



Universal diagonal estimates for minimizers of the Levy–Lieb functional

S. Di Marino¹ · A. Gerolin^{2,3} · L. Nenna⁴

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Abstract

Given a wave-function minimizing the Levy–Lieb functional, the intent of this short note is to give an estimate of the probability of the particles being in positions (x_1, \dots, x_N) in the δ -close regime $D_\delta = \cup_{i \neq j} \{|x_i - x_j| \leq \delta\}$.

Keywords Multimarginal optimal transportation · Levy-Lieb · Diagonal estimates · DFT

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✉ L. Nenna
luca.nenna@universite-paris-saclay.fr

S. Di Marino
simone.dimarino@unige.it

A. Gerolin
agerolin@uottawa.ca

¹ MaLGA, Università degli Studi di Genova (DIMA), Genoa, Italy

² Department of Mathematics and Statistics, University of Ottawa, Ottawa, Canada

³ Department of Chemistry and Biomolecular Sciences, University of Ottawa, Ottawa, Canada

⁴ Laboratoire de mathématiques d’Orsay, CNRS, Université Paris-Saclay, 91405 Orsay, France

1 Introduction

Density functional theory attempts to describe all the relevant information about a many-body quantum system at ground state in terms of the one electron density ρ . Following Levy and Lieb’s approach [29, 32], the ground state energy can be rephrased as the following variational principle involving only the electron density

$$\mathcal{E}_0[v] = \inf_{\substack{\rho \in \mathcal{A}^N \\ \int_{\mathbb{R}^3} v(x) d\rho < +\infty}} \left\{ F_{LL,\varepsilon}[\rho] + \int_{\mathbb{R}^3} v(x) d\rho \right\},$$

where $\mathcal{A}^N = \{\rho \in L^1(\mathbb{R}^3) : \rho \geq 0, \sqrt{\rho} \in H^1, \rho(\mathbb{R}^3) = N\}$ is the set of admissible densities, v is an external potential and the Levy–Lieb functional $F_{LL,\varepsilon}$ is defined as

$$F_{LL,\varepsilon}[\rho] := \min_{\substack{\psi \in \mathcal{W} \\ \psi \mapsto \rho}} \left\{ \int_{\mathbb{R}^{3N}} \varepsilon |\nabla \psi|^2(x) + v_{ee}(x) |\psi|^2(x) dx \right\}, \tag{1}$$

where $v_{ee}(x_1, \dots, x_N) = \sum_{i < j} \frac{1}{|x_i - x_j|}$ is the Coulomb interaction potential between the N electrons, $\mathcal{W} \subset H^1(\mathbb{R}^{3N}) \cap \{\|\psi\|_{L^2} = 1\}$, with an additional constraint on the symmetry of the wavefunction which we will discuss later, and $\psi \mapsto \rho$ means that the one-body density of ψ is ρ , that is $\rho = N \int_{\mathbb{R}^{3(N-1)}} |\psi|^2$. The Levy–Lieb functional is indeed the lowest possible (kinetic plus interaction) energy of a quantum system having the prescribed density ρ . This universal functional is the central object of density functional theory, since knowing it would allow one to compute the ground state energy of a system with any external potential v . For a complete review on it we refer the reader to [31].

The vector space \mathcal{W} in (1) is the search space of wavefunctions: the natural choice would be to consider $\mathcal{H}^N = \bigwedge_{i=1}^N H^1(\mathbb{R}^3; \mathbb{C})$, that is the fermionic space of anti-symmetric wavefunctions, however we will use $\mathcal{S}^N = \bigotimes_{S,i=1}^N H^1(\mathbb{R}^3; \mathbb{C})$, the bosonic space of *symmetric* wavefunctions, that is

$$\psi(x_{\sigma(1)}, \dots, x_{\sigma(N)}) = \psi(x_1, \dots, x_N), \quad \forall \sigma \in \mathcal{S}$$

Then the vector space \mathcal{W} for the bosonic case can be defined as

$$\mathcal{W} := \{\psi \in \mathcal{S}^N \text{ and } \|\psi\|_{L^2} = 1\}. \tag{2}$$

In fact, although the electrons are *fermions*, also *bosonic* wave-functions can be of interest, and they can be mathematically more treatable: for example we can assume that bosonic minimizers ψ for (1) are positive, which will guarantee that $|\psi|^2$ is a minimizer for (11), which is the functional we will actually treat. Notice that the ground-state energy of *fermionic* systems are generally higher than *bosonic* ones. In our analysis, however, the *bosonic* case is not very restrictive since we are looking at the regime ε small.

Our approach interprets the Levy–Lieb functional as a (Fisher-information regularized) multi-marginal optimal transport problem.

Connection with optimal transportation theory It has been recently shown that the limit functional as $\varepsilon \rightarrow 0$ corresponds to a multi-marginal optimal transport problem [2, 12, 13, 30] (see also the seminal works in the physics and chemistry literature [5, 36–39]): rather than wave-functions, one has now enlarged the constrained search in (1) to minimize among probability measures on \mathbb{R}^{3N} having ρ as marginal, that is

$$F_0[\rho] := \inf_{\mathbb{P} \in \Pi_N(\rho)} \left\{ \int_{\mathbb{R}^{3N}} v_{ee}(x_1, \dots, x_N) d\mathbb{P}(x_1, \dots, x_N) \right\}, \tag{3}$$

where $\Pi_N(\rho)$ denotes the set of probability measures on \mathbb{R}^{3N} having ρ/N as marginals.

The multi-marginal optimal transport with Coulomb cost (3) has garnered attention in the mathematics, physics and chemistry communities and the literature on the subject is growing considerably. Recent developments include results on the existence and non-existence of Monge-type solutions minimizing (3) (e.g., [3, 5, 7, 8, 11, 12, 19, 21]), structural properties of Kantorovich potentials (e.g., [4, 9, 18, 25]), grand-canonical optimal transport [20], efficient computational algorithms (e.g., [1, 14, 17, 23, 26]) and the design of new density functionals (e.g., [6, 24, 28, 34]). The first order expansion around the limit $\varepsilon \rightarrow 0$ of the Levy–Lieb functional was obtained in [10].

We refer to the surveys (and references therein) [18, 22] for a self-contained presentation on multi-marginal optimal transport approach in Density Functional Theory as well as the review article [40] for a the recent developments from a chemistry standpoint.

Main result of this paper In [4, 9, 16] it is shown that the support $\text{supp}(\mathbb{P}^*)$ of a solution \mathbb{P}^* of the limiting problem (3) is uniformly bounded away of the diagonal, i.e. one has always $|x_i - x_j| \geq \delta > 0$ for any $x_i, x_j \in \text{supp}(\mathbb{P}^*)$. In other words, the electrons are always at a certain distance away from each other, which is the expected behaviour since we are in a *classical* framework.

In the sequel we will denote with D_δ the enlarged diagonal

$$D_\delta = \{(x_1, \dots, x_N) \in \mathbb{R}^{3N} : \exists i \neq j \text{ s.t. } |x_i - x_j| \leq \delta\}.$$

In particular the result in [4, 9] can be rephrased saying that the solution to the multi-marginal optimal transport problem is concentrated on the complement of D_δ for some δ . An important feature of the results is that δ depends only on concentration properties of ρ . In fact defining

$$\kappa(\rho, r) := \sup_{x \in \mathbb{R}^3} \rho(B(x, r))/N, \tag{4}$$

the authors in [9] prove that if $\kappa(\rho, \beta) < \frac{1}{2(N-1)}$ then one can choose $\delta = \frac{\beta}{2N}$. Our main result is to extend this property also for $\varepsilon > 0$ small. In particular we do not expect to have $\psi_\varepsilon = 0$ on D_δ but we show that the probability of having the electrons in position $x \in D_\delta$ is very small (5).

Theorem 1.1 (Exponential off-diagonal localization for Coulomb) *Let $\rho \in \mathcal{A}^N$ and let ψ_ε be a minimizer for (1) in the bosonic case, that is $\psi_\varepsilon \in \mathcal{W}$ with \mathcal{W} as defined in (2), where $v_{ee}(x_1, \dots, x_N) = \sum_{i < j} \frac{1}{|x_i - x_j|}$. Let us consider β such that $\kappa(\rho, \beta) \leq \frac{1}{4(N-1)}$ then, let $\alpha \leq \frac{\beta}{32N}$, and suppose $\varepsilon N^2 \ll \alpha$. Then, for $\mathbb{P}_\varepsilon(x) = |\psi_\varepsilon|^2(x)$ we have*

$$\int_{D_{\alpha/2}} \mathbb{P}_\varepsilon(x) dx \leq e^{-\frac{1}{24} \sqrt{\frac{\alpha}{\varepsilon}}}. \tag{5}$$

In the proof we actually work with a general repulsive pairwise potential v_{ee} , which satisfies the hypothesis (7), stated in the next section. The result in general is the following one:

Theorem 1.2 (Exponential off-diagonal localization for general interaction cost) *Let $\rho \in \mathcal{A}_N$ and let ψ_ε be a minimizer for (1) in the bosonic case where v_{ee} satisfies (7) for some $\theta, \Theta : (0, \infty) \rightarrow [0, \infty)$ decreasing such that $\lim_{t \rightarrow 0^+} \theta(t) = +\infty$. Let β be such that $\kappa(\rho, \beta) \leq \frac{1}{4(N-1)}$ and let α such that $\theta(2\alpha) \leq 8(N-1)\Theta(\beta/2)$, and suppose $\varepsilon N^2 \ll \alpha^2 \theta(2\alpha)$. Then, for $\mathbb{P}_\varepsilon(x) = |\psi_\varepsilon|^2(x)$ we have*

$$\int_{D_{\alpha/2}} \mathbb{P}_\varepsilon(x) dx \leq e^{-\frac{1}{12} \sqrt{\frac{\alpha^2 \theta(2\alpha)}{2\varepsilon}}}. \tag{6}$$

Notice that in [9] the diagonal estimate is proven also in the weaker (and sharper) hypothesis $\kappa(\rho, \beta) < \frac{1}{N}$: while we believe that also in that case a similar generalization in the case $\varepsilon > 0$ holds true, the proof will be more technical and not so transparent. For the same reason the inequality $\kappa(\rho, \beta) \leq \frac{1}{4(N-1)}$ is used instead of $\kappa(\rho, \beta) < \frac{1}{2(N-1)}$ in order to have more transparent estimates in the end.

Organization of the paper In Sect. 2 we introduce the notations we are going to use throughout all the paper. In Sect. 3 we give the estimates concerning kinetic energy term in the Levy–Lieb functional. Section 4 is then devoted to the construction of a competitor for the Levy–Lieb functional; finally in Sect. 5 we derive the diagonal estimates for the wave-function and, thus, prove Theorems 1.1 and 1.2 via the iteration of a decay estimate.

2 Notation

Consider a subset $I \subseteq \{1, \dots, N\}$, with cardinality $k = |I|$, defined as $I = \{i_1, \dots, i_k\}$, with $1 \leq i_1 < i_2 < \dots < i_k$. Then, the I -projection is defined by

$$\pi_I : \mathbb{R}^{3N} \rightarrow \mathbb{R}^{3k}, \quad \pi_I((x_1, \dots, x_N)) = (x_{i_1}, x_{i_2}, \dots, x_{i_k}).$$

Sometimes we will denote $x_I = \pi_I(x)$ and if $I = J^c$, then $x = (x_I, x_J)$. With a slight abuse of notation, for a function $\mathbb{P} \in L^1(\mathbb{R}^{3N})$, $I \subseteq \{1, \dots, N\}$ and $J = I^c$ we denote

$$(\pi_I)_\#(\mathbb{P})(x_I) = \int \mathbb{P}(x_I, x_J) dx_J,$$

which on density of measures act precisely as the push-forward through the projection function π_I .

As we have already mentioned above, we denote by $\Pi_N(\rho)$ the set of probability measures on \mathbb{R}^{3N} having the N one body marginals equal to ρ/N .

In the following we will consider an electron–electron pair interaction repulsion potential, v_{ee} , with the following form:

$$v_{ee}(x_1, \dots, x_N) = \sum_{i < j} c(x_i, x_j), \quad \text{where} \tag{7}$$

$$\theta(|x - y|) \leq c(x, y) \leq \Theta(|x - y|) \quad \forall x, y \in \mathbb{R}^3$$

for some $\theta, \Theta : (0, \infty) \rightarrow [0, \infty)$ decreasing such that $\lim_{t \rightarrow 0^+} \theta(t) = +\infty$.

Moreover, with a slight abuse of notation, we will denote by

$$\mathbb{P} \in \mathcal{P}(\mathbb{R}^{3N}) \mapsto v_{ee}(\mathbb{P}) := \int_{\mathbb{R}^{3N}} v_{ee}(x_1, \dots, x_N) d\mathbb{P}(x_1, \dots, x_N). \tag{8}$$

Notice that we will often identify a measure \mathbb{P} with its density.

Finally, given an open set $\Omega \subseteq \mathbb{R}^{3N}$, for every $r > 0$ we denote with Ω_{-r} its r -thinning, that is the set of points inside Ω whose distance from $\partial\Omega$ is greater or equal than r . In particular

$$\Omega_{-r} := \{x \in \mathbb{R}^{3N} : B(x, r) \subseteq \Omega\}. \tag{9}$$

3 Estimate for the kinetic energy

In this section we give some preliminary estimates for the kinetic energy term of the Levy–Lieb functional. Denoting L^1_+ the cone of positive L^1 functions, we define $\mathcal{E}_{\text{kin}} : L^1_+(\mathbb{R}^{3N}) \rightarrow \mathbb{R}$ the Kinetic energy associated to some absolutely continuous N -probability measure h

$$\mathcal{E}_{\text{kin}}(h) := \begin{cases} \int_{\mathbb{R}^{3N}} \frac{\sum_{i=1}^N |\nabla_i h(x_1, \dots, x_N)|^2}{h(x_1, \dots, x_N)} dx_1, \dots, dx_N & \text{if } \sqrt{h} \in H^1(\mathbb{R}^{3N}) \\ +\infty & \text{otherwise} \end{cases} \tag{10}$$

When it will be clear from the context we will also abbreviate $\mathcal{E}_{\text{kin}}(h) = \int \frac{|\nabla h|^2}{h} dx$. Notice that the kinetic energy functional is also known as the Fisher information.

Moreover if $\psi \in H^1(\mathbb{R}^{3N}; \mathbb{R})$, then

$$4 \int |\nabla \psi|^2 dx = \mathcal{E}_{\text{kin}}(|\psi|^2) = \mathcal{E}_{\text{kin}}(\mathbb{P}_\psi),$$

where $\mathbb{P}_\psi = |\psi|^2$ is the joint probability associated to the wave-function ψ . The string of equalities above is thus true when ψ is a minimizer for the bosonic case. The following Lemma summarises some results concerning the homogeneity, sub-additivity (which is a consequence of theorem 7.8 in [33]) and the decomposability under projection of the kinetic energy (a similar result also appears in [27, 35]).

Lemma 3.1 *Let \mathcal{E}_{kin} defined as in (10). Then*

- (i) \mathcal{E}_{kin} is 1-homogeneous, that is $\mathcal{E}_{\text{kin}}(\lambda\mathbb{P}) = \lambda\mathcal{E}_{\text{kin}}(\mathbb{P})$ for every $\lambda > 0$;
- (ii) given $\mathbb{P}_1, \dots, \mathbb{P}_k \in L^1(\mathbb{R}^{3N})$, we have

$$\mathcal{E}_{\text{kin}}(\mathbb{P}_1 + \dots + \mathbb{P}_k) \leq \mathcal{E}_{\text{kin}}(\mathbb{P}_1) + \mathcal{E}_{\text{kin}}(\mathbb{P}_2) + \dots + \mathcal{E}_{\text{kin}}(\mathbb{P}_k);$$

- (iii) Let $\mathbb{P} \in L^1_+(\mathbb{R}^{3N})$. Given $I, J \subseteq \{1, \dots, N\}$ two nonempty disjoint sets such that $I = J^c$, we denote by $\mathbb{P}_I = (\pi_I)_\# \mathbb{P}$ and $\mathbb{P}_J = (\pi_J)_\# \mathbb{P}$. Then we have (here $N_I = \#I$ and $N_J = \#J$)

$$\mathcal{E}_{\text{kin}}^N(\mathbb{P}) \geq \mathcal{E}_{\text{kin}}^{N_I}(\mathbb{P}_I) + \mathcal{E}_{\text{kin}}^{N_J}(\mathbb{P}_J),$$

where the equality holds if and only if $\mathbb{P}(x) = \mathbb{P}_I(x_I)\mathbb{P}_J(x_J)/\lambda$, where $\lambda = \int \mathbb{P}$. In particular if \mathbb{P} is the density of a probability measure, we have that the equality happens if and only if x_I and x_J are independent under the probability \mathbb{P} .

Proof (i) The 1-homogeneity is obvious.

(ii) For the subadditivity it is sufficient to prove it for $k = 2$; then for every x , by Cauchy–Schwarz inequality we have

$$(\mathbb{P}_1(x) + \mathbb{P}_2(x)) \left(\frac{|\nabla \mathbb{P}_1(x)|^2}{\mathbb{P}_1(x)} + \frac{|\nabla \mathbb{P}_2(x)|^2}{\mathbb{P}_2(x)} \right) \geq (|\nabla \mathbb{P}_1(x)| + |\nabla \mathbb{P}_2(x)|)^2,$$

which, after using the triangular inequality and dividing by $\mathbb{P}_1 + \mathbb{P}_2$ can be rewritten as

$$\frac{|\nabla(\mathbb{P}_1 + \mathbb{P}_2)|^2}{\mathbb{P}_1 + \mathbb{P}_2} \leq \frac{|\nabla \mathbb{P}_1|^2}{\mathbb{P}_1} + \frac{|\nabla \mathbb{P}_2|^2}{\mathbb{P}_2},$$

which integrated gives us the conclusion.

(iii) As for the last point we fix x_J and we use the Cauchy–Schwarz inequality with respect to the measure dx_I :

$$\left(\int \mathbb{P}(x_I, x_J) dx_I \right) \cdot \left(\int \frac{|\nabla_J \mathbb{P}(x_I, x_J)|^2}{\mathbb{P}(x_I, x_J)} dx_I \right) \geq \left(\int |\nabla_J \mathbb{P}(x_I, x_J)| dx_I \right)^2$$

$$\geq \left| \nabla_J \left(\int \mathbb{P}(x_I, x_J) dx_I \right) \right|^2,$$

where in the last passage we used the triangular inequality and we took the derivative out of the integral. Now we recognize $\mathbb{P}_J(x_J) = \int \mathbb{P}(x_I, x_J) dx_I$ and so we can write this as

$$\int \frac{|\nabla_J \mathbb{P}(x_I, x_J)|^2}{\mathbb{P}(x_I, x_J)} dx_I \geq \frac{|\nabla_J \mathbb{P}_J(x_J)|^2}{\mathbb{P}_J(x_J)}.$$

Integrating this with respect to dx_J and doing a similar computation for x_I , we obtain the conclusion, that is

$$\iint \frac{|\nabla \mathbb{P}(x_I, x_J)|^2}{\mathbb{P}(x_I, x_J)} dx_I dx_J \geq \int \frac{|\nabla_J \mathbb{P}_J(x_J)|^2}{\mathbb{P}_J(x_J)} dx_J + \int \frac{|\nabla_I \mathbb{P}_I(x_I)|^2}{\mathbb{P}_I(x_I)} dx_I.$$

From the equality cases in C–S and triangular inequality combined we get $\nabla_J \mathbb{P}(x_I, x_J) = v(x_J) \mathbb{P}(x_I, x_J)$ for some vector field v ; by a simple integration we actually get $v = \nabla(\mathbb{P}_J)/\mathbb{P}_J$; this can be seen as $\nabla_J \log(\mathbb{P}) = \nabla_J \log \mathbb{P}_J$; similarly we can get $\nabla_I \log(\mathbb{P}) = \nabla_I \log \mathbb{P}_I$. Summing up this two equalities we get $\nabla(\mathbb{P}(x)/\mathbb{P}_I(x_I)\mathbb{P}_J(x_J)) = 0$. \square

The following lemma is a straightforward adaptation of Theorem 3.2 in [15] giving the IMS localization formula; we have added a short proof for sake of completeness.

Lemma 3.2 *Let $\eta_1, \eta_2, \eta_3 : \mathbb{R}^{3N} \rightarrow [0, 1]$ be C^1 functions such that $\eta_1 + \eta_2 + \eta_3 \equiv 1$. Then, for every function $\mathbb{P} \in L^1_+(\mathbb{R}^{3N})$ we have*

$$\begin{aligned} & \mathcal{E}_{\text{kin}}(\mathbb{P}\eta_1) + \mathcal{E}_{\text{kin}}(\mathbb{P}\eta_2) + \mathcal{E}_{\text{kin}}(\mathbb{P}\eta_3) \\ &= \mathcal{E}_{\text{kin}}(\mathbb{P}) + \int \left(\frac{|\nabla \eta_1|^2}{\eta_1} + \frac{|\nabla \eta_2|^2}{\eta_2} + \frac{|\nabla \eta_3|^2}{\eta_3} \right) \mathbb{P} dx. \end{aligned}$$

Proof For every $i = 1, 2, 3$ pointwisely we have:

$$\begin{aligned} \frac{|\nabla(\mathbb{P}\eta_i)|^2}{\mathbb{P}\eta_i} &= \frac{|\eta_i \nabla \mathbb{P} + \mathbb{P} \nabla \eta_i|^2}{\mathbb{P}\eta_i} = \frac{\eta_i^2 |\nabla \mathbb{P}|^2 + 2\eta_i \mathbb{P} \nabla \mathbb{P} \cdot \nabla \eta_i + \mathbb{P}^2 |\nabla \eta_i|^2}{\mathbb{P}\eta_i} \\ &= \eta_i \frac{|\nabla \mathbb{P}|^2}{\mathbb{P}} + 2 \nabla \mathbb{P} \cdot \nabla \eta_i + \mathbb{P} \frac{|\nabla \eta_i|^2}{\eta_i} \end{aligned}$$

Adding them up and using that $\sum \eta_i = 1$ and $\sum \nabla \eta_i = 0$, we get

$$\sum_i \frac{|\nabla(\mathbb{P}\eta_i)|^2}{\mathbb{P}\eta_i} = \frac{|\nabla \mathbb{P}|^2}{\mathbb{P}} + \mathbb{P} \sum_i \frac{|\nabla \eta_i|^2}{\eta_i},$$

which integrated, gives us the desired identity. \square

4 New trial state: swapping particles and estimate for the potential

The scope of this subsection is to create a competitor for the minimization of the functional

$$\mathcal{F}_{LL,\varepsilon}(\mathbb{P}) := \begin{cases} \frac{\varepsilon}{4} \mathcal{E}_{\text{kin}}(\mathbb{P}) + v_{ee}(\mathbb{P}) & \text{if } \mathbb{P} \in \Pi_N(\rho) \\ +\infty & \text{otherwise,} \end{cases} \tag{11}$$

where \mathcal{E}_{kin} is defined in (10), v_{ee} satisfies (7) and $\Pi_N(\rho)$ denotes the set of probability measures on \mathbb{R}^{3N} having ρ/N as marginals. The idea is to try to mimic what it is done in [4, 9, 16], in the semiclassical case $\varepsilon = 0$: in that case we take two points $y, z \in \mathbb{R}^{3N}$ and substitute them with \tilde{y}, \tilde{z} where we have interchanged their first component, that is $\tilde{y} = (z_1, y_2, \dots, y_n)$ and $\tilde{z} = (y_1, z_2, \dots, z_n)$.

In order to do so for the N -particle distribution \mathbb{P} , we will consider two small bumps centered around y and z

$$\eta_1(x) = \lambda_1 \eta\left(\frac{x - y}{r_1}\right) \quad \text{and} \quad \eta_2(x) = \lambda_2 \eta\left(\frac{x - z}{r_2}\right), \tag{12}$$

for some $\lambda_1, \lambda_2, r_1, r_2$ to be chosen later and some $\eta \in C_c^1(B(0, 1))$, $\eta \geq 0$. First of all we assume that $\text{supp}(\eta_1) \cap \text{supp}(\eta_2) = \emptyset$, which can be granted as long as

$$r_1 + r_2 < |y - z|, \tag{13}$$

and then we assume $\int \eta_1 \mathbb{P} = \int \eta_2 \mathbb{P} = m$ which can be accomplished again by choosing the appropriate λ_i, r_i . Let us then define

$$\rho_1^i(x_1) = (\pi_{\{1\}})_\#(\eta_i \mathbb{P}), \quad \rho_1^i(x_2, x_3, \dots, x_N) = (\pi_{\{1\}^c})_\#(\eta_i \mathbb{P}), \tag{14}$$

$$\mathbb{P}_1 = \frac{1}{m} \rho_1^2 \rho_1^1, \quad \mathbb{P}_2 = \frac{1}{m} \rho_1^1 \rho_1^2, \tag{15}$$

where ρ_1^i and ρ_1^i are the marginals of $\eta_i \mathbb{P}$ and \mathbb{P}_1 and \mathbb{P}_2 are densities concentrated around $\tilde{y} = (z_1, y_2, \dots, y_n)$ and $\tilde{z} = (y_1, z_2, \dots, z_n)$ respectively. We then finally consider

$$\bar{\mathbb{P}} := \mathbb{P} - \mathbb{P}\eta_1 - \mathbb{P}\eta_2 + \mathbb{P}_1 + \mathbb{P}_2, \tag{16}$$

which will be the competitor for a minimizer \mathbb{P} of the functional $\mathcal{F}_{LL,\varepsilon}$.

Remark 4.1 Given y, z, r_1, r_2 that satisfy (13), the only condition that remains to be checked is whether $\bar{\mathbb{P}}$ is a competitor: we will prove that this is the case for every λ_1 and λ_2 small enough.

In fact we have to check that $\bar{\mathbb{P}} \geq 0$ and that it has the correct marginals. For the positivity, notice that for λ_1 and λ_2 small enough, we have $\eta_1 + \eta_2 \leq 1$ and so $\mathbb{P} - \eta_1 \mathbb{P} - \eta_2 \mathbb{P} \geq 0$, which will guarantee that $\bar{\mathbb{P}} \geq 0$.

For the marginal constraint, notice that by (14) and (15) we have that $\eta_1 \mathbb{P} + \eta_2 \mathbb{P}$ and $\mathbb{P}_1 + \mathbb{P}_2$ have the same marginals, in particular also \mathbb{P} and $\bar{\mathbb{P}}$ share the same marginals.

Lemma 4.1 *Let \mathbb{P} be such $\mathcal{F}_{LL,\varepsilon}(\mathbb{P}) < +\infty$. Given $y, z \in \mathbb{R}^{3N}$, let $\eta_1, \eta_2, \mathbb{P}_1, \mathbb{P}_2, \bar{\mathbb{P}}$ defined by (12), (13), (14), (15) and (16). Then*

$$\begin{aligned} \mathcal{E}_{\text{kin}}(\bar{\mathbb{P}}) &\leq \mathcal{E}_{\text{kin}}(\mathbb{P}) + \int \mathbb{P}(x) \left(\frac{|\nabla\eta_1|^2}{\eta_1} + \frac{|\nabla\eta_2|^2}{\eta_2} + \frac{|\nabla\eta_1 + \nabla\eta_2|^2}{1 - \eta_1 - \eta_2} \right) dx; \\ v_{ee}(\bar{\mathbb{P}}) &= v_{ee}(\mathbb{P}) - \int \mathbb{P}(\eta_1 + \eta_2) \sum_{i>1} c(x_1, x_i) dx + \int (\mathbb{P}_1 + \mathbb{P}_2) \sum_{i>1} c(x_1, x_i) dx. \end{aligned}$$

Proof Let us consider $\eta_3 = 1 - \eta_1 - \eta_2$. Then we have $\bar{\mathbb{P}} = \eta_3\mathbb{P} + \mathbb{P}_1 + \mathbb{P}_2$. Using Lemma 3.1, in particular the subadditivity and the exact energy split in case of independent variables for \mathcal{E}_{kin} , we get (by (15))

$$\begin{aligned} \mathcal{E}_{\text{kin}}(\bar{\mathbb{P}}) &\leq \mathcal{E}_{\text{kin}}(\eta_3\mathbb{P}) + \mathcal{E}_{\text{kin}}(\mathbb{P}_1) + \mathcal{E}_{\text{kin}}(\mathbb{P}_2) \\ &= \mathcal{E}_{\text{kin}}(\eta_3\mathbb{P}) + \mathcal{E}_{\text{kin}}(\rho_1^2) + \mathcal{E}_{\text{kin}}(\rho_1^1) + \mathcal{E}_{\text{kin}}(\rho_1^1) + \mathcal{E}_{\text{kin}}(\rho_1^2); \end{aligned} \tag{17}$$

we then recall (14) and the inequality for the split energy (Lemma 3.1 (iii)) to get

$$\mathcal{E}_{\text{kin}}(\rho_1^i) + \mathcal{E}_{\text{kin}}(\rho_1^i) \leq \mathcal{E}_{\text{kin}}(\eta_i\mathbb{P}) \tag{18}$$

and so we conclude using (17), (18) and then Lemma 3.2.

For the estimate with the potential, it is clear that

$$v_{ee}(\bar{\mathbb{P}}) = v_{ee}(\mathbb{P}) - \int \mathbb{P}(\eta_1 + \eta_2)v_{ee}(x) dx + \int (\mathbb{P}_1 + \mathbb{P}_2)v_{ee}(x) dx;$$

Since $v_{ee}(x) = \sum_{i<j} c(x_i, x_j)$ we just need to show that the contribution due to $c(x_i, x_j)$ whenever $1 < i < j$ cancels out in the last two integrals. In fact in both integrals we can integrate out the first variable: denoting $I = \{1\}$ and $J = I^c$ for example we have

$$\begin{aligned} \int \mathbb{P}\eta_1 c(x_i, x_j) dx_I dx_J &= \int c(x_i, x_j) \left(\int \mathbb{P}\eta_1 dx_I \right) dx_J \\ &= \int c(x_i, x_j) \rho_1^1(x_J) dx_J \\ &= \int c(x_i, x_j) \rho_1^1(x_J) \left(\int \frac{\rho_2^1(x_I)}{m} dx_I \right) dx_J \\ &= \int c(x_i, x_j) \mathbb{P}_1 dx. \end{aligned}$$

In a similar way we can show that $\int \mathbb{P}\eta_2 c(x_i, x_j) dx = \int \mathbb{P}_2 c(x_i, x_j) dx$. Now by definition of \mathbb{P}_1 and \mathbb{P}_2 , this implies that

$$\begin{aligned}
 & - \int \mathbb{P}(\eta_1 + \eta_2) \left(\sum_{1 < i < j} c(x_i, x_j) \right) dx + \int (\mathbb{P}_1 + \mathbb{P}_2) \left(\sum_{1 < i < j} c(x_i, x_j) \right) dx \\
 & = - \int (\mathbb{P}_1 + \mathbb{P}_2) \left(\sum_{1 < i < j} c(x_i, x_j) \right) dx + \int (\mathbb{P}_1 + \mathbb{P}_2) \left(\sum_{1 < i < j} c(x_i, x_j) \right) dx = 0,
 \end{aligned}$$

which yields to the desired result. □

5 Diagonal estimates for the wave-function

We devote this last section to derive the diagonal estimates for the bosonic wave-function which minimizes the Levy–Lieb functional proving in particular Theorems 1.1 and 1.2. In the sequel we will denote $C_1(x) = \sum_{i=2}^N c(x_1, x_i)$

Lemma 5.1 *Let ρ be an one body marginal distribution with $\rho(\mathbb{R}^3) = N$ and let $\beta > 0$ be such that $\kappa(\rho, \beta) \leq \frac{1}{4(N-1)}$, where κ is defined as in (4). Then, for every $\mathbb{P} \in \Pi_N(\rho)$ and $y \in \mathbb{R}^{3N}$, for every r_1, r_2 such that $r_1 + 2r_2 < \beta$ and $\delta > 0$, there exists $z \in \mathbb{R}^{3N}$ such that, defining $\eta_1, \eta_2, \mathbb{P}_1, \mathbb{P}_2, m$ as in (12), (14) and (15)*

- (i) $\int C_1(x)(\mathbb{P}_1 + \mathbb{P}_2)dx \leq 2(N - 1)\Theta(\beta - r_1 - 2r_2)m$;
- (ii) z is a $(1 + \delta, 1/2)$ -doubling point at scale r_2 for \mathbb{P} , that is

$$\int_{B(z, r_2)} \mathbb{P} dx \leq 2(1 + \delta)^{3N} \int_{B(z, r_2/(1+\delta))} \mathbb{P} dx.$$

Proof For $\gamma > 0$, let us consider the set

$$\Omega = \left\{ z \in \mathbb{R}^{3N} : |z_1 - y_i| \geq \gamma - r_2 \text{ and } |y_1 - z_i| \geq \gamma - r_2, \forall i = 2, \dots, N \right\}.$$

We know that if $z \in \Omega$ we will have of course

$$\begin{aligned}
 & C_1(y'_1, z'_2, \dots, z'_N) + C_1(z'_1, y'_2, \dots, y'_N) \\
 & \leq 2(N - 1)\Theta(\gamma - r_1 - 2r_2) \forall y' \in B(y, r_1), z' \in B(z, r_2),
 \end{aligned}$$

which in particular implies $\int C_1(x)(\mathbb{P}_1 + \mathbb{P}_2) dx \leq 2(N - 1)\Theta(\gamma - r_1 - 2r_2)m$. Now we want to see that there exists a $1/2$ doubling point in Ω ; in order to do that, it is easy to see that

$$\Omega_{-r_2} \subseteq \{y' \in \mathbb{R}^{3N} : |y'_1 - y_i| \geq \gamma \text{ and } |y_1 - y'_i| \geq \gamma, \forall i = 2, \dots, N\}$$

And now a similar computation to what is done in [4, Lemma 2.3] and in [9, proof of Theorem 1.3 (i)] will give us

$$\int_{\Omega_{-r_2}} \mathbb{P}(x) \, dx \geq 1 - 2(N - 1)\kappa(\rho, \gamma).$$

Now if we consider $\gamma = \beta$ we have $\kappa(\rho, \beta) \leq \frac{1}{4(N-1)}$, and so we can apply Lemma 5.2 with $r = \frac{r_2}{1+\delta}$ get the existence of a $(1 + \delta, 1/2)$ -doubling point at scale r_2 in Ω . \square

Lemma 5.2 (Existence of doubling points) *Let $\mathbb{P} \in L^1_+(\mathbb{R}^{3N})$ be the density of a probability measure and let $r > 0$. Let us consider an open set $\Omega \subseteq \mathbb{R}^{3N}$; we denote $M_r := \int_{\Omega_{-r}} \mathbb{P}(x) \, dx$, where Ω_{-r} is the r -thinning of the set Ω , defined as in (9). Then, whenever $M_r > 0$, for every $\delta > 0$ there exists $y \in \Omega$, such that*

$$\int_{B(y, (1+\delta)r)} \mathbb{P}(x) \, dx \leq \frac{(1 + \delta)^{3N}}{M_r} \int_{B(y, r)} \mathbb{P}(x) \, dx,$$

that is, the measure $\mathbb{P}(x) \, dx$ is doubling at the point y at scale r , with doubling constant $\frac{(1+\delta)^{3N}}{M_r}$.

Proof Suppose on the contrary that for every $y \in \Omega$ the reversed inequality holds

$$\int_{B(y, (1+\delta)r)} \mathbb{P}(x) \, dx > \frac{(1 + \delta)^{3N}}{M_r} \int_{B(y, r)} \mathbb{P}(x) \, dx.$$

Then we can integrate this inequality on the whole Ω

$$\int_{\Omega} \int_{B(y, (1+\delta)r)} \mathbb{P}(x) \, dx \, dy > \frac{(1 + \delta)^{3N}}{M_r} \int_{\Omega} \int_{B(y, r)} \mathbb{P}(x) \, dx \, dy.$$

Let ω_{3N} be the volume of the unit ball in \mathbb{R}^{3N} . Using Fubini we get

$$\begin{aligned} \omega_{3N} \cdot ((1 + \delta)r)^{3N} &= \int_{\mathbb{R}^{3N}} \int_{B(y, (1+\delta)r)} \mathbb{P}(x) \, dx \, dy \geq \int_{\Omega} \int_{B(y, (1+\delta)r)} \mathbb{P}(x) \, dx \, dy \\ M_r \omega_{3N} \cdot r^{3N} &= \int_{\Omega_{-r}} \mathbb{P}(x) |B(x, r)| \, dz = \int_{\Omega_{-r}} \int_{|x-y| < r} \mathbb{P}(x) \, dy \, dx \\ &= \int_{\Omega} \int_{B(y, r) \cap \Omega_{-r}} \mathbb{P}(x) \, dx \, dy \leq \int_{\Omega} \int_{B(y, r)} \mathbb{P}(x) \, dx \, dy, \end{aligned}$$

where we crucially used that if $y \in B(x, r)$ and $x \in \Omega_{-r}$ then $y \in \Omega$. We thus reached a contradiction. \square

Proposition 5.1 (One step decay) *Let us consider ρ and β such that $\kappa(\rho, \beta) \leq \frac{1}{4(N-1)}$. Then there exists $\alpha_0 = \alpha(\beta, \varepsilon)$ such that if \mathbb{P} minimizes (11), we have that for every*

$y \in D_\alpha$ such that $\alpha \leq \alpha_0$, and every $\tilde{r} \leq \alpha/2$, we have

$$\int_{B(y, \tilde{r}/(1+\delta))} \mathbb{P}(x) \, dx \leq \frac{1}{\delta^2 \tilde{r}^2 \frac{\theta(2\alpha)}{2(1+\delta)^2 \varepsilon} + 1} \int_{B(y, \tilde{r})} \mathbb{P}(x) \, dx, \tag{19}$$

whenever $\delta > 0$ is such that $\theta(2\alpha) > 256\varepsilon C(\delta)/\beta^2$, where

$$C(\delta) := \frac{(1 + \delta)^2 \cdot (2(1 + \delta)^{3N} - 1)}{\delta^2}. \tag{20}$$

An implicit choice for α_0 is for example $\theta(2\alpha_0) > 8 \max\{(N - 1)\Theta(\beta/2), 832\varepsilon N^2/\beta^2\}$.

Proof Let $y \in D_\alpha$ and without loss of generality we can assume that $|y_1 - y_2| < \alpha$; let z given by Lemma 5.1. We then consider $r_1, r_2, \eta_1, \eta_2, \lambda_1, \lambda_2, \mathbb{P}_1, \mathbb{P}_2, \bar{\mathbb{P}}$ defined by (12), (13), (14), (15) and (16); being $\bar{\mathbb{P}} \in \Pi_N(\rho)$, we get, by the minimality of \mathbb{P} ,

$$\begin{aligned} \mathcal{F}_{LL, \varepsilon}(\bar{\mathbb{P}}) &\geq \mathcal{F}_{LL, \varepsilon}(\mathbb{P}), \\ \frac{\varepsilon}{4} \mathcal{E}_{\text{kin}}(\bar{\mathbb{P}}) + v_{ee}(\bar{\mathbb{P}}) &\geq \frac{\varepsilon}{4} \mathcal{E}_{\text{kin}}(\mathbb{P}) + v_{ee}(\mathbb{P}); \end{aligned}$$

now we can use the estimates in Lemma 4.1 in order to conclude that

$$\begin{aligned} &\frac{\varepsilon}{4} \int \mathbb{P} \left(\frac{|\nabla \eta_1|^2}{\eta_1} + \frac{|\nabla \eta_2|^2}{\eta_2} + \frac{|\nabla \eta_1 + \nabla \eta_2|^2}{1 - \eta_1 - \eta_2} \right) dx \\ &\geq \int \mathbb{P}(\eta_1 + \eta_2) C_1(x) dx - \int (\mathbb{P}_1 + \mathbb{P}_2) C_1(x) dx. \end{aligned}$$

Now we make the choice $\eta(x) = \min \left\{ \frac{(1+\delta)(1-|x|)_+}{\delta}, 1 \right\}$. In particular $0 \leq \eta \leq 1$ and $\eta \equiv 1$ if $|x| < \frac{1}{1+\delta}$, and moreover $\frac{|\nabla \eta|^2}{\eta} = 4|\nabla \sqrt{\eta}|^2 \equiv 4 \left(\frac{1+\delta}{\delta}\right)^2$ if $\frac{1}{1+\delta} \leq |x| \leq 1$ and 0 otherwise. Notice that η_1 and η_2 are centred in y and z respectively, we thus have

$$\begin{aligned} \frac{1}{4} \int \mathbb{P} \frac{|\nabla \eta_1|^2}{\eta_1} &= \frac{(1 + \delta)^2}{\delta^2 r_1^2} \int_{B(y, r_1) \setminus B(y, r_1/(1+\delta))} \mathbb{P} \lambda_1 \, dx \\ &= \frac{(1 + \delta)^2}{\delta^2 r_1^2} \left(\int_{B(y, r_1)} \mathbb{P} \lambda_1 \, dx - \int_{B(y, r_1/(1+\delta))} \mathbb{P} \lambda_1 \, dx \right) \end{aligned}$$

In a similar way we have

$$\begin{aligned} \frac{1}{4} \int \mathbb{P} \frac{|\nabla \eta_2|^2}{\eta_2} &= \frac{(1 + \delta)^2}{\delta^2 r_2^2} \left(\int_{B(z, r_2)} \mathbb{P} \lambda_2 \, dx - \int_{B(z, r_2/(1+\delta))} \mathbb{P} \lambda_2 \, dx \right) \\ &\leq \frac{(1 + \delta)^2 \cdot (2(1 + \delta)^{3N} - 1)}{\delta^2 r_2^2} \cdot \int_{B(z, r_2/(1+\delta))} \mathbb{P} \lambda_2 \, dx \leq \frac{C(\delta)}{r_2^2} \cdot m. \end{aligned}$$

where in the last steps we used Lemma 5.1 (ii) and the definition of $C(\delta)$ (20). Notice then that in the regime $\lambda_1, \lambda_2 \ll 1$ [we remind that λ_1, λ_2 are two parameters in the definition of the bumps η_1 and η_2 , see (12)] we have that the last contribution for the localization error $\int \frac{|\nabla\eta_1 + \nabla\eta_2|^2}{1 - \eta_1 - \eta_2} \mathbb{P}$ is of order $O(\lambda_1^2)$.

Now we use that $\int C_1(x)\mathbb{P}\eta_1 dx \geq \theta(\alpha + 2r_1) \cdot m$, the nonnegativity of C_1 (notice that we do not have any other information on C_1 on the support of η_2) and the estimates we have for $\int C_1(x)(\mathbb{P}_1 + \mathbb{P}_2) dx$ to get

$$\begin{aligned} & \int \mathbb{P}(\eta_1 + \eta_2)C_1(x)dx - \int (\mathbb{P}_1 + \mathbb{P}_2)C_1(x)dx \\ & \geq [\theta(\alpha + 2r_1) - 2(N - 1)\Theta(\beta - r_1 - 2r_2)] \cdot m. \end{aligned}$$

Putting everything together we have

$$\begin{aligned} \varepsilon \frac{(1 + \delta)^2}{\delta^2 r_1^2} \int_{B(y, r_1)} \mathbb{P}\lambda_1 dx & \geq \left[\theta(\alpha + 2r_1) - 2(N - 1)\Theta(\beta - r_1 - 2r_2) - \frac{\varepsilon C(\delta)}{r_2^2} \right] \cdot m \\ & + \varepsilon \frac{(1 + \delta)^2}{\delta^2 r_1^2} \int_{B(y, r_1/(1+\delta))} \mathbb{P}\lambda_1 dx - O(\lambda_1^2) \end{aligned} \tag{21}$$

Define

$$F(r_1, \varepsilon, \alpha) := \max \left\{ \theta(\alpha + 2r_1) - 2(N - 1)\Theta(\beta - r_1 - 2r_2) - \frac{4\varepsilon C(\delta)}{r_2^2} : r_2 > 0 \right\}.$$

We can take $r_1 = \tilde{r} \leq \alpha/2 \leq \beta/4$ and $r_2 = \beta/8$, and then choose $\alpha < \alpha_0$ such that

$$\frac{\theta(2\alpha)}{2} - 2(N - 1)\Theta(\beta/2) > \frac{\theta(2\alpha)}{4} \quad \text{and} \quad \frac{\theta(2\alpha)}{2} - \frac{\varepsilon C(\delta)}{r_2^2} > \frac{\theta(2\alpha)}{4}, \tag{22}$$

obtaining $F(r_1, \varepsilon, \alpha) \geq \theta(2\alpha)/2$.

We can now use $m \geq \int_{B(y, r_1/(1+\delta))} \lambda_1 \mathbb{P}(x) dx$ and, dividing by λ_1 , we can write the inequality (21) as

$$\int_{B(y, r_1/(1+\delta))} \mathbb{P}(x) dx \leq \frac{1 + O(\lambda_1)}{\frac{\delta^2 r_1^2 F(r_1, \varepsilon, \alpha)}{(1+\delta)^2 \varepsilon} + 1} \int_{B(y, r_1)} \mathbb{P}(x) dx, \tag{23}$$

Thanks to Remark 4.1, we can take the limit $\lambda_1 \rightarrow 0$ to get rid of the term $O(\lambda_1)$: in fact it is the only term in (23) which depends on λ_1 or λ_2 . Using then the lower bound estimate $F(r_1, \varepsilon, \alpha) \geq \theta(2\alpha)/2$ in (23) we get precisely

$$\int_{B(y, \tilde{r}/(1+\delta))} \mathbb{P}(x) dx \leq \frac{1}{\delta^2 \tilde{r}^2 \frac{\theta(2\alpha)}{2(1+\delta)^2 \varepsilon} + 1} \int_{B(y, \tilde{r})} \mathbb{P}(x) dx.$$

In order to understand for which α and δ this inequality holds, we have to ensure that the two conditions (22) are satisfied, that is

$$\theta(2\alpha) \geq \max \left\{ 8(N - 1)\Theta\left(\frac{\beta}{2}\right), 256\frac{\varepsilon C(\delta)}{\beta^2} \right\}; \tag{24}$$

notice that α_0 can be characterized as the maximal α for which there exists some δ for which (24) is satisfied that is when $C(\delta)$ as small as possible, which is approximately achieved for $\delta = \frac{2}{3N}$. With this choice we have $C(2/(3N)) \leq 26N^2$ and thus

$$\theta(2\alpha_0) \geq 8 \max \left\{ (N - 1)\Theta\left(\frac{\beta}{2}\right), 832\frac{\varepsilon N^2}{\beta^2} \right\}. \tag{25}$$

□

We will now iterate the estimate in Proposition 5.1

Theorem 5.1 *Let us consider ρ and β such that $\kappa(\rho, \beta) \leq \frac{1}{4(N-1)}$. Then let us consider $\alpha < \alpha_0$ (as in Proposition 5.1) and suppose $A := \frac{\alpha^2\theta(2\alpha)}{8\varepsilon} \gg N^2$. Then if \mathbb{P} minimizes (11) we have that*

$$\int_{D_{\alpha/2}} \mathbb{P}(x) \, dx \leq e^{-\frac{1}{6}\sqrt{\frac{\alpha^2\theta(2\alpha)}{8\varepsilon}}} \int_{D_{2\alpha}} \mathbb{P}(x) \, dx.$$

Proof Let us consider δ such that $\delta^2 A = e^2$. By the hypothesis on A we have $\delta \ll 1/N$; in particular, by (20) we can estimate $C(\delta) \leq \frac{2}{\delta^2}$, and then it is easy to see that $\theta(2\alpha) > 256\varepsilon C(\delta)/\beta^2$ and thus we can apply Proposition 5.1 with $\tilde{r} = \alpha_k = \frac{\alpha}{2}(1 + \delta)^{-k}$ to obtain for every $y \in D_\alpha$

$$\begin{aligned} \int_{B(y, \alpha_{k+1})} \mathbb{P}(x) \, dx &\leq \frac{1}{\delta^2 \alpha^2 \frac{\theta(2\alpha)}{8(1+\delta)^{2k+2\varepsilon}} + 1} \int_{B(z, \alpha_k)} \mathbb{P}(x) \, dx \\ &\leq \frac{(1 + \delta)^{2k+2}}{e^2} \int_{B(y, \alpha_k)} \mathbb{P}(x) \, dx. \end{aligned} \tag{26}$$

We can now iterate the estimate for $k = 0, \dots, k_0$ where $(1 + \delta)^{2k_0+2} \leq e^2 \leq (1 + \delta)^{2k_0+4}$. At that point we have

$$\begin{aligned} \int_{B(y, \alpha/2e)} \mathbb{P}(x) \, dx &\leq \int_{B(y, \alpha_{k_0+1})} \mathbb{P}(x) \, dx \leq \frac{(1 + \delta)^{(k_0+1)(k_0+2)}}{(e^2)^{k_0+1}} \int_{B(y, \alpha_0)} \mathbb{P}(x) \, dx \\ &\leq e^{-k_0} \int_{B(y, \alpha/2)} \mathbb{P}(x) \, dx. \end{aligned}$$

Integrating this inequality for $y \in D_\alpha$ we get

$$\omega_{3N} \left(\frac{\alpha}{2e}\right)^{3N} \int_{D_{\alpha/2}} \mathbb{P}(y) \, dy \leq \int_{D_\alpha} \int_{B(y, \alpha/2e)} \mathbb{P}(x) \, dx \, dy$$

$$\begin{aligned} &\leq e^{-k_0} \int_{D_\alpha} \int_{B(y, \alpha/2)} \mathbb{P}(x) \, dx \, dy \\ &\leq e^{-k_0} \omega_{3N} \left(\frac{\alpha}{2}\right)^{3N} \int_{D_{2\alpha}} \mathbb{P}(y) \, dy. \end{aligned}$$

Now we notice that $k_0 + 2 \geq \frac{\ln(e^2)}{2 \ln(1+\delta)} \geq \frac{2}{4\delta} = \frac{\sqrt{A}}{2e}$ and so $e^{-k_0} \leq 10e^{-\frac{\sqrt{A}}{2e}}$. In particular

$$\int_{D_{\alpha/2}} \mathbb{P}(y) \, dy \leq 10e^{-\frac{\sqrt{A}}{2e} + 3N} \int_{D_{2\alpha}} \mathbb{P}(y) \, dy;$$

notice that since $A \gg N^2$ we have $\ln(10) + \frac{\sqrt{A}}{2e} - 3N \geq \frac{\sqrt{A}}{6}$. □

Proof (Theorems 1.1 and 1.2) First we notice that if ψ_ε is a minimizer for (1) in the bosonic case then $\mathbb{P}_\varepsilon = |\psi_\varepsilon|^2$ is a minimizer for (11). Then we notice that if $\theta(2\alpha) \leq 8(N - 1)\Theta(\beta/2)$ and $\varepsilon N^2 \ll \alpha^2\theta(2\alpha)$, we have also $\alpha < \alpha_0$ and so we can apply Theorem 5.1. From that we finish using that \mathbb{P}_ε is a probability density and so $\int_{D_{2\alpha}} \mathbb{P}_\varepsilon(y) \, dy \leq 1$. The conclusions for Theorem 1.1 are then implied by using $\theta(t) = \Theta(t) = 1/t$. □

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Declarations

Conflict of interest On behalf of all authors, the corresponding author states that there is no conflict of interest.

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