 + n) + Cn.$$

$$(1.48) \quad \log \mathbb{P}_n^\beta \left( \int \zeta d\nu_n \geq \eta \right) \leq -\frac{1}{2}n\beta\eta + Cn(\beta + 1).$$

Moreover, for any smooth bounded  $U' = \sqrt{n}U \subset \mathbb{R}^2$ ,

$$(1.49) \quad \log \mathbb{P}_n^\beta \left( \int_{U'} \frac{D(x', R)^2}{R^2} \min \left( 1, \frac{|D(x', R)|}{R^2} \right) dx' \geq \eta \right) \leq n\beta(-c\eta + C|U| + C) + Cn.$$

Finally, if  $q \in [1, 2)$  there exists  $c, C > 0$  depending on  $V$  and  $q$  such that  $\forall \eta \geq 1, R > 0$ ,

$$(1.50) \quad \log \mathbb{P}_n^\beta \left( \left( 1 + \frac{R^2}{n} \right)^{\frac{1}{2} - \frac{1}{q}} \|\nu - n\mu_0\|_{W^{-1,q}(B_{R/\sqrt{n}})} \geq \eta\sqrt{n} \right) \leq -cn\beta\eta^2 + Cn(\beta + 1),$$

where  $W^{-1,q}(\Omega)$  is the dual of the Sobolev space  $W_0^{1,p}(\Omega)$  with  $\frac{1}{p} + \frac{1}{q} = 1$ ; in particular  $W^{-1,1}$  is the dual of Lipschitz functions.

These estimates can roughly be read in the following way: as soon as  $\eta$  is large enough, the events in parentheses have probability decaying like  $e^{-cn}$ . More precisely, we bound the probability that a ball contains too many or too few points compared to the expected number  $n\mu_0(B)$ , but whereas the circular law does it for a *macroscopic ball*, i.e. for  $R$  comparable to  $\sqrt{n}$ , the estimate (1.47) is effective at intermediate scales, of the order of  $n^{1/4}$ . This is sometimes called in this context “undercrowding” or “overcrowding” of points, see [JLM, NSV, Kri]. In view of similar results in [JLM] and the result of [AOC], we can expect this to hold as soon as  $R \gg 1$ , but this seems out of reach by our method. This can also be compared with analogous estimates without error terms proven in the case of Hermitian matrices, cf. [ESY]. These results, in the Hermitian case, are proven in the general setting of Wigner matrices, i.e. Hermitian matrices with random i.i.d. entries, which do not need to be Gaussian. They only concern some fixed  $\beta$  however.

The estimate (1.49) gives a global version of this result: it expresses a control on the average microscopic “discrepancy”  $D$ . This control is in  $L^2$  for large values of the discrepancy, and in  $L^3$  for small values. The estimate (1.48) allows, in view of (1.18), to control (again with some threshold to be beaten) the probability that some points may be far from the set  $\Sigma$ . Note that since  $\nu_n$  is a non-normalized empirical measure, (1.48) ensures for example that the probability that a single point lies at a distance  $\eta$  from  $\Sigma$  is exponentially small as soon as  $\eta$  is larger than some constant. All these estimates rely on controlling  $D$  by  $F_n$ .

Finally, (1.50) tells us that fluctuations around the law  $n\mu_0$  can be globally controlled (take for example  $R = \sqrt{n}$ ) by  $O(\sqrt{n})$  (except with exponentially small probability). We believe this estimate to be optimal. Its proof uses in a crucial manner the result of [STi], which controls, via Lorentz spaces, the difference  $\nu_n - n\mu_0$  in terms of  $W$  or  $\widehat{F}_n$ .

Our last result mostly expresses Theorem 2 in a “moderate” deviations language. Before stating it, let us recall for comparison the result of [BZ]:

**Theorem 4** (Ben Arous - Zeitouni). Let  $\beta = 2$  and  $V(x) = |x|^2$ . Denote by  $\tilde{\mathbb{P}}_n^\beta$  the image of the law (1.1) by the map  $(x_1, \dots, x_n) \mapsto \nu_n$ , where  $\nu_n = \frac{1}{n} \sum_{i=1}^n \delta_{x_i}$ . Then for any subset  $A$  of the set of probability measures on  $\mathbb{R}^2$  (endowed with the topology of weak convergence), we have

$$-\inf_{\mu \in \mathring{A}} \tilde{I}(\mu) \leq \liminf_{n \rightarrow \infty} \frac{1}{n^2} \log \tilde{\mathbb{P}}_n^\beta(A) \leq \limsup_{n \rightarrow \infty} \frac{1}{n^2} \log \tilde{\mathbb{P}}_n^\beta(A) \leq -\inf_{\mu \in A} \tilde{I}(\mu),$$

where  $\tilde{I} = I - \min I$ .

**Theorem 5.** Let  $V$  satisfy assumptions (1.11)–(1.14). For any  $\beta > 0$ , the following holds. For any  $n > 0$  let  $A_n \subset \mathbb{C}^n$ . Denote

$$(1.51) \quad A_\infty = \bigcap_{n>0} \overline{\bigcup_{m>n} i_m(A_m)},$$

where  $i$  is as in (1.42), and the topology is the weak convergence on  $\mathcal{P}(X)$ . Then for any  $\eta > 0$  there is  $C_\eta > 0$  depending on  $V$  and  $\eta$  only such that  $\alpha$  being as in (1.40),

$$(1.52) \quad \limsup_{n \rightarrow \infty} \frac{\log \mathbb{P}_n^\beta(A_n)}{n} \leq -\frac{\beta}{2} \left( \frac{|\Sigma|}{\pi} \inf_{P \in A_\infty} \int W(E) dP(x, E) - \alpha - \eta - \frac{C_\eta}{\beta} \right).$$

Conversely, let  $A \subset \mathcal{P}(X)$  be a set of  $T_{\lambda(x)}$ -invariant probability measures on  $X$  and let  $\mathring{A}$  be the interior of  $A$ . Then for any  $\eta > 0$ , there exists a sequence of subsets  $A_n \subset \Sigma^n$  such that

$$(1.53) \quad -\frac{\beta}{2} \left( \frac{|\Sigma|}{\pi} \inf_{P \in \mathring{A}} \int W(E) dP(x, E) - \alpha + \eta + \frac{C_\eta}{\beta} \right) \leq \liminf_{n \rightarrow \infty} \frac{\log \mathbb{P}_n^\beta(A_n)}{n},$$

and such that for any sequence  $\{\nu_n = \sum_{i=1}^n \delta_{x_i}\}_n$  such that  $(x_1, \dots, x_n) \in A_n$  for every  $n$  there exists a sequence of fields  $E_n \in L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  such that  $\text{div } E_n = 2\pi(\nu'_n - m'_0)$  and such that — defining  $P_n$  as in (1.41) with  $E_n$  replacing  $E_{\nu_n}$  — we have

$$(1.54) \quad \lim_n P_n \in \mathring{A}.$$

Note that if  $P_n$  was  $P_{\nu_n}$ , then (1.54) would be equivalent to saying that  $\bigcap_n \overline{\bigcup_{m>n} i_m(A_m)} \subset \mathring{A}$ . The difference between  $P_{\nu_n}$  and  $P_n$  is that the latter is generated by a field  $E_n$  which is not necessarily a gradient.

Compared to Theorem 4 this result can be seen as a next order (speed  $n$  instead of  $n^2$ ) deviations result, where the average of  $W$  over blow-up centers plays the role of a rate function, with a margin which becomes small as  $\beta \rightarrow \infty$ . While Theorem 4 said that empirical measures at macroscopic scale converge to  $\mu_0$ , except for a set of exponentially decaying probability, Theorem 5 says that within the empirical measures which do converge to  $\mu_0$ , the ones with large average of  $W$  (computed after blow-up) also have exponentially decaying probability, but at the slower rate  $e^{-n}$  instead of  $e^{-n^2}$ . More precisely, there is a threshold  $C/\beta$  for some  $C > 0$ , such that configurations satisfying

$$\frac{|\Sigma|}{\pi} \int W dP \geq \alpha + \frac{C}{\beta}$$

have exponentially small probability, where we recall  $\alpha$  is also the minimum possible value of  $\frac{|\Sigma|}{\pi} \int W dP$ . Since we believe that  $W$  measures the disorder of a (limit) configuration of



(blown up) points in the plane, this means that most configurations have a certain order. The threshold, or gap,  $C/\beta$  tends to 0 as  $\beta$  tends to  $\infty$ , hence in this limit, configurations have to be closer and closer to the minimum of the average of  $W$ , or have more and more order.

Modulo the conjecture that the minimum of  $W$  is achieved by the perfect “Abrikosov” triangular lattice, this constitutes a crystallisation result. Note that to solve this conjecture, it would suffice to evaluate  $\alpha$ , which in view of Theorem 1 is equivalent to being able to compute the asymptotics of  $Z_n^\beta$  as  $\beta \rightarrow \infty$ .

At nonzero temperature, the probabilities  $P$  are not expected to be concentrated on minimizers of  $W$ , indeed numerical simulations of the Ginibre ensemble<sup>4</sup> corresponding to  $\beta = 2$  show patterns of points with a certain microscopic disorder, which are certainly not crystalline. This is probably explained by the fact that at finite temperature, and in this order  $n$ , an entropy term should come to compete with the minimization of  $W$ . One may wonder if at least there exists a limiting law on the probabilities  $P_n$ , and which it is. The following theorem answers positively the question of the existence:

**Theorem 6.** *For each integer  $n$ , and a given  $\beta > 0$ , let  $\widetilde{\mathbb{P}}_n^\beta$  denote the push-forward of  $\mathbb{P}_n^\beta$  by  $i_n$  defined in (1.42). It is an element of  $\mathcal{P}(\mathcal{P}(X))$ . Then  $\{\widetilde{\mathbb{P}}_n^\beta\}_n$  is tight and converges, up to a subsequence, to a probability measure  $\widetilde{\mathbb{P}}^\beta$  on  $\mathcal{P}(X)$ .*

This shows the existence of a limiting “electric field process”.

**Remark 1.6.** *Pushing forward by  $(x, E) \mapsto \frac{1}{2\pi} \operatorname{div} E + m'_0(\sqrt{n}x + \cdot)$  as in Remark 1.5 gives the existence of a limiting point process  $\widetilde{\mathbb{Q}}^\beta$  i.e. a probability on the limiting  $Q$ 's, which themselves encode all the  $(x, \nu)$ 's.*

To conclude the following open questions naturally arise in view of our results, and are closely related to one another:

- Prove that  $\min_{\mathcal{A}_1} W$  is achieved by the triangular lattice.
- Find whether a large deviations statement is true at speed  $n$ , and if it is, find the rate function.
- Characterize the limiting processes  $\widetilde{\mathbb{P}}^\beta$  and  $\widetilde{\mathbb{Q}}^\beta$ .

In [SS4] we show that all the results we have obtained here are also true in the case of points on the real line, i.e. for 1D log gases or Hermitian random matrices. There the minimization of  $W$  is solved (the minimum is the perfect lattice  $\mathbb{Z}$ ) and the crystallisation result is complete.

The rest of the paper is organized as follows: Section 2 contains the proof of the “splitting formula”. In Section 3, we present the “spreading result” from [SS1] and some first corollaries. In Section 4, we present an explicit construction which yields the lower bound on  $Z_n^\beta$ , whose proof is postponed to Section 7. In Section 5, we show how  $W$  controls the overcrowding/undercrowding of points, and prove Theorem 3. In Section 6 we present the ergodic averaging approach (the abstract result) and apply it to conclude the proofs of Theorem 2, 5 and 1.

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<sup>4</sup>cf. e.g. Benedek Valko’s webpage <http://www.math.wisc.edu/valko/courses/833/833.html>

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## 2 Proof of the splitting formula

The connection between  $w_n$  and  $W$  originates in the following computation

**Lemma 2.1.** *For any  $x_1, \dots, x_n$  and letting  $\nu_n = \sum_{i=1}^n \delta_{x_i}$  the following holds*

$$(2.1) \quad F_n(\nu) = \frac{1}{n\pi} W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2}) + 2 \sum_{i=1}^n \zeta(x_i) \\ = \frac{1}{n} \left( w_n(x_1, \dots, x_n) - n^2 I(\mu_0) + \frac{n}{2} \log n \right),$$

where  $F_n$  is defined in (1.22),  $W$  is defined in (1.29), and  $H'_n$  is defined in (1.20).

*Proof.* Let  $\nu_n = \sum_{i=1}^n \delta_{x_i}$ , and let  $H_n$  be defined by

$$H_n = -2\pi \Delta^{-1} (\nu_n - n\mu_0).$$

First we note that since  $\nu_n$  and  $n\mu_0$  have same mass and compact support we have  $H_n(x) = O(1/|x|)$  and  $\nabla H_n(x) = O(1/|x|^2)$  as  $|x| \rightarrow +\infty$ .

We prove that, denoting by  $D$  the diagonal in  $\mathbb{R}^2 \times \mathbb{R}^2$ , we have

$$(2.2) \quad \int_{(\mathbb{R}^2 \times \mathbb{R}^2) \setminus D} -\log|x-y| d(\nu_n - n\mu_0)(x) d(\nu_n - n\mu_0)(y) = \frac{1}{\pi} W(\nabla H_n, \mathbf{1}_{\mathbb{R}^2}).$$

First, using Green's formula, we have

$$(2.3) \quad \int_{B_R \setminus \cup_{i=1}^n B(x_i, \eta)} |\nabla H_n|^2 = \int_{\partial B_R} H_n \nabla H_n \cdot \vec{\nu} + \sum_{i=1}^n \int_{\partial B(x_i, \eta)} H_n \nabla H_n \cdot \vec{\nu} \\ + 2\pi \int_{B_R \setminus \cup_{i=1}^n B(x_i, \eta)} H_n d(\nu_n - n\mu_0).$$

Here, and in all the paper,  $\vec{\nu}$  denotes the outer unit normal vector.

Let  $H^i(x) := H_n(x) + \log|x-x_i|$ . We have  $H^i = -\log*(\nu^i - n\mu_0)$ , with  $\nu^i = \nu_n - \delta_{x_i}$ , and near  $x_i$ ,  $H^i$  is  $C^1$ . Therefore, using (1.20) and the boundedness of  $m_0$  in  $L^\infty$ , we have that, as  $\eta \rightarrow 0$

$$\int_{\partial B(x_i, \eta)} H_n \cdot \vec{\nu} = -2\pi \log \eta + 2\pi H^i(x_i) + o(1),$$

while the integral on  $\partial B_R$  tends to 0 as  $R \rightarrow +\infty$  from the decay properties of  $H_n$ . We thus obtain, as  $\eta \rightarrow 0$  and  $R \rightarrow +\infty$ ,

$$\int_{B_R \setminus \cup_i B(x_i, \eta)} |\nabla H_n|^2 = -2\pi n \log \eta + 2\pi \sum_{i=1}^n H^i(x_i) - 2\pi n \int_{\mathbb{R}^2} H_n d\mu_0 + o(1),$$

and therefore, by definition of  $W$ ,

$$(2.4) \quad W(\nabla H_n, \mathbf{1}_{\mathbb{R}^2}) = \pi \sum_{i=1}^n H^i(x_i) - \pi n \int H_n d\mu_0.$$

Second we note that

$$\int_{\mathbb{R}^2 \setminus \{x_i\}} -\log |x_i - y| d(\nu_n - n\mu_0)(y) = H^i(x_i),$$

and if  $x \notin \{x_i\}$  then

$$\int_{\mathbb{R}^2 \setminus \{x\}} -\log |x - y| d(\nu_n - n\mu_0)(y) = H_n(x).$$

It follows that

$$\int_{D^c} -\log |x - y| d(\nu_n - n\mu_0)(x) d(\nu_n - n\mu_0)(y) = \sum_{i=1}^n H^i(x_i) - n \int_{\mathbb{R}^2} H_n(x) d\mu_0(x),$$

which together with (2.4) proves (2.2).

On the other hand, we may rewrite  $w_n$  as

$$w_n(x_1, \dots, x_n) = \int_{D^c} -\log |x - y| d\nu_n(x) d\nu_n(y) + n \int V(x) d\nu_n(x)$$

and, splitting  $\nu_n$  as  $n\mu_0 + \nu_n - n\mu_0$  and using the fact that  $\mu_0 \times \mu_0(D) = 0$ , we obtain

$$\begin{aligned} w(x_1, \dots, x_n) &= n^2 I(\mu_0) + 2n \int U^{\mu_0}(x) d(\nu_n - n\mu_0)(x) + n \int V(x) d(\nu_n - n\mu_0)(x) \\ &\quad + \int_{D^c} -\log |x - y| d(\nu_n - n\mu_0)(x) d(\nu_n - n\mu_0)(y). \end{aligned}$$

Since  $U^{\mu_0} + \frac{V}{2} = c + \zeta$  and since  $\nu_n$  and  $n\mu_0$  have same mass  $n$ , we have

$$2n \int U^{\mu_0}(x) d(\nu_n - n\mu_0)(x) + n \int V(x) d(\nu_n - n\mu_0)(x) = 2n \int \zeta d(\nu_n - \mu_0) = 2n \int \zeta d\nu_n,$$

using the fact that  $\zeta = 0$  on the support of  $\mu_0$ . Therefore, in view of (2.2) we have found

$$(2.5) \quad w(x_1, \dots, x_n) = n^2 I(\mu_0) + 2n \int \zeta d\nu_n + \frac{1}{\pi} W(\nabla H_n, \mathbf{1}_{\mathbb{R}^2}).$$

But, changing variables, we find

$$\frac{1}{2} \int_{\mathbb{R}^2 \setminus \cup_{i=1}^n B(x_i, \eta)} |\nabla H_n|^2 = \frac{1}{2} \int_{\mathbb{R}^2 \setminus \cup_{i=1}^n B(x'_i, \sqrt{n}\eta)} |\nabla H'_n|^2,$$

and by adding  $\pi n \log \eta$  on both sides and letting  $\eta \rightarrow 0$  we deduce that  $W(\nabla H_n, \mathbf{1}_{\mathbb{R}^2}) = W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2}) - \frac{\pi}{2} n \log n$ . Together with (2.5) this proves (2.1).  $\square$

### 3 A first lower bound on $F_n$ and upper bound on $Z_n^\beta$

The crucial fact that we now wish to exploit is that, even though  $W(E, \chi)$  or rather its associated energy density does not have a sign, there are good lower bounds for  $F_n$ . This follows from the analysis of [SS1], more specifically from the following “mass spreading result”, adapted from [SS1], Proposition 4.9 and Remark 4.10 (with slightly different notation), which itself is based on the so-called “ball construction method”, a crucial tool in the analysis of Ginzburg-Landau equations. This result, that we will use here as a black box, says that even though the energy density associated to  $W(E, \chi)$  is not positive (or even bounded below), it can be replaced by an energy-density  $g$  which is uniformly bounded below, at the expense of a negligible error.

For any set  $\Omega$ ,  $\widehat{\Omega}$  denotes its 1-tubular neighborhood, i.e.  $\{x \in \mathbb{R}^2, \text{dist}(x, \Omega) < 1\}$ .

**Proposition 3.1.** *Assume  $\Omega \subset \mathbb{R}^2$  is open and  $(\nu, E)$  are such that  $\nu = 2\pi \sum_{p \in \Lambda} \delta_p$  for some finite subset  $\Lambda$  of  $\widehat{\Omega}$  and  $\text{div } E = 2\pi(\nu - a(x)dx)$ ,  $\text{curl } E = 0$  in  $\widehat{\Omega}$ , where  $a \in L^\infty(\widehat{\Omega})$ . Then, given any  $\rho > 0$  there exists a measure  $g$  supported on  $\widehat{\Omega}$  and such that*

- *there exists a family  $\mathcal{B}_\rho$  of disjoint closed balls covering  $\text{Supp}(\nu)$ , with the sum of the radii of the balls in  $\mathcal{B}_\rho$  intersecting with any ball of radius 1 bounded by  $\rho$ , and such that*

$$(3.1) \quad g \geq -C(\|a\|_{L^\infty} + 1) + \frac{1}{4}|E|^2 \mathbf{1}_{\Omega \setminus \mathcal{B}_\rho} \quad \text{in } \widehat{\Omega},$$

where  $C$  depends only on  $\rho$ .

-

$$(3.2) \quad g = \frac{1}{2}|E|^2 \quad \text{outside } \cup_{p \in \Lambda} B(p, \lambda)$$

where  $\lambda$  depends only on  $\rho$ .

- *For any function  $\chi$  compactly supported in  $\Omega$  we have*

$$(3.3) \quad \left| W(E, \chi) - \int \chi dg \right| \leq CN(\log N + \|a\|_{L^\infty}) \|\nabla \chi\|_\infty,$$

where  $N = \#\{p \in \Lambda : B(p, \lambda) \cap \text{Supp}(\nabla \chi) \neq \emptyset\}$  for some  $\lambda$  and  $C$  depending only on  $\rho$ .

- *For any  $U \subset \Omega$ ,*

$$(3.4) \quad \#(\Lambda \cap U) \leq C \left( 1 + \|a\|_{L^\infty}^2 |\widehat{U}| + g(\widehat{U}) \right).$$

Note that the result in [SS1] is not stated for any  $\rho$  but a careful inspection of the proof there allows to show that it can be readapted to make  $\rho$  arbitrarily small. From now on, we take some  $\rho < 1/8$ .

**Definition 3.2.** *Assume  $\nu_n = \sum_{i=1}^n \delta_{x_i}$ . Letting  $\nu'_n = \sum_{i=1}^n \delta_{x'_i}$  be the measure in blown-up coordinates and  $E_{\nu_n} = -\nabla H'_n$ , where  $H'_n$  is defined by (1.20), we denote by  $g_{\nu_n}$  the result of applying the previous proposition to  $(\nu'_n, E_{\nu_n})$  in  $\mathbb{R}^2$ .*

Even though we will not use the following result in the sequel, we state it to show how we can quickly derive a first upper bound on  $Z_n^\beta$  from what precedes.

**Proposition 3.3.** *We have*

$$(3.5) \quad \log K_n^\beta \leq Cn\beta + n(\log |\Sigma| + o(1))$$

where we recall  $\Sigma = \text{Supp}(\mu_0)$ , and

$$(3.6) \quad \log Z_n^\beta \leq -\frac{\beta}{2}n^2 I(\mu_0) + \frac{\beta n}{4} \log n + Cn\beta + n(\log |\Sigma| + o(1))$$

where  $o(1) \rightarrow 0$  as  $n \rightarrow \infty$  uniformly with respect to  $\beta > \beta_0$ , for any  $\beta_0 > 0$ , and  $C$  depends only on  $V$ .

The proof uses two lemmas.

**Lemma 3.4.** *For any  $\nu_n = \sum_{i=1}^n \delta_{x_i}$ , we have*

$$(3.7) \quad F_n(\nu_n) = \frac{1}{n\pi} \int_{\mathbb{R}^2} dg_{\nu_n} + 2 \int \zeta d\nu_n,$$

where  $F_n$  is as in (1.22).

*Proof.* This follows from (3.3) applied to  $\chi_{K_R}$ , where  $\chi_{B_R}$  is as in (1.31). If  $R$  is large enough then,  $\lambda$  being the constant of Proposition 3.1,  $\#\{p \in \text{Supp}(\nu_n) : B(p, \lambda) \cap \text{Supp}(\nabla \chi_{K_R}) \neq \emptyset\} = 0$  and therefore (3.3) reads

$$W(E_{\nu_n}, \chi_{K_R}) = \int \chi_{K_R} dg_{\nu_n}.$$

Letting  $R \rightarrow +\infty$  yields  $W(E_{\nu_n}, \mathbf{1}_{\mathbb{R}^2}) = \int dg_{\nu_n}$  and the result, in view of (1.22).  $\square$

**Lemma 3.5.** *Letting  $\nu_n$  stand for  $\sum_{i=1}^n \delta_{x_i}$  we have, for any constant  $\gamma > 0$  and uniformly w.r.t.  $\beta$  greater than any arbitrary positive constant  $\beta_0$ , we have*

$$(3.8) \quad \lim_{n \rightarrow \infty} \left( \int_{\mathbb{C}^n} e^{-\gamma\beta n \int \zeta d\nu_n} dx_1 \dots dx_n \right)^{\frac{1}{n}} = |\Sigma|.$$

*Proof.* This is the place where we use the assumption (1.14). We recall that  $\zeta = U^{\mu_0} + \frac{V}{2} - c$ , and note that since  $\mu_0$  is a compactly supported probability measure  $U^{\mu_0}(x) = -\int \log|x-y| d\mu_0(y)$  behaves asymptotically like  $-\log|x|$  when  $|x| \rightarrow \infty$ , more precisely one can easily show that there exists  $C$  such that  $|U^{\mu_0}(x) + \log|x|| \leq C$  for  $|x|$  large enough. It thus follows that  $\zeta(x) \geq -\log|x| + \frac{V}{2}(x) - C$  for  $|x|$  large enough, and in view of (1.14), this implies that for some  $\beta_2 > 0$ ,  $\int_{\mathbb{C}} e^{-\beta_2 \zeta(x)} dx$  converges.

Next, by separation of variables, we have

$$\int_{\mathbb{C}^n} e^{-\gamma\beta n \sum_{i=1}^n \zeta(x_i)} dx_1 \dots dx_n = \left( \int_{\mathbb{C}} e^{-\gamma\beta n \zeta(x)} dx \right)^n$$

On the other hand, we have  $\zeta \geq 0$  and  $\{\zeta = 0\} = \Sigma$  by (1.17), hence we have  $e^{-\gamma\beta n \zeta(x)} \rightarrow \mathbf{1}_\Sigma$  pointwise, as  $\beta n \rightarrow \infty$ . In addition, if  $\beta \geq \beta_0 > 0$ , for  $n$  large enough depending on  $\beta_0$ ,  $e^{-\gamma\beta n \zeta(x)}$  is dominated by  $e^{-\beta_2 \zeta(x)}$  which is integrable. Therefore, by dominated convergence theorem, it follows that (3.8) holds uniformly w.r.t.  $\beta \geq \beta_0$ , for any  $\beta_0 > 0$ .  $\square$

*Proof of Proposition 3.3.* Let again  $\nu_n$  stand for  $\sum_{i=1}^n \delta_{x_i}$ . From (3.2) we have  $g_{\nu_n} \geq 0$  outside  $\cup_i B(x_i, \lambda)$  and from (3.1) we have  $g_{\nu_n} \geq -C$  (depending only on  $\|m_0\|_{L^\infty}$  hence on  $V$ ) in  $\cup_i B(x_i, \lambda)$ . Inserting into (3.7) we deduce that

$$F_n(\nu_n) \geq -C + 2 \int \zeta d\nu_n,$$

where  $C$  depends only on  $V$ . Inserting into (1.24) and integrating over  $\mathbb{C}^n$ , we find

$$1 \leq \frac{1}{K_n^\beta} e^{Cn\beta} \int_{\mathbb{C}^n} e^{-n\beta \int \zeta d\nu_n} dx_1 \dots dx_n.$$

Inserting (3.8) and taking logarithms, it follows that

$$\log K_n^\beta \leq Cn\beta + n(\log |\Sigma| + o(1)).$$

The relation (3.6) follows using (1.25). □

## 4 A construction and a lower bound for $Z_n^\beta$

In this section, we construct a set of explicit configurations whose  $W$  is not too large, and show that their probability is not too small, which will lead to a lower bound on  $Z_n^\beta$ . This is the longest part of our proof. The method is borrowed from [SS1] but requires various adjustments that we shall detail in Section 7. We will need (1.13) in order to simplify the construction and estimates near the boundary.

The following proves Theorem 2, part B and contains a bit more information useful for proving Theorem 5.

**Proposition 4.1.** *Let  $P$  be a  $T_{\lambda(x)}$ -invariant probability measure on  $X = \Sigma \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  with first marginal  $dx_{|\Sigma|/|\Sigma|}$  and such that for  $P$  almost every  $(x, E)$  we have  $E \in \mathcal{A}_{m_0(x)}$ . Then, for any  $\eta > 0$ , there exists  $\delta > 0$  and for any  $n$  a subset  $A_n \subset \mathbb{C}^n$  such that  $|A_n| \geq n!(\pi\delta^2/n)^n$  and for every sequence  $\{\nu_n = \sum_{i=1}^n \delta_{y_i}\}_n$  with  $(y_1, \dots, y_n) \in A_n$  the following holds.*

*i) We have the upper bound*

$$(4.1) \quad \limsup_{n \rightarrow \infty} \frac{1}{n} \left( w_n(y_1, \dots, y_n) - n^2 I(\mu_0) + \frac{n}{2} \log n \right) \leq \frac{|\Sigma|}{\pi} \int W(E) dP(x, E) + \eta.$$

*ii) There exists  $\{E_n\}_n$  in  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  such that  $\text{div } E_n = 2\pi(\nu'_n - m_0')$  and such that the image  $P_n$  of  $dx_{|\Sigma|/|\Sigma|}$  by the map  $x \mapsto (x, E_n(\sqrt{n}x + \cdot))$  is such that*

$$(4.2) \quad \limsup_{n \rightarrow \infty} \text{dist}(P_n, P) \leq \eta,$$

where  $\text{dist}$  is a distance which metrizes the topology of weak convergence on  $\mathcal{P}(X)$ .

Respectively, the image  $Q_n$  of  $dx_{|\Sigma|/|\Sigma|}$  by the map  $x \mapsto (x, \nu'_n(\sqrt{n}x + \cdot))$  is such that

$$(4.3) \quad \limsup_{n \rightarrow \infty} \text{dist}(Q_n, Q) \leq \eta,$$

where  $\text{dist}$  is a distance which metrizes the topology of weak convergence, and  $Q$  is the push-forward of  $P$  by  $(x, E) \mapsto \frac{1}{2\pi} \text{div } E + m'_0(x)$ .

Applying the above proposition with  $\eta = 1/k$  we get a subset  $A_{n,k}$  in which we choose any  $n$ -tuple  $(y_{i,k})_{1 \leq i \leq n}$ . This yields in turn a family  $\{P_{n,k}\}$  of probability measures on  $X$ . A standard diagonal extraction argument then yields

**Corollary 4.2** (Theorem 2, Part B). *Under the same assumptions as Proposition 4.1, there exists a sequence  $\{\nu_n = \sum_{i=1}^n \delta_{x_i}\}_n$  and a sequence  $\{E_n\}_n$  in  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  such that  $\text{div } E_n = 2\pi(\nu'_n - m_0'(x') dx')$  and*

$$(4.4) \quad \limsup_{n \rightarrow \infty} \frac{1}{n} \left( w_n(x_1, \dots, x_n) - n^2 I(\mu_0) + \frac{n}{2} \log n \right) \leq \frac{|\Sigma|}{\pi} \int W(E) dP(x, E).$$

Moreover, denoting  $P_n$  the image of  $dx_{|\Sigma|/|\Sigma|}$  by the map  $x \mapsto (x, E_n(\sqrt{n}x + \cdot))$ , we have  $P_n \rightarrow P$  as  $n \rightarrow +\infty$ .

Another consequence of Proposition 4.1 is, recalling (1.40) and (1.25):

**Corollary 4.3** (Lower bound part of Theorem 1). *For any  $\eta > 0$  there exists  $C_\eta > 0$  such that for any  $\beta > 0$  we have*

$$(4.5) \quad \liminf_{n \rightarrow +\infty} \frac{\log K_n^\beta}{n} \geq -\frac{\beta}{2}(\alpha + \eta) - C_\eta.$$

*Proof of the corollary.* Choose  $E_0 \in \mathcal{A}_1$  to be a minimizer for  $\min_{\mathcal{A}_1} W$ , which exists by [SS1, Theorem 1], and let  $P$  be the image of the normalized Lebesgue measure on  $\Sigma$  by the map  $x \mapsto (x, \sigma_{m_0(x)} E_0)$ , where

$$(4.6) \quad \sigma_m E(y) := \sqrt{m} E(\sqrt{m}y).$$

Then by construction  $P$ -almost every  $(x, E)$  satisfies  $E \in \mathcal{A}_{m_0(x)}$  and the first marginal of  $P$  is  $dx_{|\Sigma|/|\Sigma|}$ .

Given  $\eta > 0$ , applying Proposition 4.1 and using the notation there, we have  $|A_n| \geq n!(\delta^2/n)^n$  and from (1.24) we have

$$(4.7) \quad 1 \geq \int_{A_n} \frac{1}{K_n^\beta} e^{-n\frac{\beta}{2}F_n(\nu_n)} dy_1 \dots dy_n,$$

where  $\nu_n = \sum_{i=1}^n \delta_{y_i}$ . From (1.23) and (4.1), when  $(y_1, \dots, y_n) \in A_n$  we have

$$F_n(\nu_n) \leq \eta + \frac{|\Sigma|}{\pi} \int W(E) dP(x, E) = \eta + \frac{1}{\pi} \int_{\Sigma} W(\sigma_{m_0(x)} E_0) dx.$$

From (1.36), and since  $\int_{\Sigma} m_0 = 1$ , we obtain

$$\frac{1}{\pi} \int_{\Sigma} W(\sigma_{m_0(x)} E_0) dx = \frac{1}{\pi} W(E_0) - \frac{1}{2} \int_{\Sigma} m_0(x) \log m_0(x) dx = \alpha,$$

by definition (1.40). We deduce

$$F_n(\nu_n) \leq \alpha + \eta.$$

Together with (4.7) we find  $1 \geq \frac{|A_n|}{K_n^\beta} e^{-n\frac{\beta}{2}(\eta+\alpha)}$ . Taking logarithms, we are led to

$$\log K_n^\beta \geq \log n! + n \log \delta^2 - n \log n - \frac{1}{2} n \beta (\eta + \alpha).$$

From Stirling's formula,  $\log n! \geq n \log n - Cn$  and we deduce (4.5), with  $C_\eta = -\log \delta^2 + C$ . Note that the dependence on  $\eta$  comes from  $\delta$ .  $\square$

## 5 Consequences for fluctuations: proof of Theorem 3

In this section, we prove Theorem 3. The first step is to find, via the method first introduced in [SS3] and tools from [SS1, STi] how  $F_n$  and  $W$  control the discrepancy between  $\nu_n$  and the measure  $n\mu_0$ , as can be seen in the following

**Lemma 5.1.** *Let  $\nu_n = \sum_{i=1}^n \delta_{x_i}$  and  $E_{\nu_n} = -\nabla H'_n$  be associated through (1.20). Let  $B_R$  be any ball of radius  $R$  (not necessarily centered at 0). Assume  $\chi$  is a smooth nonnegative function compactly supported in  $U$ . Then for any  $1 < q < 2$ , we have*

$$(5.1) \quad \|\sqrt{\chi} E_{\nu_n}\|_{L^q(U)} \leq C_q |U|^{\frac{1}{q}-\frac{1}{2}} \left( W(E_{\nu_n}, \chi) + \nu'_n(\widehat{U}) (\|\chi\|_{L^\infty} + \|\nabla \chi\|_{L^\infty}) + N \log N \right)^{\frac{1}{2}},$$

where  $N = \#\{x_1, \dots, x_n | 0 < \chi(x_i) \leq \frac{1}{2}\chi\}$ . Thus

$$(5.2) \quad \int_{B_R} |E_{\nu_n}|^q \leq C_q (n + R^2)^{1-\frac{q}{2}} n^{\frac{q}{2}} \left( \widehat{F}_n(\nu_n) + 1 \right)^{\frac{q}{2}}.$$

and

$$(5.3) \quad \|\nu_n - n\mu_0\|_{W^{-1,q}(B_R)} \leq C_q (1 + R^2)^{\frac{1}{q}-\frac{1}{2}} n^{\frac{1}{2}} \left( \widehat{F}_n(\nu_n) + 1 \right)^{\frac{1}{2}}.$$

*Proof.* The first item is a rewriting of [STi, Corollary 1.2]. We then choose  $\chi$  such that  $\chi = 1$  on  $U := \Sigma' \cup (\cup_{i=1}^n B(x'_i, \frac{1}{2})) \cup B_R$  and  $\|\chi\|_\infty, \|\nabla \chi\|_\infty \leq 1$ , compactly supported on  $\widehat{U} = \{x : d(x, U) \leq 1\}$ . Using the fact that  $|\widehat{U}| \leq C(n + R^2)$  where  $C$  depends only on  $\Sigma$ , from (5.1) we find

$$(5.4) \quad \|\sqrt{\chi} E_{\nu_n}\|_{L^q(U)} \leq C_q (n + R^2)^{\frac{1}{q}-\frac{1}{2}} (W(E_{\nu_n}, \chi) + n)^{\frac{1}{2}}.$$

Since  $\nu'_n = 0$  in the support of  $1 - \chi$ , we have

$$W(E_{\nu_n}, 1 - \chi) = \frac{1}{2} \int (1 - \chi) |E_{\nu_n}|^2 \geq 0.$$

In particular  $W(E_{\nu_n}, \chi) \leq W(E_{\nu_n}, \chi) + W(E_{\nu_n}, 1 - \chi) = W(E_{\nu_n}, \mathbf{1}_{\mathbb{R}^2})$ . It then follows from (5.4) and the fact that  $\widehat{F}_n(\nu_n) = \frac{1}{\pi n} W(E_{\nu_n}, \mathbf{1}_{\mathbb{R}^2})$  (cf. (1.26)) that (5.2) holds.

By scaling we have  $\int_\Omega |\nabla H_n|^q = n^{\frac{q}{2}-1} \int_{\Omega'} |E_{\nu_n}|^q$ , where  $H_n = -2\pi \Delta^{-1}(\nu_n - n\mu_0)$ , while  $\|\nu_n - n\mu_0\|_{W^{-1,q}(\Omega)} \leq C \|\nabla H_n\|_{L^q(\Omega)}$ . Thus (5.3) follows.  $\square$

The next proposition, whose proof will be given below, shows how  $\widehat{F}_n$  controls  $D(x'_0, R)$ .

**Proposition 5.2.** *Let  $\nu_n = \sum_{i=1}^n \delta_{x_i}$ , and  $g_{\nu_n}$  be as in Definition 3.2. There exists a universal constant  $R_0 > 0$  such that for any  $R > R_0$ , and any  $x'_0 = \sqrt{n}x_0 \in \mathbb{R}^2$ , we have*

$$(5.5) \quad \int_{B_{2R}(x'_0)} dg_{\nu_n} \geq cD(x'_0, R)^2 \min\left(1, \frac{|D(x'_0, R)|}{R^2}\right) - CR^2,$$

where  $c > 0$  and  $C$  depend only on  $V$ , and where  $D$  was defined in (1.43). <sup>5</sup>

<sup>5</sup>In fact  $R_0$  could be any positive constant, and then  $c, C$  would depend on  $R_0$  as well, but this requires to adjust  $\rho$  accordingly and we omit for simplicity to prove this fact.



We now proceed to the

*Proof of Theorem 3.* We start by proving (1.47). If  $R > R_0$  and  $|D(x'_0, R)| \geq \eta R^2$  then from Proposition 5.2 and using the fact — from Proposition 3.1 — that  $g_{\nu_n}$  is positive outside  $\cup_{i=1}^n B(x'_i, \lambda)$  and that  $g_{\nu_n} \geq -C$  everywhere, we deduce from (3.7) and (5.5) that

$$(5.6) \quad F_n(\nu_n) \geq \frac{1}{n} (-CR^2 + c \min(\eta^2, \eta^3)R^4) + 2 \int \zeta d\nu_n.$$

Inserting into (1.24) we find

$$\mathbb{P}_n^\beta (|D(x'_0, R)| \geq \eta R^2) \leq \frac{1}{K_n^\beta} \exp(C\beta R^2 - c\beta \min(\eta^2, \eta^3)R^4) \int e^{-n\beta \int \zeta d\nu_n} dx_1 \dots dx_n.$$

Then, using the lower bound (4.5) and Lemma 3.5 we deduce that if  $\beta > \beta_0 > 0$  and  $n$  is large enough depending on  $\beta_0$  then

$$\log \mathbb{P}_n^\beta (|D(x'_0, R)| \geq \eta R^2) \leq -c\beta \min(\eta^2, \eta^3)R^4 + C\beta R^2 + Cn\beta + Cn,$$

where  $c, C > 0$  depend only on  $V$ . Thus (1.47) is established.

We next prove (1.49). By Fubini's theorem, and using again the facts that  $g_{\nu_n}$  is positive outside  $\cup_{i=1}^n B(x'_i, C)$  and  $\geq -C$  everywhere we have

$$\int_{\mathbb{R}^2} dg_{\nu_n} \geq \int_{U'} \left( \int_{B(x', 2R)} dg_{\nu_n} \right) dx' - C|U'| - Cn.$$

Combining with Proposition 5.2 it follows that

$$\int_{\mathbb{R}^2} dg_{\nu_n} \geq -C(|U'| + n) + \frac{1}{R^2} \int_{U'} -CR^2 + cD(x', R)^2 \min\left(1, \frac{|D(x', R)|}{R^2}\right) dx'.$$

i.e., changing the constants if necessary,

$$(5.7) \quad \int_{\mathbb{R}^2} dg_{\nu_n} \geq -C(|U'| + n) + \frac{c}{R^2} \int_{U'} D(x', R)^2 \min\left(1, \frac{|D(x', R)|}{R^2}\right) dx.$$

It follows, using as above (1.24), (3.7), (4.5) and Lemma 3.5, and since  $|U'| = n|U|$ , that

$$\log \mathbb{P}_n^\beta \left( \int_{U'} \frac{D(x', R)^2}{R^2} \min\left(1, \frac{|D(x', R)|}{R^2}\right) dx \geq \eta \right) \leq -cn\beta\eta + Cn\beta(|U| + 1) + Cn,$$

where  $c, C > 0$  depend only on  $V$ , where  $\beta > \beta_0 > 0$  and where  $n > n_0(\beta_0)$ .

We next turn to (1.48). Arguing as above, from (3.7) we have  $F_n(\nu_n) \geq -C + 2 \int \zeta d\nu_n$ . Splitting  $2 \int \zeta d\nu_n$  as  $\int \zeta d\nu_n + \int \zeta d\nu_n$ , inserting into (1.24) and using (4.5) we are led to

$$\mathbb{P}_n^\beta \left( \int \zeta d\nu_n \geq \eta \right) \leq e^{-\frac{1}{2}n\beta\eta + Cn(\beta+1)} \int e^{-n\beta \int \zeta d\nu_n} dx_1 \dots dx_n,$$

where  $C$  depends only on  $V$ . Then, using Lemma 3.5 we deduce (1.48).

There remains to prove (1.50). Reasoning as above after (5.6), the probability that  $\widehat{F}_n(\nu_n) \geq \eta$  is bounded above for any  $\beta > \beta_0 > 0$  and  $n$  large enough depending on  $\beta_0$  by  $\exp(-\frac{1}{2}n\beta\eta + Cn(\beta + 1))$ , where  $C$  depends on  $V$  only. In view of (5.3), it follows that

$$\mathbb{P}_n^\beta \left( \left( 1 + \frac{R^2}{n} \right)^{\frac{1}{2} - \frac{1}{q}} \|\nu_n - n\mu_0\|_{W^{-1,q}(B_{R/\sqrt{n}})} \geq C_q n^{\frac{1}{2}} (1 + \eta)^{\frac{1}{2}} \right) \leq \exp(-\frac{1}{2}n\beta\eta + Cn(\beta + 1)).$$

After a slight rewriting, this concludes the proof of Theorem 3.  $\square$

We now turn to the proof of Proposition 5.2. The idea of the proof is the following: if  $\operatorname{div} E = a(x)$  in a ball centered, say at 0, then we can bound from below the contribution to the energy on circles as follows, using the Cauchy-Schwarz inequality and integration by parts

$$\int_{\partial B(0,t)} |E|^2 \geq \frac{1}{2\pi t} \left( \int_{\partial B(0,t)} E \cdot \vec{\nu} \right)^2 \geq \frac{1}{2\pi t} \left( \int_{B(0,t)} a(x) dx \right)^2.$$

Thus if we can bound from below the total charge in  $B(0,t)$ , i.e.  $\int_{B(0,t)} a$ , by integrating over circles we get a bound from below on  $\int |E|^2$  which scales like the square of that charge. This is roughly how we expect the square of the discrepancy to appear in the right-hand side of (5.5). This idea needs however two modifications in order to truly work: the first is that if there is a total charge, or charge discrepancy  $D$  in a ball of radius  $R$ , we cannot be sure that the same holds in balls of radius  $t$  different from  $R$ . However, our charge density  $a = \sum \delta_{x'_i} - \mu'_0$  has a particular structure: its negative part is bounded in  $L^\infty$ , so the charge discrepancy in  $B(0,t)$  cannot decrease too quickly as  $t$  moves away from  $R$ . The second point is that what we need to bound from below is not  $\int |E|^2$  but  $\int g$ , via (5.9), and so one may only use such lower bounds on circles that do not intersect the balls of  $\mathcal{B}_\rho$ . However it is not true in general that one can find enough such circles. In order to go around this difficulty, instead of circles, we shall use curves that are defined as level lines of the distance to  $\mathcal{B}_\rho$ , this way they will automatically avoid the balls of  $\mathcal{B}_\rho$ . The co-area formula (see e.g. [EG]) will then be used to relate  $\int |E|^2$  to the integrals along these curves.

More precisely, let us introduce a modified distance function to  $x'_0$ , as follows. Two cases can be distinguished: either  $D(x'_0, R) > 0$  or  $D(x'_0, R) \leq 0$ . If  $D(x'_0, R) > 0$ , we let, for any  $x$ ,  $f(x)$  be the infimum over the set of curves  $\gamma$  joining a point in  $B_R(x'_0)$  to  $x$  of the length of  $\gamma \setminus \mathcal{B}_\rho$ . This is also the distance to  $x'_0$  for the degenerate metric which is Euclidean outside  $B_R(x'_0) \cup \mathcal{B}_\rho$  and vanishes on  $B_R(x'_0) \cup \mathcal{B}_\rho$ .

If  $D(x'_0, R) \leq 0$  we define  $f(x)$  to be the distance of  $x$  to the complement of  $B_R(x'_0)$  with respect to the metric which is Euclidean on  $B_R(x'_0) \setminus \mathcal{B}_\rho$ .

We claim the following:

**Lemma 5.3.** *In the first case, if  $|x - x'_0| \geq R + 2$  then*

$$\frac{|x - x'_0| - R}{4} \leq f(x) \leq |x - x'_0| - R.$$

*In the second case, if  $|x - x'_0| < R - 2$  then*

$$\frac{R - |x - x'_0|}{4} \leq f(x) \leq R - |x - x'_0|.$$

*Proof.* We start with the first case. The upper bound is obvious so we turn to the lower bound. Let  $\gamma(t)$  be a continuous curve joining  $x = \gamma(0)$  to  $B_R(x'_0)$ . Let us build by induction a sequence  $x^0 = x, \dots, x^K$  with  $x^{k+1}$  defined as follows: let  $t_{k+1}$  be the smallest  $t > t_k$  such that  $\gamma(t) \notin B_1(x^k) \cap \gamma$  or  $\gamma(t) \in B_R(x'_0)$ . This procedure terminates after a finite number of steps at  $x^K \in \partial B_R(x'_0)$ . By triangle inequality we have

$$(5.8) \quad |x - x'_0| \leq \sum_{k=0}^{K-1} |x^{k+1} - x^k| + |x^K - x'_0| \leq K + R.$$

On the other hand, by property of  $\mathcal{B}_\rho$ , for any  $0 \leq k \leq K - 1$  we have

$$\ell(\gamma[t_k, t_{k+1}] \setminus \mathcal{B}_\rho) \geq |x^{k+1} - x^k| - 2\rho,$$

where  $\ell$  denotes the length. Summing this over  $k$  and using (5.8), we find

$$\ell(\gamma \setminus \mathcal{B}_\rho) \geq K - 1 - 2K\rho \geq |x - x'_0| - R - 1 - 2\rho(|x - x'_0| - R).$$

Taking the infimum over all curves  $\gamma$  we deduce that

$$f(x) \geq (|x - x'_0| - R)(1 - 2\rho) - 1.$$

Since by assumption  $|x - x'_0| - R \geq 2$ , we obtain  $f(x) \geq (|x - x'_0| - R)(1 - 2\rho - \frac{1}{2})$  and the result follows since  $\rho < 1/8$ .

The proof in the second case is analogous.  $\square$

*Proof of the proposition.* From Proposition 3.1, defining  $E_{\nu_n}$  and  $g_{\nu_n}$  as in Definition 3.2, we have

$$(5.9) \quad g_{\nu_n} \geq -C + \frac{1}{4}|E_{\nu_n}|^2 \mathbf{1}_{\mathbb{R}^2 \setminus \mathcal{B}_\rho},$$

where  $\mathcal{B}_\rho$  is a set of disjoint closed balls covering  $\text{Supp}(\nu'_n)$ , and the sum of the radii of the balls in  $\mathcal{B}_\rho$  intersecting any given ball of radius 1 is bounded by  $\rho < \frac{1}{8}$ .

We distinguish again the two cases  $D(x'_0, R) \geq 0$  or  $D(x'_0, R) < 0$ . Let us start with the first case.

Since  $f$ , as defined above, is Lipschitz with constant 1, almost every  $t$  is a regular value of  $f$ . For such a  $t$  the curve  $\gamma_t := \{f = t\}$  is Lipschitz and does not intersect  $\mathcal{B}_\rho$ , since  $\nabla f = 0$  there. Moreover, restating Lemma 5.3 we have

$$(5.10) \quad f(x) < t \implies x \in B_{R+4t}(x'_0) \cup B_{R+2}(x'_0),$$

thus  $\gamma_t \subset B_{2R}(x'_0)$  if  $R + 4t < 2R$ , i.e. if  $t < R/4$ , and  $R + 2 < 2R$ , i.e. if  $R > 2$ . It follows from (5.9) that if  $R > 2$  then

$$(5.11) \quad \int_{B_{2R}(x'_0)} dg_{\nu_n} \geq -CR^2 + \frac{1}{4} \int_{\{0 < f(x) < R/4\}} |E_{\nu_n}|^2.$$

On the other hand the co-area formula (see e.g. [EG]) asserts that

$$\int_{\{0 < f(x) < R/4\}} |E_{\nu_n}|^2 = \int_0^{R/4} \left( \int_{\gamma_t} \frac{|E_{\nu_n}|^2}{|\nabla f|} d\mathcal{H}^1 \right) dt.$$

Since  $f$  is 1-Lipschitz, it follows that

$$(5.12) \quad \int_{B_{2R}(x'_0)} dg_{\nu_n} \geq -CR^2 + \frac{1}{4} \int_{t=0}^{R/4} \left( \int_{\gamma_t} |E_{\nu_n}|^2 d\mathcal{H}^1 \right) dt.$$

We proceed by estimating the innermost integral on the right-hand side. If  $t > 1/2$  then  $B_{R+2} \subset B_{R+4t}$  and using (5.10) we find, using Definition 3.2 and writing  $D$  instead of  $D(x'_0, R)$  in the course of this proof,

$$\begin{aligned} \int_{\gamma_t} E_{\nu_n} \cdot \vec{\nu} &= \int_{\{f < t\}} \operatorname{div} E_{\nu_n} \geq 2\pi\nu_n \left( B_{\frac{R}{\sqrt{n}}}(x_0) \right) - 2\pi n\mu_0 \left( B_{\frac{R+4t}{\sqrt{n}}}(x_0) \right) \\ &\geq 2\pi D - C((R+4t)^2 - R^2), \end{aligned}$$

where we have used (1.16). The right-hand side is bounded below by  $\pi D$  if  $R+4t < \sqrt{R^2 + cD}$  and  $c$  is small enough. Thus, if

$$(5.13) \quad 1/2 < T := \frac{R}{4} \left( \sqrt{1 + c\frac{D}{R^2}} - 1 \right),$$

it follows using Cauchy-Schwarz's inequality that for every  $t \in (1/2, T)$  we have

$$(5.14) \quad \int_{\gamma_t} |E_{\nu_n}|^2 d\mathcal{H}^1 \geq \frac{\pi^2 D^2}{\mathcal{H}^1(\gamma_t)},$$

Inserting into (5.12) while reducing integration to  $1/2 < t < \min(T, R/4)$  we obtain

$$(5.15) \quad \int_{B_{2R}(x'_0)} dg_{\nu_n} \geq -CR^2 + \pi^2 D^2 \int_{t=1/2}^{\min(T, R/4)} \frac{1}{\mathcal{H}^1(\gamma_t)} dt.$$

The evaluation of the last integral is complicated by the fact that  $\gamma_t$  is the level set for  $f$  rather than the usual circle, however the result will be comparable to the one we would get if we had  $\mathcal{H}^1(\gamma_t) = 2\pi(R+t)$ , this is proven as follows: From Lemma 5.3 we have for every  $t \in [1/2, \min(T, R/4)]$  that  $\gamma_t \subset \{x : 0 < |x - x'_0| - R < \min(4T, R)\}$ . From the coarea formula and the fact that  $|\nabla f| \leq 1$  it follows that

$$\int_{\frac{1}{2}}^{\min(T, R/4)} \mathcal{H}^1(\gamma_t) dt \leq |\{x : R < |x - x'_0| < R + \min(4T, R)\}| \leq CR \min(4T, R).$$

Then, using the convexity of  $x \mapsto 1/x$  and Jensen's inequality in (5.15) we obtain for some  $c > 0$  that

$$\int_{\frac{1}{2}}^{\min(T, R/4)} \frac{1}{\mathcal{H}^1(\gamma_t)} dt \geq c \frac{(\min(T, R/4) - \frac{1}{2})^2}{R \min(4T, R)}.$$

Inserting into (5.15) we obtain assuming (5.13) and

$$(5.16) \quad 1 < \min(T, R/4)$$

that

$$(5.17) \quad \int_{B_{2R}(x'_0)} dg_{\nu_n} \geq -CR^2 + c\frac{D^2}{R} \left( \min(T, R/4) - \frac{1}{2} \right).$$

One may check that (5.13), (5.16) are satisfied if  $R > 4$  and  $D > C_0 R$  for a large enough  $C_0 > 0$ . Then it is not difficult to deduce (5.5) from (5.17) by distinguishing the cases  $T < R/4$  and  $T \geq R/4$ , i.e.  $D < C_1 R^2$  and  $D \geq C_1 R^2$  for a well chosen  $C_1$ . Finally, if  $D < C_0 R$  then (5.5) is trivially satisfied, if  $C$  is chosen large enough.

Let us turn to the case  $D(x'_0, R) \leq 0$ , which implies  $|D(x'_0, R)| \leq n\mu_0(B_{R/\sqrt{n}}(x_0)) \leq \pi R^2 \bar{m}$ . As above, if  $R > 2$  and for almost every  $1/2 < t < R/4$  the curve  $\gamma_t = \{f = t\}$  is a Lipschitz curve which does not intersect  $\mathcal{B}_\rho$  and  $\{f < t\} \subset B_R(x'_0) \setminus B_{R-4t}(x'_0)$ . It follows that, writing as before  $D$  for  $D(x_0, R)$

$$(5.18) \quad \int_{\gamma_t} E_{\nu_n} \cdot \tau = \int_{\{f < t\}} \operatorname{div} E_{\nu_n} = \int_{B_R(x'_0)} \operatorname{div} E_{\nu_n} - \int_{B(x'_0, R) \setminus \{f > t\}} \operatorname{div} E_{\nu_n} \\ \leq 2\pi D + 2\pi n\mu_0(B_{\frac{R}{\sqrt{n}}} \setminus B_{\frac{R-4t}{\sqrt{n}}}) \leq 2\pi D + C(R^2 - (R-4t)^2).$$

The proof then proceeds as in the first case by using Cauchy-Schwarz's inequality and integrating with respect to  $t \in [1/2, \min(T, R/4)]$ , where

$$T = \frac{R}{4} \left( 1 - \sqrt{1 + c \frac{D}{R^2}} \right),$$

which ensures that the right-hand side in (5.18) is bounded above by  $\pi D$ . Note that  $D$  is nonpositive, but bounded below by  $-CR^2$  hence if  $c > 0$  is small enough the quantity inside the square root above is positive.  $\square$

## 6 Lower bounds via the ergodic theorem and conclusions

### 6.1 Abstract result via the ergodic theorem

In this section, we present the ergodic framework introduced in [SS1] for obtaining “lower bounds for 2-scale energies” and inspired by Varadhan. We cannot directly use the result there because it is written for a uniform “macroscopic environment”, which would correspond to the case where  $m_0(x)$  is constant on its support (as in the circular law). To account for the possibility of varying environment or weight at the macroscopic, we can however adapt Theorem 3 of [SS1] and easily prove the following variant:

Let  $X$  denote a Polish metric space, when we speak of measurable functions on  $X$  we will always mean Borel-measurable. We assume there is a  $d$ -parameter group of transformations  $\theta_\lambda$  acting continuously on  $X$ . More precisely we require that

- For all  $u \in X$  and  $\lambda, \mu \in \mathbb{R}^d$ ,  $\theta_\lambda(\theta_\mu u) = \theta_{\lambda+\mu} u$ ,  $\theta_0 u = u$ .
- The map  $(\lambda, u) \mapsto \theta_\lambda u$  is continuous with respect to each variable (hence measurable with respect to both).

Typically we think of  $X$  as a space of functions defined on  $\mathbb{R}^d$  and  $\theta$  as the action of translations, i.e.  $\theta_\lambda u(x) = u(x + \lambda)$ . Then we consider the following  $d$ -parameter group of transformations  $T_\lambda^\varepsilon$  acting continuously on  $\mathbb{R}^d \times X$  by  $T_\lambda^\varepsilon(x, u) = (x + \varepsilon\lambda, \theta_\lambda u)$ . We also define  $T_\lambda(x, u) = (x, \theta_\lambda u)$ .

For a probability measure  $P$  on  $\mathbb{R}^d \times X$  we say that  $P$  is translation-invariant if it is invariant under the action  $T$ , and we say it is  $T_{\lambda(x)}$ -invariant if for every function  $\lambda(x)$  of class

$C^1$ , it is invariant under the mapping  $(x, u) \mapsto (x, \theta_{\lambda(x)}u)$ . Note that  $T_{\lambda(x)}$ -invariant implies translation-invariant.

Let  $G$  denote a compact set in  $\mathbb{R}^d$  such that

$$(6.1) \quad |G| > 0, \quad \lim_{\varepsilon \rightarrow 0} \frac{|(G + \varepsilon x) \Delta G|}{|G|} = 0,$$

for every  $x \in \mathbb{R}^2$ , where  $\Delta$  denotes the symmetric difference of sets. We let  $\{f_\varepsilon\}_\varepsilon$  and  $f$  be measurable nonnegative functions on  $G \times X$ , and assume that for any family  $\{(x_\varepsilon, u_\varepsilon)\}_\varepsilon$  such that

$$\forall R > 0, \quad \limsup_{\varepsilon \rightarrow 0} \int_{B_R} f_\varepsilon(T_\lambda^\varepsilon(x_\varepsilon, u_\varepsilon)) d\lambda < +\infty$$

the following holds.

1. (Coercivity)  $\{(x_\varepsilon, u_\varepsilon)\}_\varepsilon$  admits a convergent subsequence (note that  $\{x_\varepsilon\}_\varepsilon$  subsequentially converges since  $G$  is compact).
2. ( $\Gamma$ -liminf) If  $\{(x_\varepsilon, u_\varepsilon)\}_\varepsilon$  converges to  $(x, u)$  then

$$\liminf_{\varepsilon \rightarrow 0} f_\varepsilon(x_\varepsilon, u_\varepsilon) \geq f(x, u).$$

Then for the sake of generality we consider an increasing family of bounded open sets  $\{\mathbf{U}_R\}_{R>0}$  such that

$$(6.2) \quad (i) \{\mathbf{U}_R\}_{R>0} \text{ is a Vitali family, } (ii) \lim_{R \rightarrow +\infty} \frac{|(\lambda + \mathbf{U}_R) \Delta \mathbf{U}_R|}{|\mathbf{U}_R|} = 0$$

for any  $\lambda \in \mathbb{R}^d$ , where Vitali means (see [Riv]) that the intersection of the closures is  $\{0\}$ , that  $R \mapsto |\mathbf{U}_R|$  is left continuous, and that  $|\mathbf{U}_R - \mathbf{U}_R| \leq C|\mathbf{U}_R|$ .

We have

**Theorem 7.** *Let  $G, X, \{\theta_\lambda\}_\lambda$ , and  $\{f_\varepsilon\}_\varepsilon, f$  be as above. For any  $u \in X$ , let*

$$F_\varepsilon(u) = \int_G f_\varepsilon(x, \theta_{\frac{x}{\varepsilon}}u) dx.$$

and let  $\phi_\varepsilon(u)$  be the probability on  $G \times X$  which is the image of the normalized Lebesgue measure on  $G$  under the map  $x \mapsto (x, \theta_{\frac{x}{\varepsilon}}u)$ .

A. Assume that  $\{u_\varepsilon\}_\varepsilon$ , a family of elements of  $X$ , is such that  $\{F_\varepsilon(u_\varepsilon)\}_\varepsilon$  is bounded, and let  $P_\varepsilon = \phi_\varepsilon(u_\varepsilon)$ . Then  $P_\varepsilon$  converges to a Borel probability measure  $P$  on  $G \times X$  whose first marginal is the normalized Lebesgue measure on  $G$ , which is  $T_{\lambda(x)}$ -invariant, such that  $P$ -a.e.  $(x, u)$  is of the form  $\lim_{\varepsilon \rightarrow 0} (x_\varepsilon, \theta_{\frac{x_\varepsilon}{\varepsilon}}u_\varepsilon)$  and such that

$$(6.3) \quad \liminf_{\varepsilon \rightarrow 0} F_\varepsilon(u_\varepsilon) \geq \int f(x, u) dP(x, u).$$

Moreover,

$$(6.4) \quad \int f(x, u) dP(x, u) = \mathbf{E}^P \left( \lim_{R \rightarrow +\infty} \int_{\mathbf{U}_R} f(x, \theta_\lambda u) d\lambda \right),$$

where  $\mathbf{E}^P$  denotes the expectation under the probability  $P$ .

B. Let  $\mathbb{P}_\varepsilon$  be a probability on  $X$  such that  $\lim_{M \rightarrow +\infty} \lim_{\varepsilon \rightarrow 0} \mathbb{P}_\varepsilon(\{F_\varepsilon(u) \geq M\}) = 0$ , then  $\{\phi_\varepsilon \# \mathbb{P}_\varepsilon\}_\varepsilon$  is tight, i.e. converges up to a subsequence to a probability measure on  $\mathcal{P}(G \times X)$ .

The proof uses the following simple lemma, whose statement and proof can be found in [SS1, Lemma 2.1].

**Lemma 6.1.** *Assume  $\{P_n\}_n$  are Borel probability measures on a Polish metric space  $X$  and that for any  $\delta > 0$  there exists  $\{K_n\}_n$  such that  $P_n(K_n) \geq 1 - \delta$  for every  $n$  and such that if  $\{x_n\}_n$  satisfies for every  $n$  that  $x_n \in K_n$ , then any subsequence of  $\{x_n\}_n$  admits a convergent subsequence (note that we do not assume  $K_n$  to be compact). Then  $P_n$  admits a subsequence which converges tightly, i.e. converges weakly to a probability measure  $P$ .*

*Proof of the theorem.* It follows the steps of [SS1, Section 2]:

1.  $P_\varepsilon$  is tight hence has a limit  $P$ . This follows from the coercivity property of  $f_\varepsilon$  as in [SS1, Section 2, Step 1] and uses Lemma 6.1.
2.  $P$  is  $T_{\lambda(x)}$ -invariant. Let  $\Phi$  be bounded and continuous, and let  $P_\lambda$  be the push-forward of  $P$  by  $(x, u) \mapsto (x, \theta_{\lambda(x)}u)$ . Then from the definition of  $P_\lambda$ ,  $P$ ,  $P_\varepsilon$ , we have

$$\begin{aligned} \int \Phi(x, u) dP_\lambda(x, u) &= \int \Phi(x, \theta_{\lambda(x)}u) dP(x, u) = \lim_{\varepsilon \rightarrow 0} \int \Phi(x, \theta_{\lambda(x)}u) dP_\varepsilon(x, u) = \\ &= \lim_{\varepsilon \rightarrow 0} \int_G \Phi(x, \theta_{\frac{x}{\varepsilon} + \lambda(x)}u_\varepsilon) dx = \lim_{\varepsilon \rightarrow 0} \int_{(I + \varepsilon\lambda)(G)} \frac{\Phi((I + \varepsilon\lambda)^{-1}(y), \theta_{\frac{y}{\varepsilon}}u_\varepsilon)}{|\det(I + \varepsilon D\lambda((I + \varepsilon\lambda)^{-1}(y)))|} dy, \end{aligned}$$

where the last equality follows by the change of variables  $y = (I + \varepsilon\lambda)(x)$ . Using the boundedness of  $\Phi$ , the  $C^1$  character of  $\lambda$ , the compactness of  $G$  and (6.1), we may replace  $(I + \varepsilon\lambda)(G)$  by  $G$  and the denominator by 1 in the last integral and we find, using the definition of  $P_\varepsilon$

$$(6.5) \quad \int \Phi(x, u) dP_\lambda(x, u) = \lim_{\varepsilon \rightarrow 0} \int \Phi((I + \varepsilon\lambda)^{-1}(x), u) dP_\varepsilon(x, u).$$

Since  $\{P_\varepsilon\}_\varepsilon$  is tight, for any  $\delta > 0$  there exists  $K_\delta$  such that  $P_\varepsilon(K_\delta^c) < \delta$  for every  $\varepsilon$ . Then by uniform continuity of  $\Phi$  on  $K_\delta$  the map  $(x, u) \mapsto \Phi((I + \varepsilon\lambda)^{-1}(x), u)$  converges uniformly on  $K_\delta$  to  $(x, u) \mapsto \Phi(x, u)$  and thus

$$\lim_{\varepsilon \rightarrow 0} \int_{K_\delta} \Phi((I + \varepsilon\lambda)^{-1}(x), u) dP_\varepsilon(x, u) = \lim_{\varepsilon \rightarrow 0} \int_{K_\delta} \Phi(x, u) dP_\varepsilon(x, u).$$

Since this is true for any  $\delta > 0$ , and using the boundedness of  $\Phi$  we get

$$\lim_{\varepsilon \rightarrow 0} \int \Phi((I + \varepsilon\lambda)^{-1}(x), u) dP_\varepsilon(x, u) = \lim_{\varepsilon \rightarrow 0} \int \Phi(x, u) dP_\varepsilon(x, u) = \int \Phi(x, u) dP(x, u),$$

by definition of  $P$ . Thus in view of (6.5) we have  $P_\lambda = P$  and  $P$  is thus  $T_{\lambda(x)}$ -invariant.

3.  $\liminf_{\varepsilon \rightarrow 0} F_\varepsilon(u_\varepsilon) \geq \int f dP$ . This follows from [SS1, Lemma 2.2], since  $F_\varepsilon(u_\varepsilon) = \int f_\varepsilon dP_\varepsilon$ .

To conclude, as in [SS1, Section 2], the fact that  $P$  is  $T_{\lambda(x)}$ -invariant (which implies  $T_\lambda$ -invariant) and Wiener's multiparametric ergodic theorem (see e.g. [Bec]) imply that

$$\int f(x, u) dP(x, u) = \mathbf{E}^P \left( \lim_{R \rightarrow +\infty} \int_{\mathbf{U}_R} f(T_\lambda(x, u)) d\lambda \right) = \mathbf{E}^P \left( \lim_{R \rightarrow +\infty} \int_{\mathbf{U}_R} f(x, \theta_\lambda u) d\lambda \right).$$

We now turn to the proof of B. Let  $A_{M,\varepsilon} = \{u \in X, F_\varepsilon(u) \leq M\}$ . Then we have  $\phi_\varepsilon \# \mathbb{P}_\varepsilon(\phi_\varepsilon(A_{M,\varepsilon}^c)) = \mathbb{P}_\varepsilon(A_{M,\varepsilon}^c) \rightarrow 0$  as  $\varepsilon \rightarrow 0$  and  $M \rightarrow \infty$ . In view of Lemma 6.1 applied with  $K_n = \phi_\varepsilon(A_{M,\varepsilon})$ , in order to prove the tightness of  $\phi_\varepsilon \# \mathbb{P}_\varepsilon$  it suffices to take  $M$  large enough and check that if  $P_\varepsilon \in \phi_\varepsilon(A_{M,\varepsilon})$  then  $P_\varepsilon$  has a convergent subsequence. But this is a direct application of what we have established in part A, since such a  $P_\varepsilon$  is the image by  $\phi_\varepsilon$  of a family  $u_\varepsilon$  for which  $F_\varepsilon(u_\varepsilon) \leq M$ . Therefore  $P_\varepsilon$  is tight and  $\phi_\varepsilon \# \mathbb{P}_\varepsilon$  as well by the lemma.  $\square$

We now apply this abstract framework to our specific situation to obtain the lower bound on  $\widehat{F}_n$ .

## 6.2 Proof of Theorem 2, part A

The proof follows essentially [SS1], Proposition 4.1 and below. Let  $\{\nu_n\}_n$  and  $P_{\nu_n}$  be as in the statement of Theorem 2. We need to prove that any subsequence of  $\{P_{\nu_n}\}_n$  has a convergent subsequence and that the limit  $P$  is a  $T_{\lambda(x)}$ -invariant probability measure such that  $P$ -almost every  $(x, E)$  is such that  $E \in \mathcal{A}_{m_0(x)}$  and (1.44) holds. Note that the fact that the first marginal of  $P$  is  $dx_{|\Sigma|/|\Sigma|}$  follows from the fact that, by definition, this is true of  $P_{\nu_n}$ .

We thus take a subsequence of  $\{P_{\nu_n}\}$  (which we don't relabel). We may assume that it has a subsequence, denoted  $\bar{\nu}_n$ , which satisfies  $\widehat{F}_n(\bar{\nu}_n) \leq C$ , otherwise there is nothing to prove. This implies that  $\bar{\nu}_n$  is of the form  $\sum_{i=1}^n \delta_{x_i}$ . We let  $\bar{E}_n$  denote the current and  $\bar{g}_n$  the measures associated to  $\bar{\nu}_n$  as in Definition 3.2 and note that  $\int d\bar{g}_n = W(\bar{E}_n, \mathbf{1}_{\mathbb{R}^2})$ . As usual,  $\bar{\nu}'_n = \sum_{i=1}^n \delta_{\sqrt{n}x_i}$ .

A first consequence of  $\widehat{F}_n(\bar{\nu}_n) \leq C$  is that, in view of (5.3), we have

$$(6.6) \quad \frac{1}{n} \bar{\nu}_n \rightarrow \mu_0,$$

in the weak sense of measures.

### Step 1: We set up the framework of Section 6.1

We will use integers  $n$  instead of  $\varepsilon$  to label sequences, and the correspondence will be  $\varepsilon = 1/\sqrt{n}$ . We let  $G = \Sigma$  and  $X = \mathcal{M}_+ \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2) \times \mathcal{M}$ , where  $p \in (1, 2)$ , where  $\mathcal{M}_+$  denotes the set of positive Radon measures on  $\mathbb{R}^2$  and  $\mathcal{M}$  the set of those which are bounded below by the constant  $-C(\|m_0\|_\infty + 1)$  of Proposition 3.1, both equipped with the topology of weak convergence.

For  $\lambda \in \mathbb{R}^2$  and abusing notation we let  $\theta_\lambda$  denote both the translation  $x \rightarrow x + \lambda$  and the action

$$\theta_\lambda(\nu, E, g) = (\theta_\lambda \# \nu, E \circ \theta_\lambda, \theta_\lambda \# g).$$

Accordingly the action  $T^n$  is defined for  $\lambda \in \mathbb{R}^2$  by

$$T_\lambda^n(x, \nu, E, g) = \left( x + \frac{\lambda}{\sqrt{n}}, \theta_\lambda \# \nu, E \circ \theta_\lambda, \theta_\lambda \# g \right).$$

Then we let  $\chi$  be a smooth cut-off function with integral 1 and support in  $B(0, 1)$  and define

$$(6.7) \quad \mathbf{f}_n(x, \nu, E, g) = \begin{cases} \frac{1}{\pi} \int_{\mathbb{R}^2} \chi(y) dg(y) & \text{if } (\nu, E, g) = \theta_{\sqrt{n}x}(\bar{\nu}'_n, \bar{E}_n, \bar{g}_n), \\ +\infty & \text{otherwise.} \end{cases}$$



Finally we let, in agreement with Section 6.1,

$$(6.8) \quad \mathbf{F}_n(\nu, E, g) = \int_{\Sigma} \mathbf{f}_n \left( x, \theta_{x\sqrt{n}}(\nu, E, g) \right) dx.$$

We have the following relation between  $\mathbf{F}_n$  and  $\widehat{F}_n$ , as  $n \rightarrow +\infty$ :

$$(6.9) \quad \mathbf{F}_n(\nu, E, g) \text{ is } \begin{cases} \leq \frac{1}{|\Sigma|} \widehat{F}_n(\bar{\nu}_n) + o(1) & \text{if } (\nu, E, g) = (\bar{\nu}'_n, \bar{E}_n, \bar{g}_n) \\ = +\infty & \text{otherwise.} \end{cases}$$

Indeed it is obvious from (6.7) that if  $(\nu, E, g) \neq (\bar{\nu}'_n, \bar{E}_n, \bar{g}_n)$  then  $\mathbf{F}_n(\nu, E, g) = +\infty$ . On the other hand, if  $(\nu, E, g) = (\bar{\nu}'_n, \bar{E}_n, \bar{g}_n)$ , then from the definition of the image measure  $\theta_{\lambda\#}\bar{g}_n$ ,

$$\mathbf{F}_n(\nu, E, g) = \frac{1}{\pi} \int_{\Sigma} \int \chi(y - x\sqrt{n}) d\bar{g}_n(y) dx = \frac{1}{\pi|\Sigma'|} \int \chi * \mathbf{1}_{\Sigma'} d\bar{g}_n.$$

Since  $\chi * \mathbf{1}_{\Sigma'}$  is bounded above by 1 and is equal to 1 on  $U := \{x' : \text{dist}(x', \mathbb{R}^2 \setminus \Sigma') \geq 1\}$  we deduce that

$$(6.10) \quad \pi \mathbf{F}_n(\nu, E, g) \leq \frac{\bar{g}_n^+(\mathbb{R}^2) - \bar{g}_n^-(U)}{|\Sigma'|} = \frac{\bar{g}_n(\mathbb{R}^2) + \bar{g}_n^-(U^c)}{n|\Sigma|} \\ = \frac{\pi \widehat{F}_n(\bar{\nu}_n)}{|\Sigma|} + \frac{\bar{g}_n^-(U^c)}{n|\Sigma|}.$$

Then we note that from (3.1)–(3.2) in Proposition 3.1 the measure  $\bar{g}_n^-$  is supported in the union of balls  $B(x', C)$  for  $x' \in \text{Supp}(\bar{\nu}'_n)$ , and bounded above by a constant. Thus  $\bar{g}_n^-(U^c)$  is bounded by a constant times the number of balls intersecting  $U^c$ , hence by  $C\bar{\nu}'_n\{x' : \text{dist}(x', U^c) \leq C\}$ . From (6.6) this is equal to

$$Cn\mu_0\{x : \text{dist}(x, \partial\Sigma) \leq C/\sqrt{n}\} + o(n) \leq Cn|\{x : \text{dist}(x, \partial\Sigma) \leq C/\sqrt{n}\}| + o(n)$$

since  $m_0$  is bounded. Using standard estimates on the volumes of tubular neighborhoods, since  $\partial\Sigma$  is  $C^1$  by assumption (1.13), we conclude that this is  $o(n)$ . Plugging this into (6.10) proves (6.9).

## Step 2: We check the hypotheses in Section 6.1

We must now check the  $\Gamma$ -liminf and coercivity properties of  $\{\mathbf{f}_n\}_n$ . The main point is again that  $\widehat{F}_n$  controls  $\nu_n - n\mu_0$  by Lemma 5.1.

**Lemma 6.2.** *Assume that  $\{(x_n, \nu_n, E_n, g_n)\}_n$  converges to  $(x, \nu, E, g)$ . Then*

$$\liminf_n \mathbf{f}_n(x_n, \nu_n, E_n, g_n) \geq \mathbf{f}(x, \nu, E, g) := \frac{1}{\pi} \int \chi dg.$$

*Proof.* We may assume that the left-hand side is finite, in which case  $\mathbf{f}_n(x_n, \nu_n, E_n, g_n) = \frac{1}{\pi} \int \chi dg_n$  for every large enough  $n$ , from which the result follows by passing to the limit.  $\square$

**Lemma 6.3.** *Assume that for any  $R > 0$  we have*

$$(6.11) \quad \limsup_{n \rightarrow +\infty} \int_{B_R} \mathbf{f}_n \left( x_n + \frac{\lambda}{\sqrt{n}}, \theta_{\lambda}(\nu_n, E_n, g_n) \right) d\lambda < +\infty.$$

*Then a subsequence of  $\{(x_n, \nu_n, E_n, g_n)\}_n$  converges to some  $(x, \nu, E, g) \in \Sigma \times X$ .*

*Proof.* Assume (6.11). Then the integrand there is bounded for a.e.  $\lambda$  and from the definition (6.7) we deduce that

$$\theta_\lambda(\nu_n, E_n, g_n) = \theta_{\sqrt{n}x_n + \lambda}(\bar{\nu}'_n, \bar{E}_n, \bar{g}_n)$$

and then that  $(\nu_n, E_n, g_n) = \theta_{\sqrt{n}x_n}(\bar{\nu}'_n, \bar{E}_n, \bar{g}_n)$ . Thus (6.11) gives, in view of (6.7), that for every  $R > 0$  there exists  $C_R > 0$  such that for any  $n$

$$\int_{B_R} \int \chi(y - \sqrt{n}x_n - \lambda) d\bar{g}_n(y) d\lambda = \int \chi * \mathbf{1}_{B_R(\sqrt{n}x_n)} d\bar{g}_n < C_R.$$

This and the fact that  $\bar{g}_n$  is bounded below implies that  $\bar{g}_n(B_R(\sqrt{n}x_n))$  is bounded independently of  $n$  and then, using (3.4), that the same is true of  $\bar{\nu}'_n(B_R(\sqrt{n}x_n))$ . In other words  $\{\nu_n = \theta_{\sqrt{n}x_n} \bar{\nu}'_n\}_n$  is a locally bounded sequence of (positive) measures hence converges weakly after taking a subsequence, and the same is true of  $\{g_n = \theta_{\sqrt{n}x_n} \bar{g}_n\}_n$ . On the other hand  $\{x_n\}_n$  is a sequence in the compact set  $\Sigma$  hence converges modulo a subsequence.

It remains to study the convergence of  $\{E_n = \bar{E}_n \circ \theta_{\sqrt{n}x_n + \lambda}\}_n$ . From (3.3) in Proposition 3.1 and the local boundedness of  $\{\nu_n\}_n$  we get that  $W(\bar{E}_n, \chi * \mathbf{1}_{B_R(\sqrt{n}x_n)}) = W(E_n, \chi * \mathbf{1}_{B_R})$  is bounded independently of  $n$  for any  $R > 0$  and then, using (5.1), that  $\{E_n\}_n$  is locally bounded in  $L^p_{loc}(\mathbb{R}^2, \mathbb{R}^2)$ , for any  $1 \leq p < 2$  hence a subsequence locally weakly converges in  $L^p_{loc}(\mathbb{R}^2, \mathbb{R}^2)$ . Moreover,  $\text{curl } E_n = 0$  and by the above  $\text{div } E_n$  is locally bounded in the sense of measures, hence weakly compact in  $W^{-1,p}_{loc}$  for  $p < 2$ . By elliptic regularity, it follows that the convergence of  $E_n$  is strong in  $L^p_{loc}(\mathbb{R}^2, \mathbb{R}^2)$ . This concludes the proof of coercivity.  $\square$

### Step 3: Conclusion

From the previous steps, we may apply Theorem 7 in this setting (choosing  $\mathbf{U}_R = K_R$ ) and we deduce in view of (6.9) that, temporarily denoting  $Q_n$  denote the push-forward of the normalized Lebesgue measure on  $\Sigma$  by the map  $x \mapsto (x, \theta_{\sqrt{n}x}(\bar{\nu}'_n, \bar{E}_n, \bar{g}_n))$ , and  $Q = \lim_n Q_n$ ,

$$(6.12) \quad \liminf_n \frac{1}{|\widehat{\Sigma}|} \widehat{F}_n(\bar{\nu}_n) \geq \liminf_n \mathbf{F}_n(\bar{\nu}'_n, \bar{E}_n, \bar{g}_n) \geq \int \left( \frac{1}{\pi} \int \chi dg \right) dQ(x, \nu, E, g) = \int \lim_{R \rightarrow +\infty} \int_{K_R} \frac{1}{\pi} \chi(y - \lambda) dg(y) d\lambda dQ(x, \nu, E, g) = \int \lim_{R \rightarrow +\infty} \left( \frac{1}{\pi |K_R|} \int \chi * \mathbf{1}_{K_R} dg \right) dQ(x, \nu, E, g).$$

Now we use the fact that for  $Q$ -almost every  $(x, \nu, E, g)$ :

- i) There exists a sequence  $\{x_n\}_n$  in  $\Sigma$  such that  $(x, \nu, E, g) = \lim_n (x_n, \theta_{\sqrt{n}x_n}(\bar{\nu}'_n, \bar{E}_n, \bar{g}_n))$ .
- ii) As a consequence of the above  $\frac{1}{\pi |K_R|} \int \chi * \mathbf{1}_{K_R} dg$  converges to a finite limit as  $R \rightarrow +\infty$ .

The first point implies, since  $\text{div } \bar{E}_n = \bar{\nu}'_n - m_0'$  and  $\text{curl } \bar{E}_n = 0$ , that by passing to the limit  $n \rightarrow \infty$  we have  $\text{div } E = \nu - m_0(x)$  and  $\text{curl } E = 0$ . The second point implies in particular using (3.4) that  $\nu(B_R) \leq CR^2$ , proving that  $(E, \nu) \in \mathcal{A}_{m_0(x)}$ .

Moreover the second point implies that for any  $C > 0$  we have  $g(K_{R+C} \setminus K_{R-C}) = o(R^2)$  as  $R \rightarrow +\infty$ , and thus from point i) above

$$\lim_{R \rightarrow +\infty} \lim_{n \rightarrow +\infty} \frac{1}{R^2} \bar{g}_n(K_{R+C}(\sqrt{n}x_n) \setminus K_{R-C}(\sqrt{n}x_n)) = 0.$$

Using (3.4) we deduce that

$$\lim_{R \rightarrow +\infty} \lim_{n \rightarrow +\infty} \frac{1}{R^2} \bar{\nu}'_n(K_{R+C}(\sqrt{n}x_n) \setminus K_{R-C}(\sqrt{n}x_n)) = 0$$

and then from (3.3),

$$\lim_{R \rightarrow +\infty} \lim_{n \rightarrow +\infty} \frac{1}{R^2} \left| W(\bar{E}_n, \chi * \mathbf{1}_{K_R(\sqrt{n}x_n)}) - \int \chi * \mathbf{1}_{K_R(\sqrt{n}x_n)} d\bar{g}_n \right| = 0.$$

Thus, using [SS1, Lemma 4.8] we may take the limit  $n \rightarrow \infty$  and deduce

$$\lim_{R \rightarrow +\infty} \frac{1}{R^2} \left| W(E, \chi * \mathbf{1}_{K_R}) - \int \chi * \mathbf{1}_{K_R} dg \right| = 0.$$

Together with (6.12) this yields, by definition of  $W$ ,

$$(6.13) \quad \liminf_n \frac{1}{|\Sigma|} \widehat{F}_n(\bar{\nu}_n) \geq \frac{1}{\pi} \int W(E) dQ(x, \nu, E, g)$$

and, we recall,  $Q$ -a.e.  $(E, \nu) \in \mathcal{A}_{m_0(x)}$ .

Now we let  $P_n$  (resp.  $P$ ) be the marginal of  $Q_n$  (resp.  $Q$ ) with respect to the variables  $(x, E)$ . Then the first marginal of  $P$  is the normalized Lebesgue measure on  $\Sigma$  and  $P$ -a.e. we have  $E \in \mathcal{A}_{m_0(x)}$ , in particular

$$W(E) \geq \min_{\mathcal{A}_{m_0(x)}} W = m_0(x) \left( \min_{\mathcal{A}_1} W - \frac{\pi}{2} \log m_0(x) \right).$$

Integrating with respect to  $P$  and noting that since only  $x$  appears on the right-hand side we may replace  $P$  by its first marginal there we find, in view of (1.40) that the lower bound (1.44) holds.

### 6.3 Proof of Theorem 2, completed

As mentioned above, Part B of the theorem is a direct consequence of Proposition 4.1, see Corollary 4.2.

Part C follows from the comparison of Parts A and B: for minimizers, the chains of inequalities (1.44) and (1.45) are in fact equalities and  $\frac{1}{\pi} \int W dP$  must be minimized hence equal to  $\alpha$ . Also we must have  $\lim_{n \rightarrow \infty} (F_n(\nu_n) - \widehat{F}_n)(\nu_n) = \lim_{n \rightarrow \infty} \int \zeta d\nu_n = 0$ , which in view of (1.18), implies that  $\lim \sum_i \text{dist}(x_i, \Sigma)^2 = 0$ .

From the fact that  $\widehat{F}_n(\nu_n) = O(1)$ , we deduce from Proposition 5.2, 3.1 and (3.7), (arguing exactly as in the proof of Theorem 3) that there exists  $C > 0$  such that for every  $x, R > 1$ , we have

$$D(x, R)^2 \min \left( 1, \frac{|D(x, R)|}{R^2} \right) \leq Cn.$$

This completes the proof of Theorem 2.

## 6.4 Deviations: proof of Theorems 5 and Theorem 1

We start with the upper bound on  $\log \mathbb{P}_n^\beta$ . Let  $A_n$  be a subset of  $\mathbb{C}^n$ . We identify points in  $\mathbb{C}^n$  with measures  $\nu_n$  of the form  $\sum_{i=1}^n \delta_{x_i}$ .

From (1.24), we have

$$\mathbb{P}_n^\beta(A_n) = \frac{1}{K_n^\beta} \int_{A_n} e^{-\frac{1}{2}\beta n F_n(\sum_{i=1}^n \delta_{x_i})} dx_1 \dots dx_n$$

hence

$$(6.14) \quad \frac{\log \mathbb{P}_n^\beta(A_n)}{n} = -\frac{\log K_n^\beta}{n} + \frac{1}{n} \log \int_{A_n} e^{-\frac{1}{2}\beta n F_n(\sum_{i=1}^n \delta_{x_i})} dx_1 \dots dx_n.$$

We deduce, since  $\widehat{F}_n(\nu_n) = F_n(\nu_n) - 2 \int \zeta d\nu_n$ , that

$$(6.15) \quad \frac{\log \mathbb{P}_n^\beta(A_n)}{n} \leq -\frac{\log K_n^\beta}{n} + \frac{1}{n} \log \left( e^{-\frac{1}{2}\beta n \inf_{A_n} \widehat{F}_n} \int_{A_n} e^{-\beta n \int \zeta d\nu_n} dx_1 \dots dx_n \right).$$

Let  $\nu_n$  such that  $\widehat{F}_n(\nu_n) \leq \inf_{A_n} \widehat{F}_n + 1/n$ . Then from (1.44) in Theorem 2 we have, using the notations there,  $\liminf_{n \rightarrow \infty} \widehat{F}_n(\nu_n) \geq \frac{|\Sigma|}{\pi} \int W(E) dP(x, E)$  where  $P = \lim_n P_{\nu_n}$ . Since  $P_{\nu_n} \in i_n(A_n)$  by definition we have  $P \in A_\infty$  since by definition  $A_\infty = \bigcap_{n>0} \overline{\bigcup_{m>n} i_m(A_m)}$ . We may thus write

$$(6.16) \quad \liminf_{n \rightarrow +\infty} \widehat{F}_n(\nu_n) \geq \frac{|\Sigma|}{\pi} \inf_{P \in A_\infty} \int W(E) dP(x, E).$$

Inserting into (6.15) we are led to

$$(6.17) \quad \frac{\log \mathbb{P}_n^\beta(A_n)}{n} \leq -\frac{\beta|\Sigma|}{2\pi} \inf_{P \in A_\infty} \int W(E) dP(x, E) - \frac{\log K_n^\beta}{n} + \frac{1}{n} \log \left( \int_{\mathbb{C}^n} e^{-\beta n \int \zeta d\nu_n} dx_1 \dots dx_n \right) + o(1)$$

thus in view of Lemma 3.5 and (4.5), we have established (1.52). An immediate corollary of (6.17), choosing  $A_n$  to be the full space and using  $\inf \frac{|\Sigma|}{\pi} \int W(E) dP(E) = \alpha$  and Lemma 3.5, is that

$$(6.18) \quad \limsup_{n \rightarrow \infty} \frac{\log K_n^\beta}{n} \leq -\frac{\beta\alpha}{2} + \log |\Sigma|.$$

We next turn to the lower bound. Fix  $\eta > 0$ . Given  $A$ , let  $P \in \mathring{A}$  be such that

$$(6.19) \quad \int W(E) dP(x, E) \leq \inf_{P \in \mathring{A}} \int W(E) dP(E) + \frac{\eta}{2}.$$

Since  $P \in \mathring{A}$ , if  $\eta$  is chosen small enough (which we assume) then  $B(P, 2\eta) \subset A$ , where the ball is for a distance metrizing weak convergence as in Proposition 4.1.

We then apply Proposition 4.1 to  $P$  and  $\eta$ . We find  $\delta > 0$  and for any  $n$  large enough a set  $A_n$  such that  $|A_n| \geq n!(\pi\delta^2/n)^n$  and, rewriting (4.1) with (2.1),

$$(6.20) \quad \limsup_{n \rightarrow \infty} \sup_{A_n} F_n \leq \frac{|\Sigma|}{\pi} \int W(E) dP(E) + \eta.$$

Moreover, for every  $(y_1, \dots, y_n) \in A_n$  and letting  $\{\nu_n = \delta_{y_1} + \dots + \delta_{y_n}\}_n$ , there exists  $\{E_n\}_n$  in  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  such that  $\text{div } E_n = 2\pi(\nu'_n - m_0')$  and such that the image  $P_n$  of  $dx_{|\Sigma|}/|\Sigma|$  by the map  $x \mapsto (x, E_n(\sqrt{n}x + \cdot))$  is such that

$$(6.21) \quad \limsup_{n \rightarrow \infty} \text{dist}(P_n, P) \leq \eta.$$

In particular (1.54) holds. Moreover, inserting (6.20) and (6.19) into (1.24), we find that

$$\frac{\log \mathbb{P}_n^\beta(A_n)}{n} \geq -\frac{\log K_n^\beta}{n} - \frac{\beta|\Sigma|}{2\pi} \inf_{P \in \mathcal{A}} \int W(E) dP(E) - \frac{1}{2}\beta\eta + \frac{1}{n} \log \left| \frac{A_n}{\sqrt{n}} \right| + o(1).$$

On the other hand, using  $|A_n| \geq n!(\pi\delta^2/n)^n$  and Stirling's formula, we have  $\log |A_n| \geq 2n \log \delta - Cn$ . Combining with (6.18), (1.53) follows, with  $C_\eta = -2 \log \delta + C + \log |\Sigma|$ .

Theorem 1 immediately follows by combining (6.18), (4.5) and (1.25).

## 6.5 Proof of Theorem 6

We apply the method of Theorem 7 part B. Let  $A_{n,M} = \{(x_1, \dots, x_n) : \widehat{F}_n(\sum_{i=1}^n \delta_{x_i}) \leq M\}$ . In view of (6.15), Corollary 4.3, and Lemma 3.5, if  $M$  is chosen large enough we have  $\mathbb{P}_n^\beta(A_{n,M}^c) \rightarrow 0$  as  $n \rightarrow \infty$ . In view of Lemma 6.1, to prove the tightness of  $i_n \# \mathbb{P}_n^\beta$  it thus suffices to check that if  $P_n \in i_n(A_{n,M})$  then  $P_n$  has a convergent subsequence. But we have just proven this in Theorem 2, part A, item 1.

## 6.6 Definition of $\mathbb{W}$

In this subsection we briefly examine how to define the renormalized energy as a function of the points only, via (1.38). We prove the following:

**Lemma 6.4.** *The function  $\mathbb{W}$  be defined by (1.38) is Borel-measurable on the set of locally finite measures.*

*Proof.* First we show that there exists a measurable map  $\nu \mapsto E_\nu$  where  $E_\nu$  satisfies (1.27). The set

$$A = \{E \in \mathcal{A}_m, W(E) < \infty\}$$

is Borel measurable, since  $W$  is (as proven in [SS1, Theorem 1]). We may partition  $A$  into equivalence classes for the relation  $E \sim E'$  if  $\text{div } E = \text{div } E'$ . In view of Lemma 1.4, denoting by  $E^*$  the equivalence class of  $E \in A$ , we have  $E^* = \{E + \vec{C}, \vec{C} \in \mathbb{R}^2\}$ . In particular this implies that if  $U$  is an open set in  $A$ , then  $U^* = \cup_{E \in U} E^*$  is open too in  $A/\sim$ . By Effros's theorem (cf. e.g. [Sri, Theorem 5.4.3]) there thus exists a Borel section  $B$  of  $A$  which contains exactly one element of each equivalence class. The map  $E^* \mapsto \frac{1}{2\pi} \text{div } E + m$  is then a Borel measurable and injective map from  $B$  to  $\{\nu \in \mathcal{M}_+ : \mathbb{W}(\nu) < \infty\}$  where  $\mathcal{M}_+$  is the set of positive Radon measures on  $\mathbb{R}^2$ . By [Co, Prop. 8.3.5] its inverse is also Borel

measurable and injective. This provides a measurable selection  $\psi : \nu \mapsto E$  satisfying (1.27) on  $\{\nu \in \mathcal{M}_+ : \mathbb{W}(\nu) < \infty\}$ . Since  $E^* = \{E + \vec{C}, \vec{C} \in \mathbb{R}^2\}$ , we may write

$$\mathbb{W}(\nu) = \inf_{\vec{C} \in \mathbb{R}^2} W(\psi(\nu) + \vec{C}).$$

Using again the fact that  $W$  is Borel measurable and  $\nu \mapsto \psi(\nu) + \vec{C}$  too, it follows that  $\mathbb{W}$  is measurable as claimed.  $\square$

We may then without too much difficulty translate the results of Theorems 2, 5 with  $\int \mathbb{W}(\nu) dQ(x, \nu)$  instead of  $\int W(E) dP(x, E)$ .

## 7 Proof of Proposition 4.1

The construction consists of the following. We are given  $\varepsilon > 0$ , which is the error we can afford. First we select a finite set of vector fields  $J_1, \dots, J_N$  ( $N$  will depend on  $\varepsilon$ ) which will represent the probability  $P(x, E)$  with respect to its  $E$  dependence, and whose renormalized energies are well-controlled. Since  $P$  is  $T_{\lambda(x)}$ -invariant, we need it to be well-approximated by measures supported on the orbits of the  $J_i$ 's under translations. Secondly, we work in blown-up coordinates and split the region  $\Sigma'$  (whose diameter is order  $\sqrt{n}$ ) into many rectangles  $K$  with centers  $x_K$  and sidelengths of order  $\bar{R}$  large enough. Even though we choose  $\bar{R}$  to be large, it will still be very small compared to the size of  $\Sigma'$ , as  $n \rightarrow \infty$ , so that the Diracs at  $x_K/\sqrt{n}$  approximate  $P(x, E)$  with respect to its  $x$  dependence. On each rectangle  $K$ , the weight  $m_0'$  is temporarily replaced by its average  $m_K$ . Then we split each rectangle  $K$  into  $q^2$  identical rectangles, with sidelengths of order  $2R = \bar{R}/q$ , where both  $R$  and  $q$  will be sufficiently large. We then select the proportion of the rectangles that corresponds to the weight that the orbit of each  $J_i$  carries in the approximation of  $P$ . In these rectangles we paste a (translated) copy of  $J_i$  at the scale  $m_K$  and suitably modified near the boundary according to a construction of [SS1] (Proposition 7.4 below) so that its tangential component on the boundary is 0 (this can be done while inducing only an error  $\varepsilon$  on  $W$ ). In the few rectangles that may remain unfilled, we paste a copy of an arbitrary  $J_0$  whose renormalized energy is finite. We perform the construction above provided we are far enough from  $\partial\Sigma'$ . The layer near the boundary must be treated separately, and there again an arbitrary (translated and rescaled) current can be pasted. Finally, we add a vector field to correct the discrepancy between  $m_K$  and  $m_0'$  in each of the rectangles.

To conclude the proof of Proposition 4.1, we collect all of the estimates on the constructed vector field to show that its energy  $w_n$  is bounded above in terms of  $\int W dP$  and that the probability measures associated to the construction have remained close to  $P$ .

### 7.1 Estimates on distances between probabilities

First we choose distances which metrize the topologies of  $L^p_{\text{loc}}(\mathbb{R}^2, \mathbb{R}^2)$  and  $\mathcal{B}(X)$ , the set of finite Borel measures on  $X = \Sigma \times L^p_{\text{loc}}(\mathbb{R}^2, \mathbb{R}^2)$ . For  $E_1, E_2 \in L^p_{\text{loc}}(\mathbb{R}^2, \mathbb{R}^2)$  we let

$$d_p(E_1, E_2) = \sum_{k=1}^{\infty} 2^{-k} \frac{\|E_1 - E_2\|_{L^p(B(0,k))}}{1 + \|E_1 - E_2\|_{L^p(B(0,k))}},$$

and on  $X$  we use the sum of the Euclidean distance on  $\Sigma$  and  $d_p$ , which we denote  $d_X$ . It is a distance on  $X$ . On  $\mathcal{B}(X)$  we define a distance by choosing a sequence of bounded continuous functions  $\{\varphi_k\}_k$  which is dense in  $C_b(X)$  and we let, for any  $\mu_1, \mu_2 \in \mathcal{B}(X)$ ,

$$d_{\mathcal{B}}(\mu_1, \mu_2) = \sum_{k=1}^{\infty} 2^{-k} \frac{|\langle \varphi_k, \mu_1 - \mu_2 \rangle|}{1 + |\langle \varphi_k, \mu_1 - \mu_2 \rangle|},$$

where we have used the notation  $\langle \varphi, \mu \rangle = \int \varphi d\mu$ .

We have the following general facts.

**Lemma 7.1.** *For any  $\varepsilon > 0$  there exists  $\eta_0 > 0$  such that if  $P, Q \in \mathcal{B}(X)$  and  $\|P - Q\| < \eta_0$ , then  $d_{\mathcal{B}}(P, Q) < \varepsilon$ . Here  $\|P - Q\|$  denotes the total variation of the signed measure  $P - Q$ , i.e. the supremum of  $\langle \varphi, P - Q \rangle$  over measurable functions  $\varphi$  such that  $|\varphi| \leq 1$ .*

In particular, if  $P = \sum_{i=1}^{\infty} \alpha_i \delta_{x_i}$  and  $Q = \sum_{i=1}^{\infty} \beta_i \delta_{x_i}$  with  $\sum_i |\alpha_i - \beta_i| < \eta_0$ , then  $d_{\mathcal{B}}(P, Q) < \varepsilon$ .

**Lemma 7.2.** *Let  $K \subset X$  be compact. For any  $\varepsilon > 0$  there exists  $\eta_1 > 0$  such that if  $x \in K, y \in X$  and  $d_X(x, y) < \eta_1$  then  $d_{\mathcal{B}}(\delta_x, \delta_y) < \varepsilon$ .*

**Lemma 7.3.** *Let  $0 < \varepsilon < 1$ . If  $\mu$  is a probability measure on a set  $A$  and  $f, g : A \rightarrow X$  are measurable and such that  $d_{\mathcal{B}}(\delta_{f(x)}, \delta_{g(x)}) < \varepsilon$  for every  $x \in A$ , then*

$$d_{\mathcal{B}}(f^{\#}\mu, g^{\#}\mu) < C\varepsilon(|\log \varepsilon| + 1).$$

*Proof.* Take any bounded continuous function  $\varphi_k$  defining the distance on  $\mathcal{B}(X)$ . Then if  $d_{\mathcal{B}}(\delta_{f(x)}, \delta_{g(x)}) < \varepsilon$  for any  $x \in X$  we have in particular

$$\frac{|\varphi_k(f(x)) - \varphi_k(g(x))|}{1 + |\varphi_k(f(x)) - \varphi_k(g(x))|} \leq 2^k \varepsilon.$$

It follows that

$$d_{\mathcal{B}}(f^{\#}\mu, g^{\#}\mu) \leq \sum_k 2^{-k} \min(\varepsilon 2^k, 1) \leq \varepsilon (|\log_2 \varepsilon| + 1) + \sum_{k=|\log_2 \varepsilon|+1}^{\infty} 2^{-k} \leq C\varepsilon(|\log \varepsilon| + 1).$$

□

## 7.2 Preliminary results

In what follows  $\Sigma' = \sqrt{n}\Sigma$ ,  $m_0'(x) = m_0(x/\sqrt{n})$ : we work in blown-up coordinates. We consider a probability measure  $P$  on  $\Sigma \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  which is as in the proposition. We let  $\tilde{P}$  be the probability measure on  $\Sigma \times \mathcal{A}_1$  which is the image of  $P$  under  $(x, E) \mapsto (x, \sigma_{1/m_0(x)}E)$ , so that

$$(7.1) \quad \tilde{P} = \int \delta_x \otimes \delta_{\sigma_{1/m_0(x)}E} dP(x, E), \quad P = \int \delta_x \otimes \delta_{\sigma_{m_0(x)}E} d\tilde{P}(x, E)$$

It is easy to check that since  $P$  is  $T_{\lambda(x)}$ -invariant,  $\tilde{P}$  is as well, and in particular it is translation-invariant.

The construction is based on the following statement which is a rewriting of Proposition 4.2 in [SS1] and the remark following it:

**Proposition 7.4** (Screening of an arbitrary vector field). *Let  $K_R = [-R, R]^2$ , let  $\{\chi_R\}_R$  satisfy (1.31).*

*Let  $G \subset \mathcal{A}_1$  be such that there exists  $C > 0$  such that for any  $E \in G$  we have*

$$(7.2) \quad \forall R > 1, \quad \frac{\nu(K_R)}{|K_R|} < C,$$

*for the associated  $\nu$ 's, and*

$$(7.3) \quad \lim_{R \rightarrow +\infty} \frac{W(E, \chi_R)}{|K_R|} = W(E),$$

*where the convergence is uniform w.r.t.  $E \in G$ .*

*Then for any  $\varepsilon > 0$  there exists  $R_0 > 0$ ,  $\eta_2 > 0$  such that if  $R > R_0$  and  $L$  is a rectangle centered at 0 whose sidelengths belong to  $[2R, 2R(1 + \eta_2)]$  and such that  $|L| \in \mathbb{N}$ , then for every  $E \in G$  there exists a  $E_L \in L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  such that the following hold*

*i)  $E_L = 0$  in  $L^c$ ,*

*ii) There is a discrete subset  $\Lambda \subset L$  such that*

$$\operatorname{div} E_L = 2\pi \left( \sum_{p \in \Lambda} \delta_p - \mathbf{1}_L \right).$$

*In particular  $E \cdot \vec{\nu} = 0$  on  $\partial L$ , there is no singular part of the divergence on  $\partial L$  and thus  $\#\Lambda = |L|$ .*

*iii) If  $d(x, L^c) > R^{\frac{3}{4}}$  then  $E_L(x) = E(x)$*

*iv)*

$$(7.4) \quad \frac{W(E_L, \mathbf{1}_L)}{|L|} \leq W(E) + \varepsilon.$$

We note that if  $E$  is such that  $\operatorname{div} E = 2\pi \sum_p \delta_p - 1$  and we have  $\operatorname{curl} E = 0$  in a neighborhood of each  $p \in \Lambda$ , then the definition (1.29) still makes sense, in particular the limit exists. This is what is meant by  $W$  in (7.4), as well as in the rest of the section.

The next lemma explains how to partition  $\Sigma$  into rectangles. The main point is to cut  $\Sigma'$  into stripes and then each stripe into rectangles in such a way that  $\int m'_0$  over each rectangle is a large integer.

**Lemma 7.5.** *There exists a constant  $C_0 > 0$  such that, given any  $R > 1$  and  $q \in \mathbb{N}^*$ , there exists for any  $n \in \mathbb{N}^*$  a collection  $\mathcal{K}_n$  of closed rectangles in  $\Sigma'$  with disjoint interiors, whose sidelengths are between  $\bar{R} = 2qR$  and  $\bar{R} + C_0\bar{R}/R^2$ , and which are such that*

$$\{x \in \Sigma' : d(x, \partial\Sigma') \leq \bar{R}\} \subset \Sigma' \setminus \bigcup_{K \in \mathcal{K}_n} K \subset \{x \in \Sigma' : d(x, \partial\Sigma') \leq C_0\bar{R}\}$$

*and, for all  $K \in \mathcal{K}_n$ ,*

$$(7.5) \quad \int_K m'_0 \in q^2\mathbb{N}.$$



*Proof.* For each  $j \in \mathbb{Z}$  we let

$$m_j(t) = \int_{x=-\infty}^t \int_{y=j\bar{R}}^{(j+1)\bar{R}} m_0'(x, y) dy dx.$$

Then each strip  $\{j\bar{R} \leq y < (j+1)\bar{R}\}$  is cut into rectangles  $[t_{ij}, t_{(i+1),j}] \times [j\bar{R}, (j+1)\bar{R}]$  where  $t_{0j} = -\infty$  and

$$t_{i+1,j} = \min\{t \geq t_{ij} + \bar{R} : m_j(t_{ij}) \in q^2\mathbb{N}\}.$$

Since by assumption (1.12) we have  $m_0'(x) \in [\underline{m}, \bar{m}]$  for any  $x \in \Sigma'$ , it is not difficult to check that if such a rectangle is included in  $\Sigma'$  then

$$t_{ij} + \bar{R} \leq t_{i+1,j} \leq t_{ij} + \bar{R} + \frac{q^2}{\underline{m}\bar{R}},$$

and thus its sidelengths are between  $\bar{R}$  and  $\bar{R} + C\bar{R}/R^2$  since  $\bar{R}/R^2 = 4q^2/\bar{R}$ . We let  $\mathcal{K}_n$  be the set of rectangles of the form  $[t_{ij}, t_{(i+1),j}] \times [j\bar{R}, (j+1)\bar{R}]$  which are included in  $\{x : d(x, \partial\Sigma') > \bar{R}\}$ . From the above, it follows that these rectangles in fact cover the set  $\{x : d(x, \partial\Sigma') > C\bar{R}\}$  for some  $C > 0$  independent of  $R > 1, q$ . By construction each  $K \in \mathcal{K}_n$  is such that

$$\int_K m_0' = m_j(t_{(i+1),j}) - m_j(t_{ij}) \in q^2\mathbb{N}.$$

□

The next lemma explains how to select a good subset of  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$ .

**Lemma 7.6.** *Let  $\tilde{P}$  be a translation invariant measure on  $X$  such that  $\tilde{P}$ -a.e.  $E \in \mathcal{A}_1$  and  $W(E) < \infty$ . Then for any  $\varepsilon > 0$ , for any  $R_\varepsilon > 0$ , there exist subsets  $H_\varepsilon \subset G_\varepsilon$  in  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  which are compact and such that*

i)  $\eta_0$  being given by Lemma 7.1 we have

$$(7.6) \quad \tilde{P}(\Sigma \times G_\varepsilon^c) < \min(\eta_0^2, \eta_0\varepsilon), \quad \tilde{P}(\Sigma \times H_\varepsilon^c) < \min(\eta_0, \varepsilon).$$

ii) For every  $E \in H_\varepsilon$ , there exists  $\Gamma(E) \subset K_{\bar{m}R_\varepsilon}$  such that

$$(7.7) \quad |\Gamma(E)| < CR_\varepsilon^2\eta_0 \text{ and } \lambda \notin \Gamma(E) \implies \theta_\lambda E \in G_\varepsilon.$$

iii) The convergence in the definition of  $W(E)$  is uniform w.r.t.  $(x, E) \in G_\varepsilon$  and, writing  $\text{div } E = 2\pi(\nu - 1)$ ,

$$(7.8) \quad W(E) \text{ and } \frac{\nu(K_R)}{R^2} \text{ are bounded uniformly w.r.t. } (x, E) \in G_\varepsilon \text{ and } R > 1.$$

iv) We have

$$(7.9) \quad d_B(P, P'') < C\varepsilon(|\log \varepsilon| + 1), \quad \text{where}$$

$$P'' = \int_{\Sigma \times H_\varepsilon} \frac{1}{m_0(x)|K_{R_\varepsilon}|} \int_{\sqrt{m_0(x)}K_{R_\varepsilon} \setminus \Gamma(E)} \delta_x \otimes \delta_{\sigma_{m_0(x)}\theta_\mu E} d\mu d\tilde{P}(x, E).$$

Moreover, there exists a partition of  $H_\varepsilon$  into  $\cup_{i=1}^{N_\varepsilon} H_\varepsilon^i$  satisfying  $\text{diam}(H_\varepsilon^i) < \eta_3$ , where  $\eta_3$  is such that

$$(7.10) \quad E \in H_\varepsilon, d_p(E, E') < \eta_3, m \in [\underline{m}, \overline{m}], \mu \in \sqrt{\overline{m}} K_{R_\varepsilon} \setminus \Gamma(E) \implies d_{\mathcal{B}}(\delta_{\sigma_m \theta_\mu E}, \delta_{\sigma_m \theta_\mu E'}) < \varepsilon;$$

and there exists for all  $i$ ,  $J_i \in H_\varepsilon^i$  such that

$$(7.11) \quad W(J_i) < \inf_{H_\varepsilon^i} W + \varepsilon.$$

At this point, denoting  $\tilde{Q}$  the projection of  $\tilde{P}$  under  $E \mapsto \frac{1}{2\pi} \text{div} E + 1$ , we may always choose  $J_i$  such that  $W(J_i) < \inf_{H_\varepsilon^i} \mathbb{W} + \varepsilon$ .

*Proof.*

*Step 1: Choice of  $G_\varepsilon$ .* Since  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  is Polish we can always find a compact set  $G_\varepsilon$  satisfying (7.6) and  $P(G_\varepsilon^c) < \eta_0$ . Then from Lemma 7.1,  $P \llcorner G_\varepsilon$  (the restriction of  $P$  to  $G_\varepsilon$ ) satisfies  $d_{\mathcal{B}}(P, P \llcorner G_\varepsilon) < \varepsilon$ .

From the translation invariance of  $\tilde{P}$  and for any  $\lambda$ , we have  $\tilde{P}(\Sigma \times \theta_\lambda G_\varepsilon) > 1 - \eta_0$  and therefore  $d_{\mathcal{B}}(\tilde{P}, \tilde{P} \llcorner \theta_\lambda G_\varepsilon) < \varepsilon$ . In view of (7.1), it follows that for any  $\lambda \in \mathbb{R}^2$  we have  $\|P - P_\lambda\| < \eta_0$  and then  $d_{\mathcal{B}}(P, P_\lambda) < \varepsilon$ , where

$$P_\lambda = \int_{\Sigma \times \theta_\lambda G_\varepsilon} \delta_x \otimes \delta_{\sigma_{m_0(x)} E} d\tilde{P}(x, j) = \int_{\Sigma \times G_\varepsilon} \delta_x \otimes \delta_{\theta_\lambda \sigma_{m_0(x)} E} d\tilde{P}(x, E).$$

Then using Lemma 7.3 we deduce that if  $A \subset \mathbb{R}^2$  is any measurable set of positive measure, then

$$(7.12) \quad d_{\mathcal{B}}(P, P') < C\varepsilon(|\log \varepsilon| + 1), \quad \text{where} \quad P' = \int_{\Sigma \times G_\varepsilon} \int_A \delta_x \otimes \delta_{\theta_\lambda \sigma_{m_0(x)} E} d\lambda d\tilde{P}(x, E).$$

Moreover, since  $P$  is  $T_{\lambda(x)}$ -invariant, choosing  $\chi$  to be a smooth positive function with integral 1 supported in  $B(0, 1)$ , the ergodic theorem (as in Section 6.1 or see again [Bec]) ensures that for  $P$ -almost every  $(x, E)$  the limit

$$\lim_{R \rightarrow +\infty} \frac{1}{|K_R|} \int_{K_R} W(E(\lambda + \cdot), \chi(\lambda + \cdot)) d\lambda$$

exists. Then  $\mathbf{1}_{K_R} * \chi$  is a family of functions which satisfies (1.31) with respect to the family of squares  $\{K_R\}_R$ , and from the definition of the renormalized energy relative to  $\{K_R\}_R$  we may rewrite the limit above as

$$(7.13) \quad W(E) = \lim_{R \rightarrow +\infty} \frac{1}{|K_R|} W(E, \mathbf{1}_{K_R} * \chi).$$

By Egoroff's theorem we may choose the compact set  $G_\varepsilon$  above to be such that, in addition to (7.12), the convergence in (7.13) is uniform on  $G_\varepsilon$ . In fact, since  $W(E) < +\infty$  and  $\limsup_R \nu(K_R)/R^2 < +\infty$  for  $P$ -a.e.  $(x, E)$ , where  $\text{div} E = 2\pi(\nu - 1)$ , we may choose  $G_\varepsilon$  such that (7.8) holds.

The arguments above show that the properties (7.12), (7.8) can be satisfied for a compact set  $G_\varepsilon$  of measure arbitrarily close to 1. We choose  $G_\varepsilon$  such that (7.6) holds.

The next difficulty we have to face is that  $\theta_\lambda E$  need not belong to  $G_\varepsilon$  if  $E$  does.

*Step 2: Choice of  $H_\varepsilon$ .* For  $E \in G_\varepsilon$ , let  $\Gamma(E)$  be the set of  $\lambda$ 's in  $\sqrt{m}K_{R_\varepsilon}$  such that  $\theta_\lambda E \notin G_\varepsilon$ . Since, from (7.6) and the translation-invariance of  $\tilde{P}$ , for any  $\lambda \in \mathbb{R}^2$  we have  $\tilde{P}(\Sigma \times \theta_\lambda(G_\varepsilon)^c) < \eta_0^2$ , it follows from Fubini's theorem that

$$\int_{G_\varepsilon} |\Gamma(E)| d\tilde{P}(x, E) = \int_{\sqrt{m}K_{R_\varepsilon}} \tilde{P}(\Sigma \times (\theta_\lambda G_\varepsilon)^c) d\lambda < 4\bar{m}R_\varepsilon^2 \min(\eta_0^2, \eta_0\varepsilon).$$

Therefore, letting

$$(7.14) \quad H_\varepsilon = \{E \in G_\varepsilon : |\Gamma(E)| < 4\bar{m}R_\varepsilon^2\eta_0\},$$

we have that (7.6) holds.

Combining (7.6) and (7.14) with Lemma 7.1, we deduce from (7.12) that (7.9) holds, where we have used the fact that  $\theta_\lambda \sigma_m E = \sigma_m \theta_{\sqrt{m}\lambda} E$  to change the integration variable to  $\mu = \sqrt{m_0(x)}\lambda$  in (7.12).

*Step 3: Choice of  $J_1, \dots, J_{N_\varepsilon}$ .* We use the fact that  $G_\varepsilon$  is relatively compact in  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$ , Lemma 7.2, and the fact that  $(m, E) \mapsto \sigma_m E$  is continuous to find that there exists  $\eta_4 > 0$  such that for any  $m \in [\underline{m}, \bar{m}]$  and any  $E \in G_\varepsilon$  it holds that

$$(7.15) \quad d_p(E, E') < \eta_4 \implies d_{\mathcal{B}}(\delta_{\sigma_m E}, \delta_{\sigma_m E'}) < \varepsilon.$$

Moreover, from the continuity of  $(\mu, E) \mapsto \theta_\mu E$ , there exists  $\eta_3 > 0$  such that

$$(7.16) \quad E \in G_\varepsilon, d_p(E, E') < \eta_3, \mu \in \sqrt{m}K_{R_\varepsilon} \implies d_p(\theta_\mu E, \theta_\mu E') < \eta_4.$$

If  $E \in H_\varepsilon$  and  $\mu \in K \setminus \Gamma(E)$  then  $\theta_\mu E \in G_\varepsilon$  hence applying (7.15), we get (7.10).

Now we cover the relatively compact set  $H_\varepsilon$  by a finite number of balls  $B_1, \dots, B_{N_\varepsilon}$  of radius  $\eta_3/2$ , derive from it a partition of  $H_\varepsilon$  by sets with diameter less than  $\eta_3$ , by letting  $H_\varepsilon^1 = B_1 \cap H_\varepsilon$  and

$$H_\varepsilon^{i+1} = B_{i+1} \cap H_\varepsilon \setminus (B_1 \cup \dots \cup B_i).$$

we then have

$$(7.17) \quad H_\varepsilon = \bigcup_{i=1}^{N_\varepsilon} H_\varepsilon^i, \quad \text{diam}(H_\varepsilon^i) < \eta_3,$$

where the union is disjoint. Then we may choose  $J_i \in H_\varepsilon^i$  such that (7.11) holds.  $\square$

### 7.3 Completing the construction

*Step 1: Choice of  $R_\varepsilon$ .* We apply Proposition 7.4 with  $G = G_\varepsilon$  and  $\sqrt{m}R$ , where  $m \in [\underline{m}, \bar{m}]$ . The proposition yields  $\eta_2 > 0$ ,  $R_\varepsilon > 1$  such that for any  $E \in G_\varepsilon$  and any  $m \in [\underline{m}, \bar{m}]$  and any rectangle  $L$  centered at 0 with sidelengths in  $[2\sqrt{m}R_\varepsilon, 2\sqrt{m}R_\varepsilon(1 + \eta_2)]$ , (7.4) is satisfied for some  $E_L$ , with  $R$  replaced by  $\sqrt{m}R_\varepsilon$ . The reason for including  $\sqrt{m}$  is that we will need to scale the construction to account for the varying weight  $m_0(x)$ .

Since our rectangles will be obtained from Lemma 7.5 and we wish to use the approximation by  $E_L$  in them, we choose  $R_\varepsilon$  large enough so that if  $m \in [\underline{m}, \bar{m}]$  and  $L$  is a rectangle centered at zero with sidelengths in  $[2\sqrt{m}R_\varepsilon, 2\sqrt{m}R_\varepsilon(1 + C_0/R_\varepsilon^2)]$  then

$$(7.18) \quad \frac{C_0}{R_\varepsilon^2} < \eta_2, \quad \frac{C_1}{R_\varepsilon^2} < \eta_0, \quad K_{\sqrt{m}R_\varepsilon(1-\eta_0)} \subset \{x : d(x, L^c) > \sqrt{m}R_\varepsilon^{\frac{3}{4}}\} \subset K_{\sqrt{m}R_\varepsilon(1+\eta_0)},$$

where  $C_0$  is the constant in Lemma 7.5,  $C_1 \geq 1$  is to be determined later, and  $\eta_0$  is the constant in Lemma 7.1.

If  $\lambda \in K_{\sqrt{m}R_\varepsilon(1-\eta_0)}$  and since  $E = E_L$  if  $d(x, L^c) > \sqrt{m}R_\varepsilon^{\frac{3}{4}}$ , we deduce from (7.18) that  $\theta_\lambda E_L = \theta_\lambda E$  in  $B(0, \sqrt{m}R_\varepsilon^{\frac{3}{4}})$ , so that from the definition of  $d_p$ , taking  $R_\varepsilon$  larger if necessary,

$$(7.19) \quad \forall E \in G_\varepsilon, m \in [\underline{m}, \overline{m}], \lambda \in K_{\sqrt{m}R_\varepsilon(1-\eta_0)}, \quad d_p(\theta_\lambda \sigma_m E, \theta_\lambda \sigma_m E_L) < \frac{\eta_1}{2},$$

where  $\eta_1$  comes from Lemma 7.2 applied on  $\{\sigma_m E : m \in [\underline{m}, \overline{m}], E \in G_\varepsilon\}$ , i.e. is such that

$$(7.20) \quad m \in [\underline{m}, \overline{m}], E \in G_\varepsilon, E' \in L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2) \text{ and } d_p(E, E') < \eta_1 \implies d_{\mathcal{B}}(\delta_{\sigma_m E}, \delta_{\sigma_m E'}) < \varepsilon.$$

*Step 2: Choice of  $q_\varepsilon$  and the rectangles.* We choose an integer  $q_\varepsilon$  large enough so that

$$(7.21) \quad \frac{N_\varepsilon}{C_1 q_\varepsilon^2} < \eta_0, \quad \frac{N_\varepsilon}{q_\varepsilon^2} \times \max_{\substack{0 \leq i \leq N_\varepsilon \\ \underline{m} \leq m \leq \overline{m}}} W_K(\sigma_m J_i) < \varepsilon$$

where  $C_1 > 1$  is to be determined later. We apply Lemma 7.5 with  $R_\varepsilon, q_\varepsilon$  and  $N_\varepsilon$  to obtain for any  $n$  a collection  $\mathcal{K}_n$  of rectangles (we omit to mention the  $\varepsilon$  dependence) which cover most of  $\Sigma'$ , and we also apply Lemma 7.6. We rewrite  $P''$  given by (7.9) as

$$(7.22) \quad P'' = \sum_{K \in \mathcal{K}_n} \int_{\frac{K}{\sqrt{n}} \times H_\varepsilon} \frac{1}{m_0(x)|K_{R_\varepsilon}|} \int_{\sqrt{m_0(x)}K_{R_\varepsilon} \setminus \Gamma(E)} \delta_x \otimes \delta_{\sigma_{m_0(x)}\theta_\mu E} d\mu d\tilde{P}(x, E).$$

Now we claim that if  $n$  is large enough and  $x \in K/\sqrt{n}, E \in H_\varepsilon^i, \mu \in \sqrt{m_0(x)}K_{R_\varepsilon} \setminus \Gamma(E)$ , then

$$(7.23) \quad d_{\mathcal{B}} \left( \delta_x \otimes \delta_{\sigma_{m_0(x)}\theta_\mu E}, \delta_{x_K} \otimes \delta_{\sigma_{m_K}\theta_\mu J_i} \right) < 2\varepsilon,$$

where  $x_K$  is the center of  $K/\sqrt{n}$  and  $m_K$  is the average of  $m_0$  over  $K/\sqrt{n}$ . Indeed, since  $m_0$  is  $C^1$  we have  $|x - x_K| < C/\sqrt{n}, |m_0(x) - m_K| < C/\sqrt{n}$  thus if  $n$  is large enough, since  $\theta_\mu E \in G_\varepsilon$  we find

$$d_{\mathcal{B}} \left( \delta_x \otimes \delta_{\sigma_{m_0(x)}\theta_\mu E}, \delta_{x_K} \otimes \delta_{\sigma_{m_K}\theta_\mu E} \right) < \varepsilon.$$

Moreover, since  $d_p(E, J_i) < \eta_3$ , we deduce from (7.10) that

$$d_{\mathcal{B}} \left( \delta_{x_K} \otimes \delta_{\sigma_{m_K}\theta_\mu E}, \delta_{x_K} \otimes \delta_{\sigma_{m_K}\theta_\mu J_i} \right) < \varepsilon,$$

which together with the previous estimate proves (7.23).

Using (7.23) together with Lemmas 7.2, 7.1, and (7.7), we deduce from (7.9) and (7.22) that  $d_{\mathcal{B}}(P, P''') < C\varepsilon(|\log \varepsilon| + 1)$ , where

$$(7.24) \quad \begin{aligned} P''' &= \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq i \leq N_\varepsilon}} \int_{\frac{K}{\sqrt{n}} \times H_\varepsilon^i} \int_{\sqrt{m_K}K_{R_\varepsilon}} \delta_{x_K} \otimes \delta_{\sigma_{m_K}\theta_\mu J_i} d\mu d\tilde{P}(x, E) \\ &= \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq i \leq N_\varepsilon}} p_{i,K} \int_{\sqrt{m_K}K_{R_\varepsilon}} \delta_{x_K} \otimes \delta_{\sigma_{m_K}\theta_\mu J_i} d\mu, \end{aligned}$$

where

$$(7.25) \quad p_{i,K} = \tilde{P} \left( \frac{K}{\sqrt{n}} \times H_\varepsilon^i \right).$$

*Step 3: Choice of subrectangles and vector field  $E_n$ .* We now replace  $p_{i,K}$  in the definition (7.24) by

$$(7.26) \quad \frac{|K|}{q_\varepsilon^2 |\Sigma'|} n_{i,K}, \quad \text{where} \quad n_{i,K} = \left\lfloor \frac{q_\varepsilon^2 |\Sigma'|}{|K|} p_{i,K} \right\rfloor.$$

We have, since  $\tilde{P}(\frac{K}{\sqrt{n}} \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)) = |K|/|\Sigma'|$ ,

$$(7.27) \quad \sum_{k=1}^{N_\varepsilon} n_{i,K} \leq \frac{q_\varepsilon^2 |\Sigma'|}{|K|} \tilde{P} \left( \frac{K}{\sqrt{n}} \times L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2) \right) = q_\varepsilon^2$$

and

$$\left| \frac{|K_{R_\varepsilon}|}{|\Sigma'|} n_{i,K} - p_{i,K} \right| < C \left( \frac{|K|}{q_\varepsilon^2 |E'|} + \frac{n_{i,K}}{R_\varepsilon^2 |\Sigma'|} \right).$$

Summing with respect to  $i$  and  $K$ , using the facts that  $\sum_{K \in \mathcal{K}_n} |K| < |\Sigma'|$ , (7.27), and the fact that the cardinal of  $\mathcal{K}_n$  is  $\frac{|\Sigma'|}{4q_\varepsilon^2 R_\varepsilon^2}$ , we find

$$\sum_{1 \leq i \leq N_\varepsilon, K \in \mathcal{K}_n} \left| \frac{|K_{R_\varepsilon}|}{|\Sigma'|} n_{i,K} - p_{i,K} \right| < C \left( \frac{N_\varepsilon}{q_\varepsilon^2} + \frac{1}{R_\varepsilon^4} \right).$$

We may always choose  $C_1$  large enough in (7.18) and (7.21) so that the right-hand side is  $< \eta_0$ . Then Lemma 7.1 implies that  $d_{\mathcal{B}}(P, P^{(4)}) < C\varepsilon(|\log \varepsilon| + 1)$  is still true after replacing  $p_{i,K}$  by  $\frac{|K_{R_\varepsilon}|}{|\Sigma'|} n_{i,K}$  in (7.24), i.e. where

$$(7.28) \quad P^{(4)} = \frac{1}{|\Sigma'|} \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq i \leq N_\varepsilon}} \frac{n_{i,K}}{m_K} \int_{\sqrt{m_K} K_{R_\varepsilon}} \delta_{x_K} \otimes \delta_{\sigma_{m_K} \theta_{\mu, J_i}} d\mu.$$

Next, we divide each  $K \in \mathcal{K}_n$  into a collection  $\mathcal{L}_K$  of  $q_\varepsilon^2$  identical subrectangles in the obvious way and we partition  $\mathcal{L}_K$  into collections  $\mathcal{L}_{K,i}$ ,  $0 \leq i \leq N_\varepsilon$  such that if  $k \geq 1$  then  $\mathcal{L}_{K,i}$  contains  $n_{i,K}$  subrectangles. This is clearly possible from (7.27). If the inequality is strict we put the extra subrectangles in  $\mathcal{L}_{K,0}$ , there will be  $n_{0,K}$  of them and then

$$(7.29) \quad \sum_{k=0}^{N_\varepsilon} n_{k,K} = q_\varepsilon^2.$$

We rewrite (7.28) as

$$(7.30) \quad P^{(4)} = \frac{1}{|\Sigma'|} \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq i \leq N_\varepsilon \\ \tilde{L} \in \mathcal{L}_{K,i}}} \frac{1}{m_K} \int_{\sqrt{m_K} K_{R_\varepsilon}} \delta_{x_K} \otimes \delta_{\sigma_{m_K} \theta_{\mu, J_i}} d\mu.$$

Now, for  $\tilde{L} \in \mathcal{L}_{K,i}$ , let  $L = \sqrt{m_K}(\tilde{L} - x_{\tilde{L}})$ , where  $x_{\tilde{L}}$  denotes the center of  $\tilde{L}$ . From Lemma 7.5, a rectangle  $K \in \mathcal{K}_n$  has sidelengths between  $2q_\varepsilon R_\varepsilon$  and  $2q_\varepsilon R_\varepsilon(1+C_0/R_\varepsilon^2)$ . Therefore  $L$  is a rectangle centered at zero with sidelengths between  $2\sqrt{m_K}R_\varepsilon$  and  $2\sqrt{m_K}R_\varepsilon(1+C_0/R_\varepsilon^2)$ , and (7.19) holds.

This, and the results of Lemma 7.6, allow us to apply Proposition 7.4 on  $L$  to any  $J_i$ ,  $1 \leq i \leq N_\varepsilon$ . Note that  $|L| \in \mathbb{N}$  follows from the fact that

$$|L| = m_K |\tilde{L}| = \int_K m_0' \frac{|K|}{q_\varepsilon^2} = \frac{1}{q_\varepsilon^2} \int_K m_0'$$

and (7.5). In this way, we define currents  $J_{i,L}$  which satisfy (7.4) and (7.19). We claim that, as a consequence of the latter, we have

(7.31)

$$E' = J_{i,L} \text{ on } L \implies d_{\mathcal{B}} \left( \int_{\sqrt{m_K}K_{R_\varepsilon}} \delta_{x_K} \otimes \delta_{\sigma_{m_K} \theta_\mu J_i} d\mu, \frac{1}{m_K |K_{R_\varepsilon}|} \int_L \delta_{x_K} \otimes \delta_{\sigma_{m_K} \theta_\mu E'} d\mu \right) < C\varepsilon.$$

This goes as follows: (i) Using Lemma 7.1 and (7.7),(7.18), we find that integrating on  $\sqrt{m_K}K_{(1-\eta_0)R_\varepsilon} \setminus \Gamma(J_i)$  instead of  $\sqrt{m_K}K_{R_\varepsilon}$  and  $L$  induces an error of  $C\varepsilon$ . (ii) From (7.19), and (7.20) applied to  $\theta_\mu J_i$  and  $\theta_\mu E'$  we have  $d_{\mathcal{B}}(\delta_{\theta_\mu J_i}, \delta_{\theta_\mu E'}) < \varepsilon$  and thus in view of Lemma 7.3 we may replace  $\theta_\mu J_i$  by  $\theta_\mu E'$  in the integral with an error of  $C\varepsilon|\log \varepsilon|$  at most. (iii) Using (7.18), (7.7) and Lemma 7.1 again, we may integrate back on  $\sqrt{m_K}K_{R_\varepsilon}$  and  $L$  rather than on  $K_{(1-\eta_0)R_\varepsilon} \setminus \Gamma(J_i)$ , with an additional error of  $C\varepsilon$ . this proves (7.31).

Combining (7.31) with (7.30) and  $d_{\mathcal{B}}(P, P^{(4)}) < C\varepsilon(|\log \varepsilon| + 1)$ , using Lemma 7.3 we find  $d_{\mathcal{B}}(P, P^{(5)}) < C\varepsilon(|\log \varepsilon| + 1)$ , where

(7.32)

$$P^{(5)} = \frac{1}{|\Sigma'|} \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq i \leq N_\varepsilon \\ \tilde{L} \in \mathcal{L}_{K,i}}} \frac{1}{m_K} \int_L \delta_{x_K} \otimes \delta_{\sigma_{m_K} \theta_\mu \tilde{J}_{i,L}} d\mu = \frac{1}{|\Sigma'|} \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq i \leq N_\varepsilon \\ \tilde{L} \in \mathcal{L}_{K,i}}} \int_{L/\sqrt{m_K}} \delta_{x_K} \otimes \delta_{\theta_\lambda \sigma_{m_K} \tilde{J}_{i,L}} d\lambda,$$

where the last equality follows by changing variables to  $\lambda = \mu/\sqrt{m_K}$ , and where  $\tilde{J}_{i,L}$  denotes an arbitrarily chosen element of  $L_{\text{loc}}^p(\mathbb{R}^2, \mathbb{R}^2)$  such that  $\tilde{J}_{i,L} = J_{i,L}$  on  $L$ , the constant  $C$  being independent of this choice.

If we choose an arbitrary  $J_0$  in  $\mathcal{A}_1$  and let the sum in (7.32) range over  $0 \leq i \leq N_\varepsilon$  instead of  $1 \leq i \leq N_\varepsilon$  we obtain a measure  $P^{(6)}$  such that, by (7.21),

$$\|P^{(5)} - P^{(6)}\| \leq \frac{1}{|\Sigma'|} \sum_{K \in \mathcal{K}_n} \frac{N_\varepsilon |K|}{q_\varepsilon^2} \leq \eta_0,$$

hence using Lemma 7.1 we have  $d_{\mathcal{B}}(P^{(5)}, P^{(6)}) < \varepsilon$  and then  $d_{\mathcal{B}}(P, P^{(6)}) < C\varepsilon(|\log \varepsilon| + 1)$ .

We now define the vector field  $E_n^{\text{int}} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  by letting  $E_n^{\text{int}}(x) = \sigma_{m_K} J_{i,L}(x - x_{\tilde{L}})$  on  $\tilde{L} = x_{\tilde{L}} + L/\sqrt{m_K}$ , for every  $K \in \mathcal{K}_n$ ,  $0 \leq i \leq N_\varepsilon$  and  $\tilde{L} \in \mathcal{L}_{K,i}$ . Then, for every  $L \in \mathcal{L}_{K,i}$  we have  $E_n^{\text{int}}(x_{\tilde{L}} + \cdot) = \sigma_{m_K} J_{i,L}$  on  $\tilde{L}$ , therefore we may choose  $\tilde{J}_{i,L} = \sigma_{1/m_K} E_n^{\text{int}}(x_{\tilde{L}} + \cdot)$  in (7.32) and then then we may summarize the above by writing

$$(7.33) \quad d_{\mathcal{B}}(P, P^{(6)}) < C\varepsilon(|\log \varepsilon| + 1), \quad P^{(6)} = \frac{1}{|\Sigma'|} \sum_{K \in \mathcal{K}_n} \int_K \delta_{x_K} \otimes \delta_{\theta_\lambda E_n^{\text{int}}} d\lambda.$$

Note that since  $J_{i,L} = 0$  outside  $L$ , we also have

$$(7.34) \quad E_n^{\text{int}} = \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq i \leq N_\varepsilon \\ \tilde{L} \in \mathcal{L}_{K,i}}} \sigma_{m_K} J_{i,L}(\cdot - x_{\tilde{L}}), \quad \text{div } E_n^{\text{int}} = 2\pi \sum_{\substack{K \in \mathcal{K}_n \\ p \in \Lambda_K}} (\delta_p - m_K),$$

where  $\Lambda_K$  is a finite subset of the interior of  $K$ . The second equation is satisfied in the sense of distributions on  $\mathbb{R}^2$ .

*Step 4: Treating the boundary.* Let  $\hat{\Sigma}' := \Sigma' \setminus \cup_{K \in \mathcal{K}_n} K$ . We let  $t \in [0, \ell\sqrt{n}]$  denote arclength on  $\partial\Sigma'$  — where  $\ell$  is the length of  $\partial\Sigma$  — and  $s$  denote the distance to  $\partial\Sigma'$ , so that  $(t, s)$  is a  $C^1$  coordinate system on  $\{x \in \Sigma' : d(x, (\Sigma')^c) < c\sqrt{n}\}$ , if  $c > 0$  is small enough, since the boundary of  $\Sigma$  is  $C^1$  by (1.13). We let  $C_t$  denote the curvilinear rectangle of points with coordinates in  $[0, t] \times [0, C\bar{R}_\varepsilon]$ , where  $\bar{R}_\varepsilon = q_\varepsilon R_\varepsilon$  and  $C$  is large enough so that  $\hat{\Sigma}' \subset \{x \in \Sigma' : d(x, \partial\Sigma') < C\bar{R}_\varepsilon\}$ , and define  $m(t) = \int_{C_t \cap \hat{\Sigma}'} m_0'$ . Since the distance of  $\cup_{K \in \mathcal{K}_n} K$  to a given  $x \in \partial\Sigma'$  is between  $\bar{R}_\varepsilon$  and  $C_0\bar{R}_\varepsilon$  from Lemma 7.5 and since  $m_0'$  is bounded above and below by (1.16), the derivative of  $t \mapsto m(t)$  is between  $\bar{R}_\varepsilon/C$  and  $C\bar{R}_\varepsilon$  for some  $C > 0$  large enough.

We let

$$(7.35) \quad k_\varepsilon = \left\lfloor \frac{\ell\sqrt{n}}{\bar{R}_\varepsilon} \right\rfloor$$

and choose  $0 = t_0, \dots, t_{k_\varepsilon} = \ell\sqrt{n}$  to be such that

$$m(t_l) = \left\lfloor \frac{l}{k_\varepsilon} m(\ell\sqrt{n}) \right\rfloor.$$

We note that indeed  $t_{k_\varepsilon} = \ell\sqrt{n}$ : Since the integral of  $m_0'$  on each square  $K \in \mathcal{K}_n$  is an integer as well as the integral on  $\Sigma'$ , we have  $\int_{\hat{\Sigma}'} m_0' \in \mathbb{N}$  and therefore  $m(\ell\sqrt{n}) \in \mathbb{N}$ .

From the above remark about the derivative of  $t \rightarrow m(t)$ , we deduce that  $\frac{m(\ell\sqrt{n})}{\ell\sqrt{n}}$  belongs to the interval  $[\bar{R}_\varepsilon/C, C\bar{R}_\varepsilon]$  for some  $C > 0$  and then it is easy to deduce that if  $\sqrt{n}$  is large enough compared to  $\bar{R}_\varepsilon$  then

$$n_l := m(t_{l+1}) - m(t_l) \in \left[ \bar{R}_\varepsilon^2/C, C\bar{R}_\varepsilon^2 \right], \quad t_{l+1} - t_l \in [\bar{R}_\varepsilon/C, C\bar{R}_\varepsilon].$$

This means that the sidelengths of the curvilinear rectangle  $C_{t_{l+1}} \setminus C_{t_l}$  are comparable to  $\bar{R}_\varepsilon$ , and that the number of points  $n_l$  to put there in is of order  $\bar{R}_\varepsilon^2$ .

We may then include each of the sets  $K_l := \hat{\Sigma}' \cap (C_{t_{l+1}} \setminus C_{t_l})$  in a ball  $B_l$  with radius in  $[\bar{R}_\varepsilon/C, C\bar{R}_\varepsilon]$  and we may also choose a set of  $n_l$  points  $\Lambda_l$  which are at distance at least  $1/C$  from each other and the complement of  $K_l$ . Let  $E_l = -\nabla H$ , where  $H$  solves  $-\Delta H = 2\pi(\sum_{p \in \Lambda_l} \delta_p - m_l)$  in  $B_l$  and  $\nabla H \cdot \vec{\nu} = 0$  on  $\partial B_l$ , where

$$m_l = \frac{n_l}{|K_l|} \mathbf{1}_{K_l}.$$

Then we have  $\text{div } E_l = 2\pi(\sum_{p \in \Lambda_l} \delta_p - m_l)$  in  $B_l$  and  $E_l \cdot \vec{\nu} = 0$  on  $\partial B_l$  and we claim that for any  $q \geq 1$ ,

$$(7.36) \quad W(E_l, \mathbf{1}_{B_l}) \leq C_\varepsilon, \quad \|E_l\|_{L^q(B_l \setminus K_l)} \leq C_{\varepsilon, q},$$

where the constants do not depend on  $n$ , but do depend on  $\varepsilon$  through  $\bar{R}_\varepsilon$ . This is proved by noting that these quantities are finite, and that a compactness argument shows that the bound is uniform for any choice of points which are at distance at least  $1/C$  from each other and the complement of some  $K_l \subset B_l$ , using for instance the explicit formulas for  $W$  in [LR]. Note that because the sets  $\{K_l\}$  and the rectangles  $\{K\}$  are disjoint, have measure between  $\bar{R}_\varepsilon^2/C$  and  $C\bar{R}_\varepsilon^2$  and diameter between  $\bar{R}_\varepsilon/C$  and  $C\bar{R}_\varepsilon$ , we know that their overlap is bounded by a constant  $C$  independent of  $\varepsilon, n$ .

*Step 5: Rectification of the weight.* We rectify the weights  $m_K, m_l$ : For  $K \in \mathcal{K}_n$  we let  $H_K$  solve  $-\Delta H_K = 2\pi(m_0' - m_K)$  on  $K$  and  $\nabla H_K \cdot \vec{\nu} = 0$  on  $\partial K$ . Similarly we let  $H_l$  solve  $-\Delta H_l = 2\pi(m_0' \mathbf{1}_{K_l} - m_l)$ ,  $\nabla H_l \cdot \vec{\nu} = 0$ . By elliptic regularity, we deduce for any  $q > 1$  that  $\|\nabla H_K\|_{L^q(K)}$  (resp.  $\|\nabla H_l\|_{L^q(B_l)}$ ) is bounded by  $C_{q,\varepsilon}\|m_0' - m_K\|_{L^\infty(K)}$  (resp.  $C_{q,\varepsilon}\|m_0' - m_l\|_{L^\infty(B_l)}$ ). Since  $m_0$  is  $C^1$  we have  $|\nabla m_0'| \leq C/\sqrt{n}$ , therefore  $\|m_0' - m_K\|_{L^\infty(K)} \leq C\bar{R}_\varepsilon/\sqrt{n}$ , while  $\|m_0' - m_l\|_{L^\infty(B_l)} \leq C$ . We deduce that

$$(7.37) \quad \|\nabla H_K\|_{L^q} \leq \frac{C_{q,\varepsilon}}{\sqrt{n}}, \quad \|\nabla H_l\|_{L^q} \leq C_{q,\varepsilon}.$$

We let

$$(7.38) \quad E_K = E_n^{\text{int}}|_K,$$

and

$$(7.39) \quad E_n = E_n^{\text{int}} + \sum_{K \in \mathcal{K}_n} -\nabla H_K + \sum_{l=1}^{k_\varepsilon} -\nabla H_l = \sum_{K \in \mathcal{K}_n} E_K - \nabla H_K + \sum_{l=1}^{k_\varepsilon} E_l - \nabla H_l,$$

$$\Lambda_n = \cup_{K \in \mathcal{K}} \Lambda_K \cup_{l=1}^{k_\varepsilon} \Lambda_l,$$

where  $E_K$  and  $\nabla H_K$  are set to 0 outside  $K$  and similarly for  $E_l, \nabla H_l$  outside  $B_l$ . Then  $\text{div } E_n = 2\pi(\sum_{p \in \Lambda_n} \delta_p - m_0')$  in  $\mathbb{R}^2$ . This completes the construction of  $E_n$ .

## 7.4 Estimating the energy

*Step 1: Energy estimate.* We have

$$W(E_K, \mathbf{1}_K) = \sum_{\substack{0 \leq i \leq N_\varepsilon \\ \tilde{L} \in \mathcal{L}_{K,i}}} W(\sigma_{m_K} J_{i,L}(\cdot - x_{\tilde{L}}), \tilde{L}).$$

From (7.4) we find, letting  $L = \sqrt{m_K}(\tilde{L} - x_{\tilde{L}})$ , using (7.29) and  $|L| = |K|/q_\varepsilon^2$ , that

$$(7.40) \quad W(E_K, \mathbf{1}_K) = \sum_{\substack{0 \leq i \leq N_\varepsilon \\ \tilde{L} \in \mathcal{L}_{K,i}}} W(\sigma_{m_K} J_{i,L}, \mathbf{1}_{\tilde{L} - x_{\tilde{L}}}) \leq |K| \left( \sum_{i=0}^{N_\varepsilon} \frac{n_{i,K}}{q_\varepsilon^2} W(\sigma_{m_K} J_i) + C\varepsilon \right).$$

We estimate the integral of  $|E_n|^2$  on  $\mathbb{R}^2 \setminus \cup_{p \in \Lambda_n} B(p, \eta)$ . From (7.39), this integral involves on the one hand the square terms

$$(7.41) \quad \sum_{l=1}^{k_\varepsilon} \int_{(B_l)_\eta} |E_l - \nabla H_l|^2 + \sum_{K \in \mathcal{K}_n} \int_{K_\eta} |E_K - \nabla H_K|^2,$$



where  $K_\eta = K \setminus \cup_{p \in \Lambda_n} B(p, \eta)$  and similarly for  $(B_l)_\eta$ , and on the other hand the rectangle terms

$$\sum_{\substack{K, K' \in \mathcal{K}_n \\ K \neq K'}} \int_{K_\eta \cap K'_\eta} (E_K - \nabla H_K) \cdot (E_{K'} - \nabla H_{K'}) + \sum_{1 \leq l \neq i \leq k_\varepsilon} \cdots + \sum_{\substack{K \in \mathcal{K}_n \\ 1 \leq l \leq k_\varepsilon}} \cdots$$

We estimate the latter as follows: Since the rectangles in  $\mathcal{K}_n$  do not overlap, the first sum is equal to zero. A nonzero rectangle term must involve some  $B_l$ , and moreover a given  $B_l$  can only be present in a number of terms bounded independently of  $n, \varepsilon$  because the overlap of the balls  $B_l$  and the rectangles  $K$  is bounded. Thus from (7.35) we have at most  $C\sqrt{n}/\bar{R}_\varepsilon$  nonzero rectangle terms. Moreover, since the  $K_l$ 's are disjoint, and disjoint from the  $K$ 's, in a rectangle term involving  $B_l \cap K$  the integral can be taken over  $K \setminus K_l$ , and in a term involving  $B_l \cap B_i$  it can be taken over  $(B_l \cap B_i \setminus K_i) \cup (B_i \cap B_l \setminus K_l)$ .

In any case we use Hölder's inequality and the bound  $\|E_l - \nabla H_l\|_{L^q(B_l \setminus K_l)} \leq C_{\varepsilon, q}$  for some  $q > 2$ , which follows from (7.36), (7.37), together with the bound

$$\|E_l - \nabla H_l\|_{L^{q'}(B_l)}, \|E_K - \nabla H_K\|_{L^{q'}(K)} \leq C_{\varepsilon, q},$$

which follows from (7.40), (7.36) using Lemma 5.1, to conclude that each rectangle term is bounded by  $C_\varepsilon$  and then that their sum is  $O(\sqrt{n})$ , meaning a quantity bounded by a constant depending on  $\varepsilon$  times  $\sqrt{n}$ .

The limit as  $\eta \rightarrow 0$  of the terms in (7.41) is estimated as above by expanding the squares and using Hölder's inequality with (7.37), (7.36), (7.40), together with the bound (7.35) to show that

$$\begin{aligned} \lim_{\eta \rightarrow 0} \frac{1}{2} \left( \sum_{l=1}^{k_\varepsilon} \int_{(B_l)_\eta} |E_l - \nabla H_l|^2 + \sum_{K \in \mathcal{K}_n} \int_{K_\eta} |E_K - \nabla H_K|^2 + \pi \# \Lambda_n \log \eta \right) \\ \leq \sum_{K \in \mathcal{K}_n} W(E_K, \mathbf{1}_K) + O(\sqrt{n}). \end{aligned}$$

In view of the bound  $O(\sqrt{n})$  for the rectangle terms and (7.40) we find using (7.29) that

$$(7.42) \quad W(E_n, \mathbf{1}_{\mathbb{R}^2}) \leq \sum_{\substack{K \in \mathcal{K}_n \\ 0 \leq i \leq N_\varepsilon}} |K| \frac{n_{i,K}}{q_\varepsilon^2} W(\sigma_{m_K} J_i) + Cn\varepsilon + O(\sqrt{n}).$$

*Step 2: We proceed to estimating  $W(E_n, \mathbf{1}_{\mathbb{R}^2})$ .* We have, using (7.26), (7.25), (7.21), then the fact that  $m_{0'} - m_K \leq C\bar{R}_\varepsilon/\sqrt{n}$  on  $K$ , then (7.11) with (1.36), then (7.11) and finally (7.1), that

$$\begin{aligned} \sum_{i=1}^{N_\varepsilon} \frac{|K| n_{i,K}}{q_\varepsilon^2} W(\sigma_{m_K} J_i) &\leq |\Sigma'| \sum_{i=1}^{N_\varepsilon} \tilde{P} \left( \frac{K}{\sqrt{n}} \times H_\varepsilon^i \right) W(\sigma_{m_K} J_i) + |K| \varepsilon \\ &\leq |\Sigma'| \sum_{i=1}^{N_\varepsilon} \int_{\frac{K}{\sqrt{n}} \times H_\varepsilon^i} W(\sigma_{m_{0'}(x)} J_i) d\tilde{P}(x, E) + |K| \left( \frac{C}{\sqrt{n}} + \varepsilon \right) \\ &\leq |\Sigma'| \sum_{i=1}^{N_\varepsilon} \int_{\frac{K}{\sqrt{n}} \times H_\varepsilon^i} W(\sigma_{m_{0'}(x)} E) d\tilde{P}(x, E) + |K| \left( \frac{C}{\sqrt{n}} + C\varepsilon \right) \\ (7.43) \quad &= |\Sigma'| \int_{\frac{K}{\sqrt{n}} \times H_\varepsilon} W(E) dP(x, E) + |K| \left( \frac{C}{\sqrt{n}} + C\varepsilon \right). \end{aligned}$$

Here we have used the fact that  $W$  is bounded below by some (negative) constant, a fact proved in [SS1] that we use below several times.

We proceed by estimating  $n_{0,K}$ . From (7.26) we deduce that

$$\sum_{i=1}^{N_\varepsilon} (n_{i,K} + 1) \geq \frac{q_\varepsilon^2 |\Sigma'|}{|K|} \tilde{P} \left( \frac{K}{\sqrt{n}} \times H_\varepsilon \right) \geq \frac{q_\varepsilon^2 |\Sigma'|}{|K|} \left( \frac{|K|}{|\Sigma'|} - \tilde{P} \left( \frac{K}{\sqrt{n}} \times H_\varepsilon^c \right) \right),$$

and then it follows from (7.29) that

$$n_{0,K} = q_\varepsilon^2 - \sum_{i=1}^{N_\varepsilon} n_{i,K} \leq N_\varepsilon + \frac{q_\varepsilon^2 |\Sigma'|}{|K|} \tilde{P} \left( \frac{K}{\sqrt{n}} \times H_\varepsilon^c \right).$$

Summing over  $K \in \mathcal{K}_n$ , using the fact that

$$(7.44) \quad |\Sigma' \setminus \cup_{K \in \mathcal{K}_n} K| < C_\varepsilon \sqrt{n}$$

and then (7.21), (7.6), we find that

$$\sum_{K \in \mathcal{K}} \frac{|K|}{q_\varepsilon^2} n_{0,K} W(\sigma_{m_K} J_0) \leq C |\Sigma'| \left( \tilde{P}(\Sigma \times H_\varepsilon^c) + \frac{1}{\sqrt{n}} + \varepsilon \right) \leq Cn \left( \frac{C_\varepsilon}{\sqrt{n}} + \varepsilon \right).$$

Summing (7.43) with respect to  $K \in \mathcal{K}_n$  and adding the above estimate we find, in view of (7.44), (7.42) and (7.6), that

$$(7.45) \quad W(E_n, \mathbf{1}_{\mathbb{R}^2}) \leq n |\Sigma| \int_{\Sigma \times L_{\text{loc}}^p} W(E) dP(x, E) + Cn \left( \varepsilon + \frac{C_\varepsilon}{\sqrt{n}} \right).$$

Note that at this point if we had chosen  $J_i$  such that  $W(J_i) < \inf_{H_\varepsilon^i} \mathbb{W} + \varepsilon$ , we obtain

$$W(E_n, \mathbf{1}_{\mathbb{R}^2}) \leq n |\Sigma| \int_{\Sigma \times \mathcal{M}_+} \mathbb{W}(\nu) dQ(x, \nu) + Cn \left( \frac{C_\varepsilon}{\sqrt{n}} + \varepsilon \right).$$

*Step 3: Energy bound for  $(x_1, \dots, x_n)$ .* From (7.45), the constructed fields  $\{E_n\}$  and points  $\{\Lambda_n\}_n$  satisfy  $\text{div } E_n = 2\pi \left( \sum_{p \in \Lambda_n} \delta_p - m_0' \right)$  in  $\mathbb{R}^2$  with  $\#\Lambda_n = n$  (cf. item (iii) in Proposition 7.4) and

$$(7.46) \quad \limsup_n \frac{W(E_n, \mathbf{1}_{\mathbb{R}^2})}{n} \leq |\Sigma| \int W(E) dP(x, E) + C\varepsilon.$$

Now let  $\{x_i\}_i = \{p/\sqrt{n}\}_{p \in \Lambda_n}$  be the points in  $\Lambda_n$  in the initial scale, and let still  $\nu_n = \sum_i \delta_{x_i}$ . The next step is to show that modifying  $E_n$  to make it curl-free can only decrease its energy. To see that, defining  $H'_n$  by (1.20), we have that  $-\Delta H'_n = \text{div } E_n$  and we may thus write  $E_n = -\nabla H'_n + \nabla^\perp f_n$  for some function  $f_n$ . But  $E_n = 0$  outside of  $\Sigma$ , by construction, while  $H'_n$  decays fast at infinity by its definition (1.20) and the fact that the right-hand side of (1.20) has integral 0. Letting  $U_\eta = \cup_{p \in \Lambda_n} B(p, \eta)$ , we first have

$$\begin{aligned} \int_{B_R \setminus U_\eta} |\nabla^\perp f_n - \nabla H'_n|^2 - \int_{B_R \setminus U_\eta} |\nabla H'_n|^2 &= -2 \int_{B_R \setminus U_\eta} \nabla^\perp f_n \cdot \nabla H'_n + \int_{B_R \setminus U_\eta} |\nabla f_n|^2 \\ &\geq -2 \int_{B_R \setminus U_\eta} \nabla^\perp f_n \cdot \nabla H'_n. \end{aligned}$$

Since  $E_n \in L_{loc}^q$  for any  $q < 2$  and since  $f_n \in W_{loc}^{1,q}(\mathbb{R}^2)$  for all  $q$ , the last term on the right-hand side converges as  $\eta \rightarrow 0$  to the integral over  $B_R$ . Also integrating by parts, using the Jacobian structure and the decay of  $f_n$  and  $H'_n$ , we have  $\int_{B_R} \nabla^\perp f_n \cdot \nabla H'_n \rightarrow 0$  as  $R \rightarrow +\infty$ . Therefore, letting  $\eta \rightarrow 0$  then  $R \rightarrow +\infty$  in the above yields

$$W(E_n, \mathbf{1}_{\mathbb{R}^2}) - W(\nabla H'_n, \mathbf{1}_{\mathbb{R}^2}) \geq 0.$$

Since  $\Lambda_n \subset \Sigma'$  by construction, we have  $\text{Supp}(\nu_n) \subset \Sigma$  and thus  $\int \zeta d\nu_n = 0$ . Together with (7.46), we deduce in view of (2.1) that

$$(7.47) \quad \limsup_{n \rightarrow \infty} \frac{1}{n} \left( w_n(x_1, \dots, x_n) - n^2 I(\mu_0) + \frac{n}{2} \log n \right) \leq \frac{|\Sigma|}{\pi} \int W(E) dP(x, E) + C\varepsilon.$$

(Respectively  $\leq \frac{|\Sigma|}{\pi} \int \mathbb{W}(\nu) dQ(x, \nu) + C\varepsilon$ .)

*Step 4: Existence of  $A_n$ .* We claim that if  $n$  is large enough and if  $E \in L_{loc}^p(\mathbb{R}^2, \mathbb{R}^2)$  is such that

$$(7.48) \quad d_p(E(\sqrt{n}x + \cdot), E_n^{\text{int}}(\sqrt{n}x + \cdot)) < \eta_1/2$$

for any  $x \in \Sigma \setminus \Xi$  for some set  $\Xi$  satisfying  $|\Xi| < \eta_0 |\Sigma|$ , then

$$(7.49) \quad d_{\mathcal{B}} \left( \int_{\Sigma} \delta_x \otimes \delta_{\theta_{\sqrt{n}x} E} dx, \int_{\Sigma} \delta_x \otimes \delta_{\theta_{\sqrt{n}x} E_n^{\text{int}}} dx \right) < C\varepsilon(|\log \varepsilon| + 1).$$

This would follow immediately from Lemmas 7.1, 7.2 and 7.3 if  $\theta_{\sqrt{n}x} E_n^{\text{int}}$  belonged to some compact set independent of  $x \notin \Xi$  and  $n$ . In our case we note that if  $x$  belongs to some  $\tilde{L} \in \mathcal{L}_{K,i}$ , where  $K \in \mathcal{K}_n$  and  $0 \leq i \leq N_\varepsilon$ , then

$$E_n^{\text{int}}(x + \cdot) = \sigma_{m_K} J_{i,L}(\cdot + x - x_{\tilde{L}}).$$

Moreover, since  $J_i \in H_\varepsilon$ , from (7.14) it follows that if  $x - x_{\tilde{L}} \notin \Gamma(J_i)/\sqrt{m_K}$  then  $E' := \sigma_{m_K} J_i(\cdot + x - x_{\tilde{L}}) \in G_\varepsilon$ . If in addition,  $\text{dist}(x, \partial \tilde{L}) > \eta_0 R_\varepsilon$ , then we deduce from (7.19) that  $d_p(E_n^{\text{int}}(x + \cdot), E') < \eta_1/2$  and  $d_p(E(x + \cdot), E') < \eta_1$ . Lemma 7.2 then yields  $d_{\mathcal{B}}(\delta_{E_n^{\text{int}}(x+\cdot)}, \delta_{E'}) < \varepsilon$  and  $d_{\mathcal{B}}(\delta_{E(x+\cdot)}, \delta_{E'}) < \varepsilon$  thus

$$d_{\mathcal{B}}(\delta_{E_n^{\text{int}}(x+\cdot)}, \delta_{E(x+\cdot)}) < 2\varepsilon.$$

In view of Lemma 7.3 we find

$$d_{\mathcal{B}} \left( \frac{1}{|\Sigma|} \int_{\Sigma \setminus \tilde{\Xi}} \delta_x \otimes \delta_{\theta_{\sqrt{n}x} E} dx, \frac{1}{|\Sigma|} \int_{\Sigma \setminus \tilde{\Xi}} \delta_x \otimes \delta_{\theta_{\sqrt{n}x} E_n^{\text{int}}} dx \right) < C\varepsilon(|\log \varepsilon| + 1),$$

where  $\tilde{\Xi}$  is the union of  $\Xi$  and of the union with respect to  $0 \leq i \leq N_\varepsilon$ ,  $K \in \mathcal{K}_n$  and  $\tilde{L} \in \mathcal{L}_{K,i}$  of  $\frac{1}{\sqrt{n}}(x_{\tilde{L}} + \Gamma(J_i)/\sqrt{m_K})$ , of  $\frac{1}{\sqrt{n}}\{x \in \tilde{L} : \text{dist}(x, \partial \tilde{L}) \leq \eta_0 R_\varepsilon\}$ , and of  $\Sigma \setminus \cup_{\mathcal{K}_n} \frac{K}{\sqrt{n}}$ . It turns out that  $|\tilde{\Xi}| < C\eta_0$  if  $n$  is large enough,  $C$  being of course independent of  $\varepsilon$ , and thus using Lemma 7.1 we deduce (7.49). The claim is proved.

To prove the existence of the set  $A_n$ , we note that the currents  $J_i$  used in constructing  $E_n^{\text{int}}$  depend on  $\varepsilon$  but are independent on  $n$ . Then they are truncated to obtain  $J_{i,K}$  where the sidelengths of  $L$  are in  $[R_\varepsilon/C, CR_\varepsilon]$ , i.e. in an interval independent of  $n$ . It follows at

once that there exists  $\delta > 0$  such that the points in  $L$  may be perturbed by an amount  $\delta$  so that for every  $i$ ,  $K$  and  $\tilde{L} \in \mathcal{L}_{K,i}$  the perturbed  $J_{i,L}^{pert}$  is at a distance at most  $\eta_1/4$  of  $J_{i,K}$ , for every  $n$ . Then in view of (7.44) and (7.37) it follows that for  $n$  large enough the resulting  $E_n^{pert}$  will satisfy (7.48) for  $x$  far enough from  $\partial\Sigma'$ , i.e. outside a set of proportion relative to  $|\Sigma'|$  tending to 0 as  $n \rightarrow \infty$ . We deduce that  $E_n^{pert}$  satisfies (7.49), hence if  $n$  is large enough

$$d_{\mathcal{B}}(P_{E_n^{pert}}, P) < C\varepsilon(|\log \varepsilon| + 1).$$

The same reasoning implies that if we let  $\{x_1, \dots, x_n\}$  be the points in  $\Lambda_n$  in original coordinates, then perturbing the points in  $\Lambda_n$  by an amount  $\delta > 0$  small enough, i.e. perturbing the  $x_i$ 's by an amount  $\delta/\sqrt{n}$  at most we obtain points  $y_i$  such that  $w_n(y_i) \leq w_n(x_i) + \varepsilon$ . Since the ordering of the points is irrelevant, we let  $S_n$  denote the set of permutations of  $1 \dots n$  and define

$$A_n = \{(y_1, \dots, y_n) : \exists \sigma \in S_n, |x_i - y_{\sigma(i)}| < \delta.\}$$

Then, given  $\eta > 0$ , from the previous discussion and choosing  $\varepsilon > 0$  small enough we have for any  $n$  and any  $(y_1, \dots, y_n) \in A_n$  that (4.1) is satisfied and the existence of  $E_n$  such that  $\operatorname{div} E_n = 2\pi (\sum_i \delta_{y_i'} - m_0')$  and such that  $\{P_{E_n}\}_n$  satisfies (4.3).

This concludes the proof of Proposition 4.1.

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