1. Introduction. In this paper we investigate the large-scale limit of the equilibrium state of Dyson’s hierarchical vector valued $p$ dimensional, $p \geq 2$, model with parameter $c$, $1 < c < \sqrt{2}$, at low temperatures. More precisely, in Theorem 1 we construct a probability measure $\hat{\mu} = \hat{\mu}(T)$ on $(\mathbb{R}^p)^\mathbb{Z}$ with $\mathbb{Z} = \{1, 2, \ldots\}$ which is an equilibrium state of the model. In Theorem 2 we determine the large-scale limit of a $\hat{\mu}$ distributed random field together with the right scaling, i.e. we prove that if $\bar{\sigma} = \{\bar{\sigma}(j) = (\sigma^{(1)}(j), \ldots, \sigma^{(p)}(j)) \in \mathbb{R}^p, j \in \mathbb{Z}\}$ is a $\hat{\mu}$ distributed random field then the finite dimensional distributions of the random fields

$$
\mathcal{R}_n \bar{\sigma} = \left\{(\mathcal{R}_n \sigma^{(1)}(j), \ldots, \mathcal{R}_n \sigma^{(p)}(j)) \in \mathbb{R}^p, j \in \mathbb{Z}\right\},
$$

$$
\mathcal{R}_n \sigma^{(1)}(j) = c^n 2^{-n} \sum_{k=(j-1)2^n+1}^{j2^n} \left[\sigma^{(1)}(k) - E\sigma^{(1)}(k)\right], \quad j \in \mathbb{Z},
$$

$$
\mathcal{R}_n \sigma^{(s)}(j) = c^{n/2} 2^{-n} \sum_{k=(j-1)2^n+1}^{j2^n} \sigma^{(s)}(k), \quad j \in \mathbb{Z}, \quad s = 2, \ldots, p
$$

tend to those of a limit random field, and describe the finite dimensional distributions of this limit field.

The distributions of the fields $\mathcal{R}_n \sigma$ defined in (1.1)–(1.3) are called the renormalizations of the distribution $\hat{\mu}$ of the underlying field $\sigma$. More precisely, they are its renormalization with parameters $\alpha = 1 - \frac{\log c}{\log 2}$ in the first coordinate and $\bar{\alpha} = 1 - \frac{1}{2} \frac{\log c}{\log 2}$ in the coordinates $s = 2, \ldots, p$, because we multiplied by $2^{-n\alpha}$ in
(1.2), by \(2^{-n\alpha}\) in (1.3), and the number of summands is \(2^n\) in these formulas. If the finite dimensional distributions of the fields \(\mathcal{R}_n\sigma\) converge to those of a limit field then this limit field, more precisely its distribution, is called the large-scale limit of the measure \(\bar{\mu}\).

Given some \(h \in R^1, h \geq 0\), and positive integer \(N\) let the Gibbs measure \(\mu_h^N = \mu_N^h(T,t)\) be defined on \((R^p)^{2N}\) with density function

\[
p_N^h(x_1, \ldots, x_{2N}) = p_N^h(x_1, \ldots, x_{2N}, t, T),
\]

\[
x_j = (x_j^{(1)}, \ldots, x_j^{(p)}) \in R^p, \quad j = 1, \ldots, 2^N,
\]

where

\[
p_N^h = Z_N^{-1}(T,t,h) \exp\left\{-\frac{1}{T} \left( - \sum_{i=1}^{2^N-1} \sum_{j=i+1}^{2^N} U(i,j)x_i x_j - h \sum_{j=1}^{2^N} x_j^{(1)} \right) \right\} \prod_{j=1}^{2^N} p(x_j, t),
\]

\[
Z_N = \int \exp\left\{-\frac{1}{T} \left( - \sum_{i=1}^{2^N-1} \sum_{j=i+1}^{2^N} U(i,j)x_i x_j - h \sum_{j=1}^{2^N} x_j^{(1)} \right) \right\} \prod_{j=1}^{2^N} p(x_j, t) \, dx_j
\]

is the grand partition function, and \(p(x, t)\) is defined in (1.3) of Part I. Let \(p_N^h(x)\) denote the density function of the average \(2^{-N} \sum_{j=1}^{2^N} \sigma(j)\) of the \(\mu_h^N\) distributed random vector \((\sigma(1), \ldots, \sigma(2^n))\). Put \(\mu_N = \mu_h^N, p(x_1, \ldots, x_{2N}) = p_N^h(x_1, \ldots, x_{2N})\) and \(p_N(x) = p_N^h(x)\) in the case \(h = 0\).

In Part I we have described the asymptotic behaviour of the above defined density function \(p_N(x)\). The result of Theorem A formulated below are contained in Theorems 1, 2 and Lemma 13 of Part I. Let us consider the integral equation

\[
g(x) = \left(\frac{2}{c\sqrt{\pi}}\right)^{p-1} \int_{R^p} \exp(-v^2)g\left(\frac{x}{c} + u + \frac{v^2}{2}\right)g\left(\frac{x}{c} - u + \frac{v^2}{2}\right) \, du \, dv,
\]

\[
x, \; u \in R^1, \quad v \in R^{p-1},
\]

where \(v^2\) denotes scalar product. In Part I we have proved that equation (1.5) has a unique non-trivial (i.e. not identically zero) solution in the class of functions \(A = \{g, \int e^{tx}|g(x)|\,dx < \infty \text{ if } t < t_0(g), \; t_0(g) > 0\}\). In this work we consider this function as the solution of equation (1.5). It is a density function which is positive for all \(x\). Since the function \(p_n(x, t)\) depends on \(x\) only through \(|x|\) we can define a function \(\tilde{p}_n(z) = \tilde{p}_n(z, t, T), \; z \in R^1,\) such that \(p_n(x) = \tilde{p}_n(|x|)\) for all \(x \in R^p\).

Now we formulate the following

**Theorem A.** If \(1 < c < \sqrt{2}\) then there exist some thresholds \(T_0 = T_0(c) > 0\) such that for all \(0 < T < T_0, \; 0 < t < t_0, \; t_0 = t_0(c), \; \) (\(t\) is the parameter of \(p(x, t)\) in formula (1.3) of Part I) the following relations hold:
There are some $M = M(c,T,t) > 0$ and $n_0 = n_0(c,T,t) > 0$ such that for $n > n_0$
\begin{equation}
(1.6) \quad c^{-n} p_n(x,T) = c^{-n} \bar{p}_n(|x|,T)
\quad = B \exp \left\{ -\frac{a_0 c^n}{T} M(|x| - M) \right\} g \left( \frac{a_1 c^n}{T} M(|x| - M) \right) \left( 1 + r_n(x) \right)
\end{equation}
for $-\eta n c^{-n} < |x| - M < \eta n^{1/\alpha} c^{-n}$ with some $B = B(c,T,t) > 0$, $\eta = \eta(c,T,t) > 0$, and the error term $r_n(x)$ satisfies the inequality
\begin{equation}
(1.7) \quad |r_n(x)| \leq K q^n
\end{equation}
with some $K > 0$ and $0 < q < 1$ depending on $c$, $T$ and $t$. In formula (1.6) $g(x)$ is the solution of the equation (1.5), $a_0 = \frac{2}{2-c}$, $a_1 = a_0 + 1$ and $\alpha = \frac{\log 2}{\log c}$.
\begin{equation}
(1.8) \quad c^{-n} \bar{p}_n(x,T) \leq K q^n \exp \left\{ -L(c^n|x-M|)^{2+\delta} \right\} \quad \text{if } x > M + \eta n^{1/\alpha} c^{-n}
\end{equation}
with some $\delta > 0$, $K > 0$ and $L > 0$ which depend on $c$, $T$ and $t$. The solution of the equation (1.6) satisfies the inequality
\begin{equation}
(1.9) \quad 0 < g(x) < C \exp(-A x^\alpha) \quad \text{for } x > 0
\end{equation}
with some $C > 0$, $A > 0$. We have
\begin{equation}
(1.10) \quad c^{-n} \bar{p}_n(x,T) \leq C_1 \exp\{ -C_2 c^n |x - M| \} \quad \text{for } 0 < x < M
\end{equation}
with some $C_1 > 0$, $C_2 > 0$ depending on $c$, $T$ and $t$. We also have
\begin{equation}
M^2 = \frac{a_0 - T}{tT} + R(t,T)
\end{equation}
with some $|R(t,T)| \leq \text{const.}$, and such that $R(t,T) \to 0$ and $T \to 0$.

Given some integers $N \geq k \geq 0$ we define the probability measure $\mu_{k,N}^h$ as the projection of the measure $\mu_N^h$ to the first $2^k$ coordinates, i.e. $\mu_{k,N}^h$ is a probability measure on $(R^p)^{2^k}$, and for all measurable $A \subset (R^p)^{2^k}$ $\mu_{k,N}^h(A) = \mu_N^h(A) = \mu_N^h \left( A \times (R^p)^{2^k-2^k} \right)$. Our first result is the following

**Theorem 1.** Let the conditions of Theorem A be satisfied. Consider an arbitrary sequence of real numbers $h_N$, $N = 0,1,2,\ldots$ such that
\begin{equation}
(1.11) \quad \frac{2}{2-c} \frac{M}{T} \left( \frac{c}{2} \right)^N \leq \frac{h_N}{T} \leq D \left( \frac{c}{2} \right)^N
\end{equation}
with some $\infty > D > \frac{2}{2-c} \frac{M}{T}$, where $M$ and $T$ are the same as in Theorem A. Then the measures $\mu_{N}^{h_N}$ tend to a probability measure $\bar{\mu} = \bar{\mu}(t,T,c)$ on $(R^p)^{2^k}$. More precisely, for all $k \geq 0$ the measures $\mu_{k,N}^{h_N}$ converge to the projection of $\bar{\mu}$ to the first $2^k$ coordinates in variational metric as $N \to \infty$. The measure $\mu$ does not depend on the choice of sequences $h_N$.

Then we prove the following
Theorem 2. Let \( \sigma = \{ \sigma(n) = (\sigma^{(1)}(n), \ldots, \sigma^{(p)}(n)) \in R^p, \ n \in Z \} \) be a \( \tilde{\mu} \) distributed random field with the distribution \( \tilde{\mu} \) defined in Theorem 1. Then the finite dimensional distributions of the random fields \( R_n \sigma \) defined in (1.1), (1.2), (1.3) tend to those of a random field \( Y = (Y(n) = (Y^{(1)}(n), \ldots, Y^{(p)}(n)) \in R^p, \ n \in Z) \). For all \( k \geq 0 \) the density function \( h_k(x_1, \ldots, x_{2^k}) \), \( x_j = (x_j^{(1)}, \ldots, x_j^{(p)}) \in R^p \), of the random vector \( (Y(1), \ldots, Y(2^k)) \) is given by the formula

(1.12)

\[
\begin{align*}
    h_k(x_1, \ldots, x_{2^k}) &= C(k) \exp \left\{ \frac{-1}{T} \sum_{s=2}^{p} \left( \frac{1}{2} \sum_{j=1}^{2^k} x_j^{(s)} \right)^2 - \frac{2 - c}{c} \left( \frac{c}{4} \right)^k \left( \sum_{j=1}^{2^k} x_j^{(s)} \right)^2 \\
    &\quad + \sum_{i=1}^{2^k-1} \sum_{j=i+1}^{2^k} U(i, j) x_i^{(s)} x_j^{(s)} \right\} \prod_{j=1}^{2^k} \frac{4 - c}{(2-c)T} \left( \frac{1}{2} \sum_{s=2}^{p} x_j^{(s)^2} \right)\right),
\end{align*}
\]

where the function \( g \) is defined in (1.5), the constant \( M \) is the same as in Theorem A, and \( C(k) \) is an appropriate norming constant.

In Appendix E we prove the following

Theorem B. The measure \( \tilde{\mu} = \tilde{\mu}(T, t, c) \) constructed in Theorem 1 is a Gibbs state with Hamiltonian \( \mathcal{H} \) and free measure \( \nu \) defined in formulas (1.1)–(1.3) of Part I at temperature \( T \).

Theorem B is very plausible. Its proof depends on a rather standard limiting procedure in statistical physics literature. Nevertheless, we have found no result which could have been directly applied in our case. We present the proof of Theorem B in Appendix E.

Let us discuss the role of condition (1.11) in Theorem 1. The lower bound

(1.11')

\[
    h_N > \frac{2}{2-c} M \left( \frac{2}{c} \right)^N
\]

is essential in Theorem 1, it is needed to get a pure state with magnetization in the direction \( e_1 = (1, 0, \ldots, 0) \) for the limit measure \( \tilde{\mu} \). If it were violated we would get a Gibbs state with Hamiltonian \( \mathcal{H} \) again for the limit, but this Gibbs state would be a mixture of Gibbs states with different directions of magnetization, and it is not natural to renormalize such a mixture. On the other hand the upper bound for \( h_N \) in (1.11) seems not to be essential. We believe that the same limit measure \( \tilde{\mu} \) would be obtained for any sequence \( h_N, \ h_N > 0 \) satisfying (1.11') or with the help of the double limiting procedure \( \mu^h = \lim_{N \to \infty} \mu_N^h, \ h > 0, \ \tilde{\mu} = \lim_{h \to 0} \mu^h \). This second way was chosen to construct the equilibrium state in the case \( \sqrt{2} < c < 2 \) in paper [5]. However, to prove these statements we would need a large deviation result on the behaviour of \( p_n(x) \) which is stronger than Theorem A. Since we are not able to prove such a result we have proved Theorem 1 under the condition (1.11), but we think that this is not an essential restriction.

In formula (1.12) we have a quadratic form inside the exponent. This means that the random variables \( Y^{(\ell)}(j), \ \ell = 2, \ldots, p \) appearing in Theorem 2 are jointly Gaussian. We describe the structure of this limit field in more detail. The random
fields \( \{ R_n \sigma^{(s)}(j), j \in \mathbb{Z} \} \), \( s = 2, \ldots, p \), and \( \{ M R_n \sigma^{(1)}(j) + \frac{1}{2} \sum_{s=2}^{p} R_n \sigma^{(s)}(j)^2, j \in \mathbb{Z} \} \) tend to independent random fields as \( n \to \infty \). The limit of the random fields \( \{ R_n(\sigma^{(s)}(j)), j \in \mathbb{Z} \} \) is the (disregarding a multiplying factor) unique Gaussian self-similar field with self-similarity parameter \( 1 - \frac{1}{2} \log c \), whose distribution is invariant under all permutations of the index set \( \mathbb{Z} \) which preserves the hierarchical distance \( d(i, j) \). The random fields \( \{ M R_n \sigma^{(1)}(j) + \frac{1}{2} \sum_{s=2}^{p} R_n \sigma^{(s)}(j), j \in \mathbb{Z} \} \) tend to a random field consisting of independent random variables with the density function \( \frac{2-c}{(4-c)g(\frac{4-c}{2})} x \). This is a quadratic functional of a Gaussian field (see Lemma 12 in Part I).

The above result can also be interpreted in the following way: Given a \( \bar{\mu} \) distributed random field \( \sigma(n), n \in \mathbb{Z} \), define the absolute value of the appropriately normalized partial sums \( |R_n| \sigma(j) = c^n 2^{-n} \left( |R_n \sum_{k=(j-1)2^n+1}^{j2^n} \sigma(k)| - M \right) \), \( j \in \mathbb{Z} \). Then the random fields \( R_n \sigma^{(s)}(j), j = 2, \ldots, p \), and the random fields \( |R_n| \sigma(j) \) tend in distribution to independent random fields. The limit of \( R_n \sigma^{(s)}(j) \) \( j = 2, \ldots, p \) is Gaussian, and the limit of \( |R_n| \sigma(j) \) consists of independent random variables. This follows immediately from the above description of the limit behaviour of the fields \( R_n \sigma \) together with the observation that \( |R_n| \sigma(j) - (R_n \sigma^{(1)}(j) + \frac{1}{2M} \sum_{s=2}^{p} R_n \sigma^{(s)}(j)^2) \to 0 \) stochastically as \( n \to \infty \).

We believe that the above property is a special case of a more general law. Let us remark that an analogous statement also holds in the case \( \sqrt{2} < c < 2 \), but this is a degenerate case. It follows from the results of [5] that if \( \{ \sigma(j), j \in \mathbb{Z} \} \) is a \( \bar{\mu} = \bar{\mu}(c) \), \( \sqrt{2} < c < 2 \) distributed random field with the equilibrium state \( \bar{\mu} \) constructed in [5] then the random fields

\[
|R_n| \sigma(j) = 2^{-n/2} \left( \sum_{k=(j-1)2^n+1}^{j2^n} \sigma(k) - M \right), \quad j \in \mathbb{Z},
\]

have the same limit as the random fields

\[
R_n \sigma^{(1)}(j) = 2^{-n/2} \sum_{k=(j-1)2^n+1}^{j2^n} \left( \sigma^{(1)}(k) - M \right)
\]

as \( n \to \infty \), since in this case \( |R_n| \sigma(j) - R_n \sigma^{(1)}(j) \to 0 \). This limit consists of independent (Gaussian) random variables which is also independent of the limit of the random fields \( R_n \sigma^{(j)}, s = 2, \ldots, p \).

The method of this paper is very similar to that of [5]. The two main steps in the proofs consist of the description of the limit behaviour of the function \( p_n(x) \) done in Part I, and a good asymptotic formula for the Radon–Nikodym derivatives \( \frac{dp_n^{(s)}(x)}{dp_n} \). Then an appropriate limiting procedure supplies the proof of Theorems 1 and 2. The investigation of the Radon–Nikodym derivatives can be considered as an adaptation of the method of [5] to the present case. The main difference between the two cases is that now \( p_n(x) \) is not asymptotically Gaussian. But although the Radon–Nikodym derivative \( \frac{dp_n^{(s)}(x)}{dp_n} \) depends on \( p_n(x) \), its asymptotic behaviour does
not. As we shall see, in the investigation of the asymptotic behaviour of the above Radon–Nikodym derivative we only need some estimates on the tail behaviour of \( p_n(x) \), but not its explicit form. This is the reason why we can adapt the method of [5].

2. On the basic estimates needed in the proof. Reduction to integral equations. We need a good asymptotic formula for the Radon–Nikodym derivative \( \frac{d\mu^h}{d\mu} \). It can be expressed exactly with the help of the following formulas:

\[
\begin{align*}
(2.1) & \quad \frac{d\mu^h}{d\mu_n}(x_1, \ldots, x_{2^n}) = f_h^{N,N}(2^{-n} \sum_{j=1}^{2^n} x_j), \quad n \leq N \\
(2.2) & \quad f^{h_N}(x) = K(N, h_N) \exp \left( \frac{2^Nh_Nx^{(1)}}{T} \right) \\
(2.3) & \quad f^{h_N}(x) = K(n, N, h_N)S_n f^{h_{N+1,N}}(x) \\
\end{align*}
\]

with

\[
(2.3') \quad S_n f(x) = \int_{\mathbb{R}^p} \exp \left( \frac{c^n}{T} xy \right) f \left( \frac{x + y}{2} \right) p_n(y) \, dy
\]

where \( K(n, N, h_N) \) are appropriate norming factors, \( xy \) denotes scalar product, and \( p_n \) is the density function appearing in Theorem A. For scalar valued models formulas (2.1)–(2.3') are proved in the main formula in [4]. The proof for the vector valued case is the same, but since the proof in [4] is a bit complicated we present it in Appendix C.

Let us define the sequences \( g_n = g_n(N, h_N) \) and \( A_n = A_n(N, h_N) \) by the recursive relations

\[
\begin{align*}
(2.4) & \quad g_N = g_N(N, h_N) = \frac{2^Nh_N}{T} \\
(2.4') & \quad g_n = g_n(N, h_N) = \frac{g_{n+1}}{2} + \frac{c^n}{T} M \quad \text{for } n < N \\
(2.5) & \quad A_N = A_N(N, h_N) = 0 \\
(2.5') & \quad A_n = A_n(N, h_N) = \frac{A_{n+1}}{4} + \frac{(\frac{c^n}{T} + \frac{A_{n+1}}{2})^2}{2c^n + \frac{A_{n+1}}{M} - A_{n+1}} \quad \text{for } n < N, \\
\end{align*}
\]

where \( M \) and \( T \) are the same as in Theorem A. In Section 7 of [6] we have claimed that

\[
f_n(x) = f_{n,N}(x) \sim K_n \exp \left\{ g_n(x^{(1)} - M) + A_n \sum_{s=2}^{p} x^{(s)/2} \right\},
\]
and have given a heuristic explanation. In the following Proposition 1 we formulate this result in a more precise form. For the sake of simpler notation we assume that \( R^0 = R^2 \). From now on \( C, \ C_1, K, \ L \) etc. denote appropriate constants. The same letter may denote different constants in different formulas. Let us define the domains

\[
\Omega_n^1 = \{ x \in R^2, |x| - M < c^{-0.4n}, |x(2)| < c^{-0.4n}, \ x(1) > 0 \} \\
(2.6') \Omega_n^2 = \{ x \in R^2, |x| - M < c^{-0.4n} \} - \Omega_n^1 \\
(2.6'') \Omega_n^3 = \{ x \in R^2, |x| - M \geq c^{-0.4n} \}.
\]

Clearly \( \Omega_n^1 \cup \Omega_n^2 \cup \Omega_n^3 = R^2 \). Now we formulate the following

**Proposition 1.** For all \( q \), \( c^{-0.2} < q < 1 \), there is some \( n_0 = n_0(T, M, c, D, q) \) such that if \( (1.11) \) holds and \( N \geq n \geq n_0 \) then the Radon–Nikodym derivative \( f_n(x) = f_{n,N}^n(x) \) appearing in \( (2.1) \) satisfies the following relations:

a) In the domain \( \Omega_n^1 \)

\[
f_n(x) = L_n \exp \left\{ g_n \left( x(1) - M \right) + A_n x(2)^2 + \varepsilon_n(x) \right\}
\]

with

\[
\sup_{x \in \Omega_n^1} |\varepsilon_n(x)| \leq q^n.
\]

b) In the domain \( \Omega_n^2 \)

\[
0 \leq f_n(x) \leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} + q^n \right\}.
\]

c) In the domain \( \Omega_n^3 \)

\[
0 \leq f_n(x) \leq L_n \exp \left\{ \frac{g_n}{2M} (|x|^2 - M^2) \right\},
\]

where the numbers \( A_n \) and \( g_n \) are defined in \( (2.4)-(2.5') \), and \( L_n = L_n(N, h_N) \) is an appropriate norming constant.

We also prove the following result which is a slight modification of Lemma 1 in [5].

**Lemma 1.** Let us choose some integer \( N \) and \( h_N > 0 \). Define the sequences \( g_n \) and \( A_n \), \( 0 \leq n \leq N \), by formulas \( (2.4)-(2.5') \) and put \( \bar{g}_n = c^{-n} g_n \), \( \bar{A}_n = c^{-n} A_n \). If \( h_N \) satisfies relation \( (1.11) \) then \( \bar{g}_N \geq \bar{g}_{N-1} \geq \cdots \geq \bar{g}_0 \geq \bar{g} \) and \( 0 = \bar{A}_N = \bar{A}_{N-1} = \cdots \leq \bar{A}_0 \leq \bar{A} \) with \( \bar{g} = 2^{2-M} \bar{T} \) and \( \bar{A} = 2^{2-M} \). If the relations \( N > N_0 \) and \( N > n^B \) also hold with some appropriate \( N_0 = N_0(c, M, T, D) \) and \( B = B(c, M, T, D) \) then \( |\bar{g}_n - \bar{g}| < 4^{-n} \), \( |\bar{A}_n - \bar{A}| < 4^{-n} \).

Proposition 1 together with the characterization of the asymptotic behaviour of the sequences \( g_n \) and \( A_n \) made in Lemma 1 supplies a good asymptotic formula for
the Radon–Nikodym derivative $f_n$. Here $\Omega_n^1$ is the typical region, where we have a good asymptotic formula, in $\Omega_n^2$ and $\Omega_n^3$ we have only given an upper bound. Actually we are interested in the density function

$$p_n\left(2^{-n}\sum_{j=1}^{2^n}x_j\right)f_{n,N}^{h_N}(2^{-n}\sum_{j=1}^{2^n}x_j)$$

of the measure $\mu_{n,N}$. The tail behaviour of the functions $p_n(x)$ and $f_n(x)$ together show that $2^{-n}\sum_{j=1}^{2^n}x_j$ is contained in $\Omega_n^3$ with a negligible small $\mu_{n,N}$ probability. It is contained in $\Omega_n^2$ also with a small probability, since in this domain $f_n(x)$ is small. To see it, let us observe that by Lemma 1

$$\frac{g_n}{2M} - A_n \geq c_n\left(\frac{g_n}{2M} - \bar{A}\right) = \frac{c_n}{T}\left(\frac{1}{2 - c} - \frac{2 - c}{c}\right) = \frac{c_n}{T}\left(4 - c\right)\left(\frac{c - 1}{2 - c}\right) > 0,$$

hence the term $-\left(\frac{g_n}{2M} - A_n\right)c^{-0.8n}$ in the exponent of (2.8), makes this upper bound (2.8) sufficiently small for our purposes.

In Section 7 of [6] we have given a heuristic argument for formula (2.7). The following remark explains the content of the estimate (2.8).

**Remark.** If $x \in \Omega_n^1$ then

$$|x| = \left(x^{(1)^2} + x^{(2)^2}\right)\frac{1}{2} = x^{(1)}\left(1 + \frac{x^{(2)^2}}{x^{(1)^2}}\right)^{1/2} = x^{(1)} + \frac{x^{(2)^2}}{2x^{(1)}} + O(x^{(2)^4})$$

$$= x^{(1)} + \frac{x^{(2)^2}}{2M} + O\left(x^{(2)^4} + x^{(2)^2}|x^{(1)} - M|\right) = x^{(1)} + \frac{x^{(2)^2}}{2M} + O(c^{-1.2n}),$$

hence

$$\exp\left\{g_n(|x| - M) - \left(\frac{g_n}{2M} - A_n\right)x^{(2)^2}\right\} = \exp\left\{g_n(x^{(1)} - M) + A_nx^{(2)^2} + O(c^{-0.2n})\right\}.$$

The above calculation shows that on the boundary of the domains $\Omega_n^1$ and $\Omega_n^2$, the right-hand side of formulas (2.7) and (2.8) have the same magnitude. The estimate (2.8) expresses the fact that this is the worst region, where the weakest upper bound can be given for $f_n(x)$ in $\Omega_n^2$.

With the help of Proposition 1, Lemma 1 and Theorem A we are able to carry out a limiting procedure which supplies Theorem 1. Moreover, it yields the following Proposition 2. Let $\bar{\mu}_n$ denote the projection of the measure $\bar{\mu}$ constructed in Theorem 1 to the first $2^n$ coordinates, i.e. let $\bar{\mu}_n$ be the measure on $(R^2)^{2^n}$ defined by the relation $\bar{\mu}_n(A) = \bar{\mu}\left(A \times (R^2)^\infty\right)$ for all measurable sets $A \subset (R^2)^{2^n}$. The following result holds true:
Proposition 2. For all \( q < 1 \), there is some \( n_0 = n_0(T,M,c,q) \) such that for all \( n \geq n_0 \) the measure \( \tilde{\mu}_n \) is absolute continuous with respect to the measure \( \mu_n \), and its Radon–Nikodym derivative satisfies the relations:

\[
\frac{d\tilde{\mu}_n}{d\mu_n}(x_1, \ldots, x_{2^n}) = \tilde{f}_n\left(2^{-n} \sum_{i=1}^{2^n} x_i\right)
\]

and

a) For \( x \in \Omega_1^n \)

\[
\tilde{f}_n(x) = L_n \exp\left\{ c^n \bar{g}(x^{(1)} - M) + c^n \bar{A}x^{(2)} + \varepsilon_n(x)\right\}
\]

with

\[
\sup_{x \in \Omega_1^n} |\varepsilon_n(x)| \leq q^n.
\]

b) For \( x \in \Omega_2^n \)

\[
0 \leq \tilde{f}_n(x) \leq L_n \exp\left\{ c^n \bar{g}(\|x\| - M) - c^0.2n \left( \frac{\bar{g}}{2M} - \bar{A} \right) + q^n\right\}.
\]

c) For \( x \in \Omega_3^n \)

\[
0 \leq \tilde{f}_n(x) \leq L_n \exp\left\{ \frac{\bar{g}c^n}{4M}(x^2 - M^2)\right\} \quad \text{if } 0 < x < M - c^{-0.4n},
\]

\[
0 \leq \tilde{f}_n(x) \leq L_n \exp\left\{ \frac{\bar{g}c^n}{M}(x^2 - M^2)\right\} \quad \text{if } x > M + c^{-0.4n},
\]

with \( \bar{g} = \frac{2}{2c} M, \bar{A} = \frac{2-c}{cT} \) and an appropriate norming constant \( L_n \). This norming constant satisfies the relation

\[
C_1 < c^{-n/2}L_n < C_2 \quad \text{with some } 0 < C_1 < C_2 < \infty.
\]

Theorem 2 can be deduced from Proposition 2 and Theorem A.

Let us finally remark that the function \( f_n(x) = f_{n,N}^h(x) \) clearly satisfies Proposition 1 for \( n = N \), since in this case \( f_N(x) = L_N \exp\{g_N(x^{(1)} - M)\} \). Hence Proposition 1 follows from Lemma 1 and the following

Proposition 1’. For all \( q < 1 \), there exists some \( n_0 = n_0(T,M,c,D,q) \) such that if for \( n \geq n_0 \) the function \( f(x) \) satisfies the following relations with some \( \bar{g}c^{n+1} < g_{n+1} \leq Dc^{n+1}, 0 \leq A_{n+1} \leq \bar{A}c^{n+1}, \bar{g} = \frac{2}{2c} M, \bar{A} = \frac{2-c}{cT}, D > \bar{g} \):

a) For \( x \in \Omega_1^{n+1} \)

\[
f(x) = \exp\left\{ g_{n+1}(x^{(1)} - M) + A_{n+1}x^{(2)} + \varepsilon_{n+1}(x)\right\}
\]

\[
\sup_{x \in \Omega_1^{n+1}} |\varepsilon_{n+1}(x)| \leq q^{n+1}.
\]
b) For \( x \in \Omega^2_{n+1} \)

\[
0 \leq f(x) \leq \exp \left\{ g_{n+1}(|x| - M) + \left( \frac{g_{n+1}}{2M} - A_{n+1} \right) e^{-0.8(n+1) + q^{n+1}} \right\};
\]

(2.16)

\[
0 \leq f(x) \leq \exp \left\{ \frac{g_{n+1}}{2M} (|x|^2 - M^2) \right\};
\]

(2.17)

then the function \( S_n f(x) \) defined by (2.3) satisfies, with the constants \( g_n \) and \( A_n \) defined by (2.4) and (2.5) with the above \( g_{n+1} \) and \( A_{n+1} \), the following relations with some appropriate norming constant \( L_n \):

a) In the domain \( \Omega^1_n \)

\[
S_n f(x) = L_n \exp \left\{ g_n(x^{(1)} - M) + A_n x^{(2)^2} + \varepsilon_n(x) \right\}
\]

(2.18)

with

\[
\sup_{x \in \Omega^1_n} |\varepsilon_n(x)| \leq q^n.
\]

(2.18’)

b) In the domain \( \Omega^2_n \)

\[
0 \leq S_n f(x) \leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) e^{-0.8n + q^n} \right\}.
\]

(2.19)

c) In the domain \( \Omega^3_n \)

\[
0 \leq S_n f(x) \leq L_n \exp \left\{ \frac{g_n}{2M} (|x|^2 - M^2) \right\}.
\]

(2.20)

3. The proof of Lemma 1. The proof is a modification of that given for Lemma 1 in [5]. Simple calculation shows that \( \bar{g}_n - \bar{g} = (\frac{\bar{c}}{2})^{N-n} (\bar{g}_N - \bar{g}). \) The statements of Lemma 1 about the sequence \( g_n \) follow from this identity. To investigate \( \bar{A}_n \) let us introduce the function

\[
T(x, g) = \frac{c}{4} x + \left( \frac{1}{2} + \frac{\bar{c}}{2M} \right)^2, \quad x \in R^1, \quad g \in R^1.
\]

Clearly, \( \bar{A}_n = T(\bar{A}_{n+1}, \bar{g}_{n+1}) \). On the other hand \( T(\bar{A}, \bar{g}) = \bar{A} \), and some calculation shows that \( T \) has the following monotonicity properties: \( T(x, g) < T(x, g') \) if \( 0 < x < A \) and \( g > g' > \bar{g} \); and \( T(x', g) > T(x, g) \) if \( 0 < x < x' < A \) and \( g > \bar{g} \). (These properties follow e.g. from the relations

\[
\partial_{\bar{g}} T(x, g) = -\frac{c}{M} \left( \frac{1}{2} + \frac{\bar{c}}{2M} \right)^2 < 0,
\]
and
\[ \frac{\partial}{\partial x} \mathbb{T}(x, g) = \frac{c\left(\frac{c}{T} + \frac{c|g|}{2M}\right)^2}{\left(\frac{c}{T} + \frac{c|g|}{2M} - cx\right)^2} > 0, \]

and the fact that \( T(x, g) \) has no singularity in the domain \( \{(x, g), 0 < x < A, g > \bar{g}\} \). We have \( 0 < A_{N-1} < \bar{A} \), since \( A_{N-1} = T(0, \bar{g}_N) > 0 \) and \( \bar{A}_{N-1} = T(0, \bar{g}_N) < T(\bar{A}, \bar{g}) = \bar{A} \). Then we get by induction that \( 0 < \bar{A}_{n+1} < \bar{A}_n < \bar{A} \) implies that \( 0 < \bar{A}_n < A_{n-1} < \bar{A} \). Indeed, \( \bar{A}_{n-1} = T(\bar{A}_n, \bar{g}_n) < T(\bar{A}, \bar{g}) = \bar{A} \), and \( \bar{A}_{n-1} = T(\bar{A}_n, \bar{g}_n) > T(\bar{A}_{n+1}, \bar{g}_{n+1}) = \bar{A}_n \), as we have claimed.

The conditions \( N > N_0 \) and \( N > n^B \) with sufficiently large \( N_0 \) and \( B \) imply that \( |\bar{g}_t - \bar{g}| < 10^{-n} \) for all \( 0 < \ell \leq \sqrt{N} \), and \( A_9 - 5^{-n} < A^* < A \), where \( A^* \) is the smaller solution of the equation \( T(x, g^*) = x \) with \( g^* = \bar{g}\sqrt{N} \). Indeed, the last equation is a small perturbation of the equation \( T(x, \bar{g}) = x \), which has two solutions \( A_1 = \bar{A} \) and \( A_2 = \frac{1}{(2-c)T} > \bar{A} \). Hence the solutions of the equation \( T(x, g^*) = x \) are very close to \( A_1 \) and \( A_2 \). We claim that the monotonicity properties of the sequences \( \bar{g}_n \) and \( \bar{A}_n \) and the function \( T(x, g) \) imply that \( \bar{A} > A_n > T_{g^*}^{-N-n}(0) \), where \( T_{g^*}^k \) denotes the \( k \)-th iteration of the function \( T(x, g^*) \) with fixed \( g^* \) in the variable \( x \). Indeed, \( \bar{A} > A_n > \bar{A}_n \), \( 0 < \bar{A}_{N-n} < \bar{A} \), and we get by induction that for all \( \ell \geq 0 \)

\[ \bar{A}_{N-n-\ell} \geq T_{g^*}^\ell(0), \]

which implies the required statement with \( \ell = \sqrt{N} - n \).

To complete the proof of Lemma 1 it is enough to show that \( T_{g^*}^n(0) \) tends exponentially fast in \( n \) to the smaller solution \( A^* \) of the equation \( T_{g^*}(x) = x \). Since \( T(x, g^*) \) is a convex increasing function (in the variable \( x \)) it is enough to show that \( \frac{\partial T(x, g^*)}{\partial x} \leq \alpha < 1 \) for \( x = A^* \) if \( (A^*, g^*) \) is in a small neighbourhood of the point \( (\bar{A}, \bar{g}) \). But this follows from the continuity of the function \( \frac{\partial T(x, g)}{\partial x} \), and the fact that its value in the point \( (\bar{A}, \bar{g}) \) equals \( c^{-1} < 1 \). Lemma 1 is proved.

4. Some preparatory remarks to the proof of Proposition 1’. We shall prove the following estimates under the conditions of Proposition 1’.

Put
\[ S_n^i f(x) = \int_{\{y, x+y \in \Omega_n^i\}} \exp\left(\frac{c^n}{T} xy\right) f\left(\frac{x+y}{2}\right) p_n(y) dy, \quad i = 1, 2, 3. \]

Then we have with some appropriate \( \varepsilon’ = \varepsilon’(c), \varepsilon’ > 0 \), and the same \( q \) as in Proposition 1’:

In the domain \( x \in \Omega_n^1 \)
\[ S_n^1 f(x) = L_n \exp\left\{g_n(x(1) - M) + A_n(x(2))^2 + \varepsilon_n(x) + \hat{\varepsilon}_n(x)\right\} \]

with
\[ \sup_{x \in \Omega_n^1} |\varepsilon_n(x)| \leq q^{n+1}, \quad \sup_{x \in \Omega_n^1} |\varepsilon_n(x)| \leq Kc^{-0.2n}, \]

\[ \sup_{x \in \Omega_n^1} |\hat{\varepsilon}_n(x)| \leq q^{n+1}, \quad \sup_{x \in \Omega_n^1} |\hat{\varepsilon}_n(x)| \leq Kc^{-0.2n}. \]
where $K$ does not depend on $n$,

\begin{equation}
(4.1'') \quad L_n \geq e^{-n} \exp \left( \frac{e^n}{T} M^2 \right),
\end{equation}

\begin{equation}
(4.2) \quad S_n^2 f(x) \leq L_n \exp \left\{ g_n(x(1) - M) + A_n x(2)^2 - \varepsilon' c^{0.2n} \right\},
\end{equation}

\begin{equation}
(4.3) \quad S_n^3 f(x) \leq L_n \exp \left\{ g_n(x(1) - M) + A_n x(2)^2 - \frac{1}{6} c^{n/2} \right\}.
\end{equation}

In the domain $x \in \Omega_n^2$

\begin{equation}
(4.4) \quad S_n^1 f(x) \leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} + q^{n+1} + K c^{-0.2n} \right\},
\end{equation}

\begin{equation}
(4.5) \quad S_n^2 f(x) \leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} - \varepsilon' c^{0.2n} \right\},
\end{equation}

\begin{equation}
(4.6) \quad S_n^3 f(x) \leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} - \frac{1}{6} c^{n/2} \right\}.
\end{equation}

In the domain $x \in \Omega_n^3$

\begin{equation}
(4.7) \quad S_n f(x) \leq L_n \exp \left\{ \frac{g_n}{2M} (|x|^2 - M^2) \right\}.
\end{equation}

We show that these estimates imply Proposition 1’. Indeed, for $x \in \Omega_n^1$

\begin{equation*}
S_n f(x) = S_n^{(1)} f(x) + S_n^{(2)} f(x) + S_n^{(3)} f(x) = L_n \exp \left\{ g_n(x(1) - M) + A_n x(2)^2 + \varepsilon_n(x) \right\}
\end{equation*}

with

\begin{equation*}
\sup_{x \in \Omega_n^1} |\varepsilon_n(x)| \leq \sup_{x \in \Omega_n^1} |\varepsilon_1(x)| + \sup_{x \in \Omega_n^1} |\varepsilon_2(x)| + 2 \exp(-\varepsilon' c^{0.2n}) + 2 \exp \left( -\frac{1}{6} c^{n/2} \right).
\end{equation*}

(Here we have exploited that $e^t < 1 + 2|t|$ for small $t$.) Hence

\begin{equation*}
\sup_{x \in \Omega_n^1} |\varepsilon_n(x)| \leq q^{n+1} + K c^{-0.2n} + 2 \exp(-\varepsilon' c^{0.2n}) + 2 \exp \left( -\frac{1}{6} c^{n/2} \right) \leq q^n
\end{equation*}

if $c^{-0.2} < q < 1$, and $n \geq n_0(q, D, \varepsilon')$.

For $x \in \Omega_n^2$ we have analogously

\begin{equation*}
S_n f(x) = S_n^{(1)} f(x) + S_n^{(2)} f(x) + S_n^{(3)} f(x)
\end{equation*}

\begin{equation*}
\leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} + q^{n+1} + K c^{-0.2n} + 2 \exp(-\varepsilon' c^{0.2n}) + 2 \exp \left( -\frac{1}{6} c^{n/2} \right) \right\}
\end{equation*}

\begin{equation*}
\leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} + q^n \right\}.
\end{equation*}
as we have claimed in Proposition 1'. For \( x \in \Omega^3_n \) (4.7) contains the needed estimate. The above estimates will be proved in the next Section. In this Section we prove two lemmas which we need during the proof. Put

\[
S^n f(x) = \int_{R^2 - V^c_n(x)} \exp\left( \frac{c^n}{T} xy \right) f_n \left( \frac{x + y}{2} \right) p_n(y) \, dy
\]

with

\[
V^c_n(x) = \{ y \in R^2, \quad ||y| - M| \leq \epsilon c^{-0.4n}, \quad xy \geq |x||y| - \epsilon c^{-0.4n} \},
\]

where \( xy \) denotes scalar product.

**Lemma 2.** There is some \( \epsilon_0 = \epsilon_0(c) \) and \( n_0 = n_0(T, M, D, c, \epsilon) \) such that if \( n \geq n_0, \quad 0 < \epsilon < \epsilon_0, \quad 0 \leq g_{n+1} < D c^{n+1} \) and

\[
0 \leq f(x) \leq \exp \left\{ g_{n+1} \left( \frac{||x||^2 - M^2}{2M} \right) \right\} \quad \text{for all } x \in R^2
\]

then

a) \( 0 \leq S_n f(x) \leq c^n \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (||x||^2 - M^2) - \frac{c^n}{3T} (||x|| - M)^2 \right\}, \quad x \in R^2 \)

and

b) \( 0 \leq S^n f(x) \leq \exp \left\{ \frac{c^n}{T} M^2 + g_n (||x|| - M) - c^{n/2} \right\} \quad \text{if } ||x|| - M < c^{-0.4n} \)

with \( g_n = \frac{g_{n+1}}{2} + \frac{c^n}{T} M \).

**Lemma 3.** There is some \( n_0 = n_0(T, M, D, c) \) such that if \( n \geq n_0 \)

\[
0 \leq f(x) \leq \exp \left\{ g_{n+1} (||x|| - M) \right\} \quad \text{for } ||x|| - M < c^{-0.4(n+1)}
\]

\[
f(x) = 0 \quad \text{for } ||x|| - M \geq c^{-0.4(n+1)}
\]

then

\[
0 \leq S_n f(x) \leq K \exp \left\{ \frac{c^n}{T} M^2 + g_n (||x|| - M) \right\} \quad \text{for } ||x|| - M \leq c^{-0.4n}
\]

with

\[
g_n = \frac{g_{n+1}}{2} + \frac{c^n}{T} M
\]

and some \( K = K(T, M, D, c) > 0 \).

**Proof of Lemma 2.**

Part a). We have

\[
0 \leq S_n f(x) \leq \int \exp \left\{ \frac{c^n}{T} xy + \frac{g_{n+1}}{2M} \left( \left| \frac{x + y}{2} \right| - M \right)^2 \right\} p_n(y) \, dy.
\]
Clearly, \( \max_{|y|=r} xy \) is taken in the point \( y = \frac{x}{|x|} r \), and it equals \( |x|r \). Similarly \( \max_{|y|=r} (x+y)^2 = (|x| + r)^2 \). Hence
\[
(4.10) \quad 0 \leq S_n f(x) \leq 2\pi \int_0^\infty \exp \left(\frac{c^n}{T} |x|r + \frac{g_{n+1}}{2M} \left( \left| \frac{|x| + r}{2} \right|^2 - M^2 \right) \right) r\tilde{p}_n(r) \, dr.
\]

Let us split up the integral into two parts, \( \int_0^M \) and \( \int_M^\infty \). Put
\[
(4.11) \quad f_n(t) = c^{-n}\tilde{p}_n(M + c^{-n}t).
\]

It follows from (1.10) that
\[
(4.11') \quad f_n(t) \leq C_1 \exp(-C_2|t|) \quad \text{for } -c^n M < t < 0
\]
and from (1.8), (1.7) and (1.9) that
\[
(4.11'') \quad f_n(t) \leq C_3 \exp(-\ell^2) \quad \text{for } t > 0
\]
with some appropriate \( C_1 > 0, C_2 > 0 \) and \( C_3 > 0 \). (Relations (1.7) and (1.9) are needed in the domain \( 0 < |x| < \eta_{n,1/\alpha} e^{-n} \).)

First we estimate the integral \( \int_0^M \). For \( 0 \leq r \leq M \) we have
\[
\frac{c^n}{T} |x|r + \frac{g_{n+1}}{2M} \left( \left| \frac{|x| + r}{2} \right|^2 - M^2 \right) \leq \frac{c^n}{T} |x|M + \frac{g_{n+1}}{2M} \left( \left| \frac{|x| + M}{2} \right|^2 - M^2 \right).
\]

Hence (4.11) and (4.11') imply that
\[
\int_0^M \ldots dr \leq C_1 \exp \left\{ \frac{c^n}{T} |x|M + \frac{g_{n+1}}{2M} \left[ \left( \left| \frac{|x| + M}{2} \right|^2 - M^2 \right) \right] \right\} \\
\int_{-c^n M}^0 (M + c^{-n}t) e^{-C_2|t|} \, dt \leq C_4 M \exp \left\{ \frac{c^n}{T} |x|M + \frac{g_{n+1}}{2M} \left[ \left( \left| \frac{|x| + M}{2} \right|^2 - M^2 \right) \right] \right\}.
\]

Simple calculation yields the identity
\[
\frac{c^n}{T} |x|M + \frac{g_{n+1}}{2M} \left[ \left( \left| \frac{|x| + M}{2} \right|^2 - M^2 \right) \right] = \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) - \left( \frac{c^n}{2T} + \frac{g_{n+1}}{8M} \right) (|x| - M)^2.
\]

Then, since \( C_4 M < c^{n/2} \) we get that
\[
(4.12) \quad \int_0^M \ldots dr \leq c^{n/2} \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) + \left( \frac{c^n}{2T} + \frac{g_{n+1}}{8M} \right) (|x| - M)^2 \right\}.
\]
Let us estimate \( \int_M^\infty \ldots \, dr \). We make the change of variables \( r = M + c^{-n}t \), introduce \( f_n(t) = c^{-n} \tilde{p}_n(M + c^{-n}t) \), \( \tilde{g}_n = c^{-n} g_n \) and \( \tilde{g}_{n+1} = c^{-(n+1)} g_{n+1} \). Since

\[
\frac{c^n}{T} |x| r + \frac{g_{n+1}}{2M} \left( \frac{|x| + r}{2} \right)^2 - M^2
\]

\[
= \frac{c^n}{T} |x| M + \frac{g_{n+1}}{2M} \left( \frac{|x| + M}{2} \right)^2 - M^2
\]

(4.13)

\[
= \frac{c^n}{T} |x|(r - M) + \frac{g_{n+1}}{8M}(r - M)(2|x| + r + M)
\]

\[
= \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) - \left( \frac{c^n}{2T} + \frac{g_{n+1}}{8M} \right) (|x| - M)^2
\]

\[
+ \left( \frac{c^n}{T} + \frac{g_{n+1}}{4M} \right) |x|(r - M) + \frac{g_{n+1}}{4} (r - M) + \frac{g_{n+1}}{8M} (r - M)^2
\]

hence

(4.14)

\[
\int_M^\infty \ldots \, dr = \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) - \left( \frac{c^n}{2T} + \frac{g_{n+1}}{8M} \right) (|x| - M)^2 \right\} J_n(|x|)
\]

with

\[
J_n(|x|) = \int_0^\infty \exp \left\{ \left( \frac{1}{T} + \frac{c}{4M} \tilde{g}_{n+1} \right) |x| t
\]

\[
+ \frac{c}{4} \tilde{g}_{n+1} t + c \frac{c^{-n}}{8M} \tilde{g}_{n+1} t^2 \right\} (M + c^{-n} t) f_n(t) \, dt
\]

(4.14')

\[
= \int_0^\infty \exp \left\{ \left( \frac{1}{T} + \frac{c}{4M} \tilde{g}_{n+1} \right) (|x| - M) t
\]

\[
+ \left( \frac{M}{T} + \frac{c}{2} \tilde{g}_{n+1} \right) t + c \frac{c^{-n+1}}{8M} \tilde{g}_{n+1} t^2 \right\} (M + c^{-n} t) f_n(t) \, dt.
\]

Relation (4.11'') implies that

\[
J_n(|x|) \leq C_5 \int_0^\infty \exp \left\{ \left( \frac{1}{T} + \frac{c}{4M} \tilde{g}_{n+1} \right) (|x| - M) t - \frac{t^2}{2} \right\} \, dt
\]

\[
\leq C_6 \{ \exp C_7 (|x| - M)^2 \}.
\]

Thus

\[
\int_M^\infty \ldots \, dr \leq C_6 \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2)
\]

\[
- \left( \frac{c^n}{2T} + \frac{g_{n+1}}{8M} \right) (|x| - M)^2 + C_7 (|x| - M)^2 \right\}.
\]
Since \(\frac{c_n^2}{T^2} + \frac{g_{n+1}}{8M} - C_7 \geq \frac{c_n^2}{3T}\) and \(C_6 \leq c^{n/2}\) for large \(n\), hence
\[
\int_M^\infty \ldots dr \leq c^{n/2} \exp \left\{ \frac{c_n^2}{T}M^2 + \frac{g_n}{2M}(|x|^2 - M^2) - \frac{c_n^2}{3T}(|x| - M)^2 \right\}.
\]
This inequality together with (4.12) imply that
\[
(4.15)
\]
\[
S_n f(x) \leq 4\pi c^{n/2} \exp \left\{ \frac{c_n^2}{T}M^2 + \frac{g_n}{2M}(|x|^2 - M^2) - \frac{c_n^2}{3T}(|x| - M)^2 \right\} \leq \leq c^n \exp \left\{ \frac{c_n^2}{T}M^2 + \frac{g_n}{2M}(|x|^2 - M^2) - \frac{c_n^2}{3T}(|x| - M)^2 \right\}
\]
as we have claimed.

Part b). Let us introduce
\[
R_n^\varepsilon f(x) = \int_{\{y : ||y| - M| \geq \varepsilon c^{-0.4n}\}} \exp \left( \frac{c_n^2}{T}xy \right) f\left( \frac{x+y}{2} \right) p_n(y) dy
\]
and
\[
Q_n^\varepsilon f(x) = \int_{\{y : ||y| - M| < \varepsilon c^{-0.4n}, xy \leq |x| \varepsilon c^{-0.4n}\}} \ldots dy.
\]
Clearly, \(S_n^\varepsilon f(x) = R_n^\varepsilon f(x) + Q_n^\varepsilon f(x)\), and by (4.10)
\[
0 \leq R_n^\varepsilon f(x) \leq 2\pi \left[ \int_0^{M-\varepsilon c^{-0.4n}} + \int_{M+\varepsilon c^{-0.4n}}^\infty \right] \exp \left\{ \frac{c_n^2}{T}|x|r + \frac{g_{n+1}}{2M} \left( \left( \frac{\left| x \right| + r}{2} \right)^2 - M^2 \right) \right\} r\bar{p}_n(r) dr.
\]
Moreover, similarly to part a), we get by using (4.13) and the observation
\[
\left( \frac{c_n^2}{T} + \frac{g_{n+1}}{4M} \right) |x|(r - M) + \frac{g_{n+1}}{4}(r - M) + \frac{g_{n+1}}{8M}(r - M)^2
\]
\[
< \frac{g_{n+1}}{4}(r - M)\left( 1 + \frac{r - M}{2M} \right) < -2c^{n/2} \quad \text{if} \quad -M < r - M < -c^{-0.4n}
\]
that
\[
(4.16)
\]
\[
\int_0^{M-\varepsilon c^{-0.4n}} \ldots dr
\]
\[
\leq C_1 \exp \left\{ -2c^{n/2} + \frac{c_n^2}{T}M^2 + \frac{g_n}{2M}(|x|^2 - M^2) - \left( \frac{c_n^2}{2T} + \frac{g_{n+1}}{8M} \right)(|x| - M)^2 \right\}
\]
\[
\int_0^{M-\varepsilon c^{-0.4n}} r\bar{p}_n(r) dr
\]
\[
\leq c^{-n/2} \exp \left\{ -2c^{n/2} + \frac{c_n^2}{T}M^2 + \frac{g_n}{2M}(|x|^2 - M^2)
\]
\[
- \left( \frac{c_n^2}{2T} + \frac{g_{n+1}}{8M} \right)(|x| - M)^2 \right\}
\]
\[
\leq c^{-n/2} \exp \left\{ -2c^{n/2} + \frac{c_n^2}{T}M^2 + \frac{g_n}{2M}(|x|^2 - M^2) \right\},
\]

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and by (4.13) and (4.11’), similarly to (4.14), (4.14’)
\[ \int_{M + \varepsilon c^{-0.4n}}^{\infty} \cdots dr \leq \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) - \left( \frac{c^n}{2T} + \frac{g_{n+1}}{8M} \right) (|x| - M)^2 \right\} J_n^\varepsilon(|x|) \]
with
\[ J_n^\varepsilon(|x|) = C_5 \int_{\varepsilon c^{0.6n}}^{\infty} c^n \exp \left\{ \left( \frac{1}{T} + \frac{c}{4M} \bar{g}_{n+1} \right) (|x| - M)t + \left( \frac{M}{T} + \frac{c}{2} \bar{g}_{n+1} \right) t - \frac{t^2}{2} \right\} dt. \]
Since $|x| - M \leq c^{-0.4n}$, and $\bar{g}_{n+1} < D$
\[ \sup_{t \geq 0} \left\{ \left( \frac{1}{T} + \frac{c}{4M} \bar{g}_{n+1} \right) (|x| - M)t + \left( \frac{M}{T} + \frac{c}{2} \bar{g}_{n+1} \right) t - \frac{t^2}{3} \right\} \leq C', \]
and therefore
\[ J_n^\varepsilon(|x|) \leq C_6 \int_{\varepsilon c^{0.6n}}^{\infty} c^n \exp \left( -\frac{t^2}{3} \right) dt \leq \exp(-\varepsilon' c^{1.2n}) \leq c^{-n} \exp(-2c^{n/2}) \]
with some $\varepsilon' = \varepsilon'(\varepsilon) > 0$. Hence
\[ \int_{M + \varepsilon c^{-0.4n}}^{\infty} \cdots dr \leq c^{-n} \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) - 2c^{n/2} \right\}. \]
The last inequality together with (4.16) imply that
\[ (4.17) \quad R_n^\varepsilon f(x) \leq 4\pi c^{-n/2} \exp \left\{ -2c^{n/2} + \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) \right\}. \]
Now we estimate $Q_n^\varepsilon f(x)$. We have
\[ 0 \leq Q_n^\varepsilon f(x) \leq \int_{\{ y, |y| - M \leq \varepsilon c^{-0.4n}, \ x y \leq |x| |y| - \varepsilon c^{-0.4n} \}} \exp \left\{ \frac{c^n}{T} x y + \frac{g_{n+1}}{2M} \left( \frac{x + y}{2} \right)^2 - M^2 \right\} p_n(y) dy. \]
Since $|y| = r$, $x y \leq r |x| - \varepsilon c^{-0.4n}$ and $(\frac{x+y}{2})^2 \leq (\frac{|x| + r}{2})^2$ in the last integral, hence
\[ Q_n^\varepsilon f(x) \leq 2\pi \int_{0}^{\infty} \exp \left\{ \frac{c^n}{T} (|x| - r - \varepsilon c^{-0.4n}) + \frac{g_{n+1}}{2M} \left( \frac{|x| + r}{2} \right)^2 - M^2 \right\} r \tilde{p}_n(r) dr \]
\[ \leq 2\pi \exp \left( -\frac{\varepsilon c^{0.6n}}{T} \right) \int_{0}^{\infty} \exp \left\{ \frac{c^n}{T} |x| r + \frac{g_{n+1}}{2M} \left( \frac{|x| + r}{2} \right)^2 - M^2 \right\} r \tilde{p}_n(r) dr. \]
The last integral has already appeared in (4.10). We have estimated it in Part a), and bounded it by the right hand side of (4.15). Hence
\[ Q_n^\varepsilon f(x) \leq 2\pi c^n \exp \left\{ -\frac{\varepsilon c^{0.6n}}{T} + \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) \right\} \]
\[ \leq \frac{1}{2} \exp \left\{ -2c^{n/2} + \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) \right\}. \]
This inequality together with (4.17) imply that

\[ S_n f(x) \leq \exp \left\{ -2c^n/2 + \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) \right\}. \]

Since \( \frac{1}{2M} (|x|^2 - M^2) = (|x| - M) + \frac{1}{2M} (|x| - M)^2 \) and \( \frac{g_n}{2M} (|x| - M)^2 \leq Dc^{0.2n} \) if \( |x| - M \leq c^{-0.4n} \), hence the last inequality implies that under the conditions of Part b)

\[ S_n f(x) \leq \exp \left\{ -2c^n/2 + \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x| - M) + Kc^{0.2n} \right\}, \]

as we have claimed.

**Proof of Lemma 3.**

\[
0 \leq S_n f(x) \leq \int_{\{y, \ |y + x| - M| \leq c^{-0.4(n+1)}\}} \exp \left\{ \frac{c^n}{T} xy + g_{n+1} \left( \frac{|x + y|}{2} - M \right) \right\} p_n(y) \, dy.
\]

Since

\[
\max_{|y|=r} \left\{ \frac{c^n}{T} xy + g_{n+1} \left( \frac{|x + y|}{2} - M \right) \right\} \leq \frac{c^n}{T} r + g_{n+1} \left[ \left( \frac{|x| + r}{2} \right) - M \right],
\]

hence

\[
S_n f(x) \leq 2\pi \int_0^\infty \exp \left\{ \frac{c^n}{T} r + g_{n+1} \left( \frac{|x| + r}{2} - M \right) \right\} r \tilde{p}_n(r) \, dr.
\]

Writing \( \frac{c^n}{T} r = \frac{c^n}{T} M^2 + \frac{c^n}{T} M(|x| - M + r - M) + \frac{c^n}{T} (|x| - M)(r - M) \) we get that

\[
S_n f(x) \leq 2\pi \exp \left\{ \frac{c^n}{T} M^2 + g_n(|x| - M) \right\} \int_0^\infty \exp \left\{ g_n(r - M) + \frac{c^n}{T} (|x| - M)(r - M) \right\} r \tilde{p}_n(r) \, dr.
\]

The change of variables \( r = M + c^{-n}t \) and the introduction of \( \tilde{g}_n = c^{-n}g_n \) yields that

\[
S_n f(x) \leq 2\pi \exp \left\{ \frac{c^n}{T} M^2 + g_n(|x| - M) \right\} \int_{-c^n M}^\infty \exp \left\{ \tilde{g}_n t + \frac{(|x| - M) t}{T} \right\} (M + c^{-n}t) f_n(t) \, dt.
\]
Since \(0 < \bar{g}_n < D\), and \(||x| - M| < e^{-0.4n}\) relations (4.11') and (4.11'') imply that for large \(n\)

\[
\int_{-c^n M}^{\infty} \exp \left\{ \bar{g}_n t + \frac{1}{T}(|x| - M)t \right\} (M + e^{-n}t) f_n(t) \, dt < \bar{K}
\]

with some \(\bar{K} > 0\) independent of \(n\). Hence

\[
S_n f(x) \leq 2 \pi \bar{K} \exp \left\{ \frac{c^n}{T} M^2 + g_n(|x| - M) \right\}
\]

\[
\leq K \exp \left\{ \frac{c^n}{T} M^2 + g_n(|x| - M) \right\},
\]

as we have claimed.

5. The proof of Proposition 1'. In this Section we prove the estimates (4.1)–(4.7) which imply Proposition 1'.

a) The estimation of \(S_n^1 f(x)\) for \(x \in \Omega_n^1\).

It follows from (2.15) and (2.15') that

\[
S_n^1 f(x) = \int_{\{y, \frac{x+y}{2} \in \Omega_{n+1}^1\}} \exp \left\{ \frac{c^n}{T} xy + g_{n+1} \left( \frac{x^{(1)} + y^{(1)}}{2} - M \right) + A_{n+1} \right. \\
\left. \left( \frac{x^{(2)} + y^{(2)}}{2} \right)^2 + \varepsilon_{n+1} \left( \frac{x + y}{2} \right) \right\} p_n(y) \, dy.
\]

Hence

\[
S_n^1 f(x) = \exp(\varepsilon_n(x)) \int_{\{y, \frac{x+y}{2} \in \Omega_{n+1}^1\}} \exp \left\{ \frac{c^n}{T} xy + g_{n+1} \left( \frac{x^{(1)} + y^{(1)}}{2} - M \right) \\
+ A_{n+1} \left( \frac{x^{(2)} + y^{(2)}}{2} \right)^2 \right\} p_n(y) \, dy.
\]

with some

\[
(5.1) \quad \sup_{x \in \Omega_{n+1}^1} |\varepsilon_n(x)| \leq q^{n+1}.
\]

Let us rewrite the last expression in polar coordinate system. We get that

\[
(5.2) \quad S_n^1 f(x) = \exp(\varepsilon_n(x)) \int_0^{\infty} I_n(r) \overline{p_n}(r) \, dr
\]

with

\[
I_n(r) = \int_{\{\varphi \in \Gamma_n(r,x)\}} r \exp \left\{ \frac{c^n}{T} \left( x^{(1)} y^{(1)} + x^{(2)} y^{(2)} \right) + g_{n+1} \left( \frac{x^{(1)} + y^{(1)}}{2} - M \right) \\
+ A_{n+1} \left( \frac{x^{(2)} + y^{(2)}}{2} \right)^2 \right\} d\varphi
\]
where \( y^{(1)} = r \cos \varphi \), \( y^{(2)} = r \sin \varphi \), \(-\pi < \varphi < \pi\), \(y = (y^{(1)}, y^{(2)})\), and \(\Gamma_n(r, x) = \{\varphi, \frac{x+y}{2} \in \Omega_{n+1}^1\}\). We shall express \(I_n(r)\) as an asymptotically Gaussian integral with respect to \(\varphi\). For this aim we give some bounds on \(x^{(1)}\), \(x^{(2)}\), \(y^{(1)}\) and \(y^{(2)}\) if \(x \in \Omega_{n+1}^1\) and \(\frac{x+y}{2} \in \Omega_{n+1}^1\).

We have
\[
|x^{(2)}| < c^{-0.4n} \quad \text{if} \quad x \in \Omega_{n+1}^1.
\]

The second relation in (5.3) holds, since
\[
|x^{(1)} - M| < \frac{3}{2} c^{-0.4n}
\]
if \(x \in \Omega_{n+1}^1\), hence
\[
|x^{(1)} - M| + \left| \frac{x^{(2)}}{2|x|} + O(x^{(2)^4}) - M \right| \leq c^{-0.4n} + Kc^{-0.8n} \leq \frac{3}{2} c^{-0.4n}.
\]

Similarly, if \(\frac{x+y}{2} \in \Omega_{n+1}^1\) then \(\left| \frac{x^{(2)}+y^{(2)}}{2} \right| < c^{-0.4n}\), and \(\left| \frac{x^{(1)}+y^{(1)}}{2} - M \right| < 2c^{-0.4n}\).

Hence
\[
|y^{(2)}| \leq |x^{(2)}| + |x^{(2)} + y^{(2)}| \leq 3c^{-0.4n},
\]
and
\[
|y^{(1)} - M| = |M - x^{(1)}| + |x^{(1)} + y^{(1)} - 2M| \leq 6c^{-0.4n}
\]
and
\[
|r - M| = ||y| - M| = \left| (y^{(1)^2} + y^{(2)^2})^{1/2} - M \right| = \left| y^{(1)} + y^{(2)^2} + O(y^{(2)^4}) - M \right| \leq 10c^{-0.4n}
\]
if \(x \in \Omega_{n+1}^1\) and \(\frac{x+y}{2} \in \Omega_{n+1}^1\). In particular, (5.4′) implies that
\[
I_n(r) = 0 \quad \text{if} \quad |r - M| > 10c^{-0.4n}, \quad \text{and} \quad x \in \Omega_{n+1}^1.
\]

Furthermore, \(|\varphi| \leq 2|\sin \varphi| = 2 |y^{(2)}| = O(c^{-0.4n})\) and \(y^{(1)} = r(1 - \cos \varphi) = r(1 - \varphi^2/2 + O(c^{-1.6n}))\), \(y^{(2)} = r \sin \varphi = r \varphi + O(c^{-1.2n})\) and \(r = M + O(c^{-0.4n})\) if \(\varphi \in \Gamma_n(r, x)\). Hence
\[
I_n(r) = \left(1 + O(c^{-0.2n})\right) \int_{\{\varphi \in \Gamma_n(r, x)\}} M \exp \left\{ \frac{c^n}{T} \left( x^{(1)} + x^{(2)}\varphi \right) \left( 1 - \frac{\varphi^2}{2} \right) + x^{(2)}\varphi \right\} d\varphi
\]
\[
= \left(1 + O(c^{-0.2n})\right) M \exp \left\{ \frac{c^n}{T} x^{(1)} + \frac{g_{n+1}}{2} \left( x^{(1)} + \frac{x^{(2)}}{2} - M \right) + \frac{A_{n+1}}{4} x^{(2)^2} \right\}
\]
\[
\int_{\{\varphi \in \Gamma_n(r, x)\}} \exp \left\{ - \left( \frac{c^n}{2T} x^{(1)} + \frac{g_{n+1}}{4} r \frac{A_{n+1}}{4} x^{(2)^2} \right) - r \varphi^2 \right\} d\varphi.
\]
Moreover, since \( c^n x^{(1)} r \varphi^2 = c^n M^2 \varphi^2 + O(c^{-0.2n}) \), \( g_{n+1} r \varphi^2 = g_{n+1} M \varphi^2 + O(c^{-0.2n}) \), \( A_{n+1} r^2 \varphi^2 = A_{n+1} M^2 \varphi^2 + O(c^{-0.2n}) \) and \( \left( \frac{c^n}{T} \right) x^{(2)} r \varphi = \left( \frac{c^n}{T} + \frac{A_{n+1}}{4} \right) x^{(2)} M \varphi + O(c^{-0.2n}) \) in our case (observe that \( g_{n+1} \leq Dc^{n+1} \) and \( A_{n+1} \leq \bar{A}c^{n+1} \)), hence we make an error of order \( O(c^{-0.2n}) \) by substituting \( x^{(1)} \) and \( r \) by \( M \) in the last integral, i.e. 

\[
I_n(r) = \left( 1 + O(c^{-0.2n}) \right) M \exp \left\{ \frac{c^n}{T} r x^{(1)} + \frac{g_{n+1}}{2} \left( \frac{x^{(1)} + r}{2} - M \right) + \frac{A_{n+1}}{4} x^{(2)} \right\} \\
\int_{\{ \varphi \in \Gamma_n(r,x) \}} \exp \left\{ \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right) M^2 \varphi^2 + \left( \frac{c^n}{T} + \frac{A_{n+1}}{2} \right) M x^{(2)} \varphi \right\} d\varphi ,
\]

or equivalently 

\[
I_n(r) = \left( 1 + O(c^{-0.2n}) \right) \exp \left\{ \frac{c^n}{T} r x^{(1)} + \frac{g_{n+1}}{2} \left( \frac{x^{(1)} + r}{2} - M \right) \right. \\
+ \frac{A_{n+1}}{4} x^{(2)} \right\} + \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right) \\
\int_{\{ \varphi \in \Gamma_n(r,x) \}} M \exp \left\{ - M^2 \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right) (\varphi - \gamma_n x^{(2)})^2 \right\} d\varphi ,
\]

with 

\[
\gamma_n = \frac{\left( \frac{c^n}{T} + \frac{A_{n+1}}{2} \right)}{\left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)} .
\]

By (2.5') we get from this relation that 

\[
(5.6)
\]

\[
I_n(r) = \left( 1 + O(c^{-0.2n}) \right) \exp \left\{ \frac{c^n}{T} r x^{(1)} + \frac{g_{n+1}}{2} \left( \frac{x^{(1)} + r}{2} - M \right) + A_n x^{(2)} \right\} \\
\int_{\{ \varphi \in \Gamma_n(r,x) \}} M \exp \left\{ - M^2 \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right) (\varphi - \gamma_n x^{(2)})^2 \right\} d\varphi .
\]

Since 

\[
(5.7) \quad \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right) \geq c^n \left( \frac{1}{2T} + c \frac{\bar{g}}{4M} - c \frac{\bar{A}}{4} \right) = Ke^n
\]

with \( K = \frac{1}{2T} + c \frac{\bar{g}}{4M} - c \frac{\bar{A}}{4} = \frac{c(4-c)}{4(2-c)T} > 0 \) relation (5.6) implies that 

\[
(5.8) \quad I_n(r) \leq \sqrt{\pi} \left( 1 + O(c^{-0.2n}) \right) \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)^{-1/2} \\
\exp \left\{ \frac{c^n}{T} r x^{(1)} + \frac{g_{n+1}}{2} \left( \frac{x^{(1)} + r}{2} - M \right) + A_n x^{(2)} \right\} 
\]
Moreover, we claim that
\begin{equation}
I_n(r) = (1 + O(c^{-0.2n})) \sqrt{\pi} \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)^{-1/2} \exp \left\{ \frac{c^n}{T} r x^{(1)} + \frac{g_{n+1}}{2} \left( \frac{x^{(1)} + r}{2} - M \right) + A_n x^{(2)} \right\}
\end{equation}
if $|r - M| < \varepsilon c^{-0.4n}$
with some appropriate $\varepsilon > 0$. Because of (5.6) and (5.7) to prove (5.9) it is enough to show that there is some $\varepsilon = \varepsilon_0(\varepsilon) > 0$ such that
\begin{equation}
\{ \varphi : |\varphi - \gamma_n x^{(2)}| < \varepsilon_0 c^{-0.4n} \} \subset \Gamma_n(x, r) \quad \text{if } |r - M| < \varepsilon c^{-0.4n}.
\end{equation}
Since $A_{n+1} \leq \tilde{A} c^{n+1}$ and $g_{n+1} \geq \tilde{g} c^{n+1}$
\begin{equation}
\gamma_n = \frac{\left( \frac{c^n}{T} + \frac{A_{n+1}}{2} \right)}{2 \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)^{1/2}} \leq \frac{1}{2} \frac{1 + \frac{2c}{2 - c} - \frac{2c}{2 - c} M}{(1 + \frac{c}{2 - c} - \frac{2c}{2 - c} M) M} = \frac{2 - c}{cM},
\end{equation}
and we prove (5.10) by showing that
\begin{equation}
(5.10') \quad \left\{ \frac{x^{(1)} + r \cos \varphi}{2}, \frac{x^{(2)} + r \sin \varphi}{2} \right\} \in \Omega^{1}_{n+1}
\end{equation}
if $|\varphi - \gamma_n x^{(2)}| < \varepsilon_0 c^{-0.4n}$, $|r - M| \leq \varepsilon c^{-0.4n}$ and $x \in \Omega^{1}_{n}$. But in this case
\begin{equation}
\left| \frac{x^{(2)} + r \sin \varphi}{2} \right| \leq \frac{1}{2} \left( |x^{(2)}| + r |\varphi| \right) \leq \frac{1}{2} \left( |x^{(2)}| + r (\gamma_n |x^{(2)}| + |\varphi - \gamma_n x^{(2)}|) \right) \leq \frac{1}{2} \left( c^{-0.4n} + (M + \varepsilon c^{-0.4n}) \left( \frac{2 - c}{cM} c^{-0.4n} + \varepsilon_0 c^{-0.4n} \right) \right) \leq \frac{1}{\sqrt{c}} c^{-0.4n} \leq c^{-0.4(n+1)}
\end{equation}
if $\varepsilon_0 > 0$ and $\varepsilon > 0$ are sufficiently small. We also get with the help of (5.3') that
\begin{equation}
\left| \frac{x^{(1)} + r \cos \varphi - M}{2} \right| \leq \left| \frac{x^{(1)} - M}{2} \right| + r \left| \cos \varphi - 1 \right| + \left| \frac{r - M}{2} \right| \leq \left| \frac{x^{(1)} - M}{2} \right| + r \left| \gamma_n |x^{(2)}| + \varepsilon_0 c^{-0.4n} \right| + \left| \frac{r - M}{2} \right| \leq \left( \frac{3}{4} + \varepsilon \right) c^{-0.4n} + K c^{-0.8n} \leq c^{-0.1} c^{-0.4(n+1)}
\end{equation}
if $\varepsilon_0 > 0$ and $\varepsilon > 0$ are sufficiently small. The above estimates imply (5.10') hence also (5.10). Now we can estimate the term $\int_0^{\infty} I_n(r) \bar{p}_n(r) \; dr$. Relation (5.9) yields that
\begin{equation}
\int_{|r - M| < \varepsilon c^{-0.4n}} I_n(r) \bar{p}_n(r) \; dr = (1 + O(c^{-0.2n})) \sqrt{\pi} \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)^{-1/2} \exp \left\{ \frac{c^n}{T} x^{(1)} M + \frac{g_{n+1}}{4} \left( x^{(1)} - M \right) + A_n x^{(2)} \right\}
\end{equation}
The large-scale limit of Dyson’s hierarchical model

with

\[ J_{n,\varepsilon} = J_{n,\varepsilon}(x^{(1)}) = \int_{|r-M|<\varepsilon c^{-0.4n}} \exp \left\{ \frac{c^n}{T} x^{(1)}(r-M) + \frac{g_{n+1}}{4} (r-M) \right\} \bar{p}_n(r) \, dr \]

\[ = \int_{|t|<\varepsilon c^{-0.6n}} \exp \left\{ \frac{tx^{(1)}}{T} + c \frac{g_{n+1}}{4} t \right\} f_n(t) \, dt \]

with the function \( f \) defined in (4.11). On the other hand, by (5.8) and (5.5)

\[ \int_{|r-M|>\varepsilon c^{-0.4n}} I_n(r) \bar{p}_n(r) \, dr \]

\[ \leq (1+O(e^{-0.2n})) \sqrt{\pi} \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)^{-1/2} \exp \left\{ \frac{c^n}{T} x^{(1)} M + \frac{g_{n+1}}{4} (x^{(1)} - M) + A_n x^{(2)} \right\} \bar{J}_{n,\varepsilon} \]

with

\[ \bar{J}_{n,\varepsilon} = \int_{10c^{0.6n}>|t|>\varepsilon c^{0.6n}} \exp \left\{ \frac{tx^{(1)}}{T} + c \frac{g_{n+1}}{4} t \right\} f_n(t) \, dt . \]

Let us remark that \( J_{n,\varepsilon} = J_{n,\varepsilon}(x^{(1)}) \) depends on \( x^{(1)} \). We show that this dependence is very weak. Namely, since

\[ \left| \frac{d}{dx^{(1)}} J_{n,\varepsilon}(x^{(1)}) \right| = \int_{|t|<\varepsilon c^{0.6n}} \exp \left\{ \frac{tx^{(1)}}{T} + c \frac{g_{n+1}}{4} t \right\} \frac{|t|}{T} f_n(t) \, dt \leq C < \infty \]

by (4.11’) and (4.11’’), and for \( x^{(1)} = M \) the expression

\[ J_{n,\varepsilon}(M) = \int_{|t|<\varepsilon c^{0.6n}} \exp \left\{ \frac{tM}{T} + c \frac{g_{n+1}}{4} t \right\} f_n(t) \, dt \]

satisfies the relation

\[ (5.11) \quad 0 < K_1 < J_{n,\varepsilon}(M) < K_2 < \infty \]

because of (4.11’), (4.11’’), the inequality \( 0 \leq \bar{g}_{n+1} \leq D \) and the relation \( f_n(t) \geq \text{const.} > 0 \) for \(|t| < 1\) that follows from Theorem A. Hence

\[ J_{n,\varepsilon}(x^{(1)}) = (1+O(e^{-0.4n})) J_{n,\varepsilon}(M) \quad \text{if} \quad x \in \Omega^1_n. \]

Similarly,

\[ 0 \leq \bar{J}_{n,\varepsilon} = O \left( \exp(-Kc^{0.6n}) \right) . \]

The above relations imply that

\[ \int I_n(r) \bar{p}_n(r) \, dr = (1+O(e^{-0.2n})) \sqrt{\pi} \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)^{-1/2} J_{n,\varepsilon}(M) \]

\[ \exp \left\{ \frac{c^n}{T} M^2 + g_n (x^{(1)} - M) + A_n x^{(2)} \right\} . \]

The last formula together with (5.1) and (5.2) imply (4.1) with

\[ L_n = \sqrt{\pi} \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right)^{-1/2} J_{n,\varepsilon}(M) \exp \left( \frac{c^n}{T} M^2 \right) . \]

Since \( \left( \frac{c^n}{2T} + \frac{g_{n+1}}{4M} - \frac{A_{n+1}}{4} \right) < \text{const.} e^n \) relation (5.11) and the last formula imply (4.1’’).
b) The estimation of $S_n^1 f(x)$ for $x \in \Omega_n^2$.

We divide $\Omega_n^2$ to two subsets $\tilde{\Omega}_n^2$ and $\tilde{\Omega}_n^2$, where we apply different arguments. Put

$$\Omega_{n, \varepsilon}^1 = \left\{ x, \left| |x| - M \right| \leq c^{-0.4n}, |x^{(2)}| \leq (1 + \varepsilon)c^{-0.4n}, x^{(1)} > 0 \right\}, \varepsilon > 0,$$

$$\tilde{\Omega}_n^2 = \Omega_{n, \varepsilon}^1 - \Omega_n^1,$$

and

$$\tilde{\Omega}_n^2 = \left\{ x, \left| |x| - M \right| \leq c^{-0.4n} \right\} - \Omega_{n, \varepsilon}^1.$$

Clearly, $\Omega_n^2 = \tilde{\Omega}_n^2 \cup \tilde{\Omega}_n^2$. The domain $\Omega_{n, \varepsilon}^1$ is a slight enlargement of $\Omega_n^1$. It is not difficult to see by analyzing the proof of relation (4.1) that for sufficiently small $\varepsilon > 0$

$$S_n^1 f(x) = L_n \exp \left\{ g_n(x^{(1)} - M) + A_n x^{(2)^2} + \tilde{\varepsilon}(x) + \tilde{\varepsilon}(x) \right\}$$

if $x \in \Omega_{n, \varepsilon}^1$ with some $|\tilde{\varepsilon}_n(x)| < q^{n+1}$ and $\tilde{\varepsilon}_n(x) < Kc^{-0.2n}$. Since $x^{(1)} - M = |x| - M - x^{(2)^2}/2M + O(c^{-1.2n})$ and $|x^{(2)}| \geq c^{-0.4n}$ for $x \in \tilde{\Omega}_n^2$, the above relation implies that

$$S_n^1 f(x) = L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) x^{(2)^2} \right.$$  

$$\left. + \tilde{\varepsilon}_n(x) + \tilde{\varepsilon}_n(x) + O(c^{-0.2n}) \right\}$$

$$\leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} + q^{n+1} + Kc^{-0.2n} \right\}$$

in this case, what we had to show. For $x \in \tilde{\Omega}_n^2$ we define

$$S_n^\varepsilon f(x) = \int_{R^2 - V_n^\varepsilon(x)} \exp \left( \frac{c^n}{T} xy \right) f \left( \frac{x + y}{2} \right) p_n(y) dy$$

and

$$T_n^\varepsilon f(x) = \int_{\{ y, \frac{x+y}{2} \in \Omega_{n+1}^1, y \in V_n^\varepsilon(x) \}} \exp \left( \frac{c^n}{T} xy \right) f \left( \frac{x + y}{2} \right) p_n(y) dy,$$

where $V_n^\varepsilon(x)$ is defined in (4.8'), and $\varepsilon > 0$ is appropriately chosen. The function $S_n^\varepsilon f(x)$ will be bounded with the help of Part b) of Lemma 2, and $T_n^\varepsilon f(x)$ similarly to $S_n^1 f(x)$ in the case $x \in \Omega_n^1$. To apply Lemma 2 first we show that under the conditions of Proposition 1'

$$0 \leq f(x) \leq 2 \exp \left\{ \frac{g_n+1}{2M} \left( |x|^2 - M^2 \right) \right\} \text{ for all } x \in R^2.$$

For $x \in \Omega_{n+1}^1$

$$x^{(1)} - M = |x| - M - \frac{x^{(2)^2}}{2|x|} + O \left( x^{(2)^4} \right) = |x| - \frac{x^{(2)^2}}{2M} - M + O \left( c^{-1.2n} \right),$$
hence

\[ 0 \leq f(x) \leq \exp \left\{ g_{n+1}(|x| - M) - \left( \frac{g_{n+1}}{2M} - A_{n+1} \right)x^{(2)} + \varepsilon_{n+1}(x) + O(c^{-0.2n}) \right\}, \]

and since \( \frac{g_{n+1}}{2M} - A_{n+1} > 0 \) hence

\[ 0 \leq f(x) \leq \frac{3}{2} \exp\{g_{n+1}(|x| - M)\}. \]

This inequality together with the relation

\[ |x| - M = \frac{1}{2M} \left( (|x|^2 - M^2) - (|x| - M)^2 \right) \leq \frac{1}{2M} (|x|^2 - M^2) \]

imply (5.12) for \( x \in \Omega_{n+1}^1 \). Similarly, for \( x \in \Omega_{n+1}^2 \) the relation

\[
0 \leq f(x) \leq \exp \left\{ g_{n+1}(|x| - M) - \left( \frac{g_{n+1}}{2M} - A_{n+1} \right)c^{-0.8(n+1)} + q^{n+1} \right\} \\
\leq \frac{3}{2} \exp\{g_n(|x| - M)\}
\]

implies (5.12), and this relation also holds for \( x \in \Omega_{n+1}^3 \) by relation (2.17).

By (5.12) part b) of Lemma 2 can be applied for \( \frac{1}{2} f(x) \), and it yields that

\[
S_n^\varepsilon f(x) \leq 2 \exp \left\{ \frac{c^n}{T} M^2 + g_n(|x| - M) - c^n/2 \right\} \quad \text{if } x \in \Omega_n^2 \quad \text{(or } x \in \Omega_n^1 \text{)}.
\]

Since \( L_n \geq c^{-n} \exp\{c^n M^2\} \), and \( (\frac{g_n}{2M} - A_n)c^{-0.8n} = O(c^{0.2n}) \leq \frac{1}{3}c^{n/2} \), the last inequality implies that

\[
(5.13) \quad S_n^\varepsilon f(x) \leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right)c^{-0.8n} - \frac{1}{2}c^{n/2} \right\} \\
\text{if } x \in \Omega_n^2 \quad \text{(or } x \in \Omega_n^1 \text{)}.
\]

To estimate \( T_n^\varepsilon f(x) \) first we show that if \( x \in \Omega_n^2 \), \( y \in \mathcal{V}_n^\varepsilon (x) \) and \( \frac{x+y}{2} \in \Omega_{n+1}^1 \) then

\[
(5.14) \quad |x^{(2)}| < 2\sqrt{\varepsilon}c^{-0.2n}, \quad |y^{(2)}| < 2\sqrt{\varepsilon}c^{-0.2n}, \quad |x^{(1)} - M| \leq 2\sqrt{\varepsilon}c^{-0.2n}, \\
|y^{(1)} - M| \leq 2\sqrt{\varepsilon}c^{-0.2n}, \quad |x| - M| \leq \varepsilon c^{-0.4n}.
\]

Indeed, in this case \((x, y) \geq |x| |y| - \varepsilon c^{-0.4n}\), and

\[
0 \leq |x - y|^2 = |x|^2 + |y|^2 - 2(x, y) \leq |x|^2 + |y|^2 - 2|x||y| + \varepsilon c^{-0.4n} \\
= (|x| - |y|)^2 + \varepsilon c^{-0.4n} \leq 2(|x| - M)^2 + 2(|y| - M)^2 + \varepsilon c^{-0.4n} \\
\leq 2(1 + \varepsilon)c^{-0.8n} + \varepsilon c^{-0.4n} \leq 2\varepsilon c^{-0.4n}.
\]
Hence \( \left| \frac{x^{(2)} - y^{(2)}}{2} \right| \leq \sqrt{\varepsilon} c^{-0.2n} \) and \( \frac{1}{2} \left| (x^{(1)} - M) - (y^{(1)} - M) \right| \leq \sqrt{\varepsilon} c^{-0.2n} \). Since 
\[
\frac{x+y}{2} \in \Omega_{n+1}^1, \quad \left| \frac{x^{(2)}+y^{(2)}}{2} \right| \leq c^{-0.4(n+1)} \leq \sqrt{\varepsilon} c^{-0.2n} \text{ and } \frac{1}{2} \left| (x^{(1)} - M) + (y^{(1)} - M) \right| \leq \frac{3}{2} c^{-0.4(n+1)} \leq \sqrt{\varepsilon} c^{-0.2n} \text{ by (5.3).}
\]
These relations together with the definition of \( V^\varepsilon_n(x) \) imply (5.14) that enables us to estimate \( T_n^\varepsilon f(x) \) similarly to \( S_n^\varepsilon f(x) \) for \( x \in \Omega_{n+1}^1 \).

We get that

\[
T_n^\varepsilon f(x) \leq (1 + q^{n+1}) \int_0^\infty I_n(r)\bar{p}_n(r) \, dr \tag{5.15}
\]

with

\[
I_n(r) = \int_{\{\varphi \in \Gamma_n(r, x)\}} \left[ r \exp \left\{ \frac{c^n}{T} \left( x^{(1)} y^{(1)} + x^{(2)} y^{(2)} \right) \right. \right. \\
+ g_{n+1} \left( \frac{x^{(1)} + y^{(1)}}{2} - M \right) A_{n+1} \left( \frac{x^{(2)} + y^{(2)}}{2} \right)^2 \bigg\} \, d\varphi,
\tag{5.16}
\]

where \( y^{(1)} = r \cos \varphi, \ y^{(2)} = r \sin \varphi, \ y = (y^{(1)}, y^{(2)}) \) and

\[
\Gamma_n(r, x) = \{ \varphi : \frac{x+y}{2} \in \Omega_{n+1}^1, y \in V^\varepsilon_n(x) \}.
\]

Observe that by (5.14) \( |r - M| < \varepsilon c^{-0.4n}, \ |\varphi| \leq 2 |\sin \varphi| = \frac{2}{\varepsilon} |y^{(2)}| \leq \frac{2}{\varepsilon} \sqrt{\varepsilon} c^{-0.2n} \). Let us make the change of variables \( z = \sin \varphi \) in the integral \( I_n(r) \). We have \( y^{(2)} = rz, \ y^{(1)} = r(1 - \frac{y^{(2)2}}{r})^{1/2} = r(1 - \frac{y^{(2)2}}{2r}) + O(y^{(2)4}), \ z \leq 2 \frac{\sqrt{\varepsilon}}{r} c^{-0.2n} \), hence \( |y^{(1)} - rz^2| \leq K \varepsilon^2 c^{-0.8n} \) with some \( K > 0 \) independent of \( \varepsilon \). These relations imply that if \( \varphi \in \Gamma_n(r, x) \) then

\[
\frac{c^n}{T} \left( x^{(1)} y^{(1)} + x^{(2)} y^{(2)} \right) + g_{n+1} \left( \frac{x^{(1)} + y^{(1)}}{2} - M \right) A_{n+1} \left( \frac{x^{(2)} + y^{(2)}}{2} \right)^2 \\
\leq \frac{c^n}{T} r \left( x^{(1)} \left( 1 - \frac{z^2}{2} \right) + x^{(2)} z \right) + g_{n+1} \left( \frac{x^{(1)} + r(1 - \frac{z^2}{2})}{2} - M \right) \\
+ A_{n+1} \left( \frac{x^{(2)} + rz}{2} \right)^2 + K \varepsilon^2 c^{0.2n}
\]

with some \( K > 0 \). Since \( \frac{dz}{d\varphi} > \frac{1}{2} \) and \( |r - M| < \varepsilon c^{-0.4n} \) if \( \Gamma_n(r, x) \) is not empty, the above inequality together with (5.16) imply that

\[
I_n(r) = 0 \quad \text{if } |r - M| \geq \varepsilon c^{-0.4n} \tag{5.17}
\]

and

\[
I_n(r) \leq 3 \exp(K \varepsilon^2 c^{0.2n}) \hat{I}_n(r) \quad \text{for } |r - M| < \varepsilon c^{-0.4n} \tag{5.17'}
\]
with
\[ \hat{I}_n(r) = \int_{-\infty}^{\infty} M \exp \left\{ \frac{c^n}{T} r \left( x^{(1)} \left( 1 - \frac{z^2}{2} \right) + x^{(2)} z \right) \right. \\
+ \left. g_{n+1} \left( \frac{x^{(1)} + r(1 - \frac{z^2}{2}) - M}{2} \right) + A_{n+1} \left( \frac{x^{(1)} + rz}{2} \right)^2 \right\} dz. \]

The expression \( \hat{I}_n(r) \) can be calculated explicitly, and we get that for \( |r - M| < \varepsilon_c^{-0.4n} \)
\[ \hat{I}_n = \sqrt{\pi} M \left( \frac{c^n}{2T} r x^{(1)} + \frac{g_{n+1}}{4} r - \frac{A_{n+1}}{4} r^2 \right)^{-1/2} \exp \left\{ \frac{c^n x^{(1)} r}{T} + g_{n+1} \left( \frac{x^{(1)} + r}{2} - M \right) + A_n (x^{(1)}, r) x^{(2)} \right\} \]
with
\[ A_n(x^{(1)}, r) = \frac{A_{n+1}}{4} + \frac{(\frac{c^n}{T} + \frac{A_{n+1}}{2})^2}{2(\frac{c^n x^{(1)}}{T} + \frac{g_{n+1}}{2} + \frac{A_{n+1}}{2})}. \]

Hence
\[ \hat{I}_n(r) \leq K c^{-n/2} \exp \left\{ \frac{c^n x^{(1)} r}{T} + g_{n+1} \left( \frac{x^{(1)} + r}{2} - M \right) + A_n (x^{(1)}, r) x^{(2)} \right\} \]
\[ = K c^{-n/2} \exp \left\{ \frac{c^n}{T} M^2 + g_n (x^{(1)} - M) + A_n (x^{(1)}, r) x^{(2)} \right\} \]
\[ + \left( \frac{c^n}{T} + \frac{g_{n+1}}{2} \right) (r - M) \].

Observe that \( A_n = A_n(M, M) \), and if \( x \in \tilde{\Omega}_n^2 \), \( y \in V_n^\varepsilon(x) \) and \( \frac{x+y}{2} \in \Omega_1^1 \) then
\[ |A_n(x^{(1)}, r) - A_n| \leq K_0 c^n (|x^{(1)} - M| + |r - M|) \]
\[ |x^{(1)} - M - \left( |x| - M - \frac{x^{(2)}}{2M} \right)| \leq K_1 x^{(2)} + |x^{(1)} - M|, \]
hence (5.17') implies that
\[ I_n(r) \leq \bar{K} c^{-n/2} \exp \left\{ K_2 c^{0.2n} + \frac{c^n}{T} M^2 + g_n (x^{(1)} - M) + A_n x^{(2)} \right\} \]
\[ + K_0 c^n (|x^{(1)} - M| + |r - M|) x^{(2)} + \left( \frac{c^n x^{(1)}}{T} + \frac{g_{n+1}}{2} \right) (r - M) \]
\[ \leq \bar{K} c^{-n/2} \exp \left\{ K_2 c^{0.2n} + \frac{c^n}{T} M^2 + g_n (|x| - M) - \left( \frac{g_n}{2M} - A_n \right) x^{(2)} \right\} \]
\[ + \left[ K_0 c^n (|x^{(1)} - M| + |r - M|) + K_1 g_n (x^{(2)} + |x^{(1)} - M|) \right] x^{(2)} \]
\[ + \left( \frac{c^n x^{(1)}}{T} + \frac{g_{n+1}}{2} \right) (r - M) \]
if $|r - M| \leq \varepsilon c^{-0.4n}$. Since $\frac{g_{n}}{2M} - A_{n} > \alpha c^{n}$ with some $\alpha > 0$ and $x^{(2)} > (1 + \varepsilon)c^{-0.4n}$ if $x \in \hat{\Omega}_{n}^{2}$

$$K \varepsilon^{2}c^{0.2n} + \left[ K_{0}c^{n}(|x^{(1)} - M| + |r - M|) + K_{1}g_{n}(x^{(2)} + |x^{(1)} - M|) \right] x^{(2)^{2}}$$

$$- \left( \frac{g_{n}}{2M} - A_{n} \right) x^{(2)^{2}}$$

$$= K \varepsilon^{2}c^{0.2n} - \left( \frac{g_{n}}{2M} - A_{n} + O \left( c^{0.8n} \right) \right) x^{(2)^{2}}$$

$$\leq K \varepsilon^{2}c^{0.2n} - \left( \frac{g_{n}}{2M} - A_{n} + O \left( c^{0.8n} \right) \right) (1 + \varepsilon)^{2}c^{-0.8n}$$

$$\leq - \left( \frac{g_{n}}{2M} - A_{n} \right) (1 + \varepsilon)c^{-0.8n}$$

if $\varepsilon > 0$ is chosen sufficiently small. This relation together with (5.18) imply that

$$I_{n}(r) \leq Kc^{-n/2} \exp \left\{ c^{n} \cdot M^{2} + g_{n}(|x| - M) - (1 + \varepsilon) \left( \frac{g_{n}}{2M} - A_{n} \right) c^{-0.8n} \right\}$$

$$+ \left( \frac{c^{n}x^{(1)}}{T} + \frac{g_{n+1}}{2} \right) (r - M) \right\} \text{ if } |r - M| < \varepsilon c^{-0.4n}.$$
By Lemma 1 \( g_{n+1}c^{-(n+1)} \geq g_n c^n \), \( A_n c^{-n} \geq A_{n+1}c^{-(n+1)} \) i.e. \( g_{n+1} \geq c g_n \) and \( cA_n \geq A_{n+1} \). Hence

\[
\left( \frac{g_{n+1}}{2M} - A_{n+1} \right)c^{-0.8(n+1)} \geq \left( \frac{g_n}{2M} - A_n \right)c^{-0.8n}c^{0.2} \geq (1 + \varepsilon)\left( \frac{g_n}{2M} - A_n \right)c^{-0.8n}
\]

if \( 0 < \varepsilon < c^0.2 - 1 \). Since \( \varepsilon(\frac{g_n}{2M} - A_n)c^{-0.8n} > c'c^{0.2} \) with some appropriate \( c'(\varepsilon) > 0 \) we get that

\[
\left( \frac{g_{n+1}}{2M} - A_{n+1} \right)c^{-0.8(n+1)} \geq \left( \frac{g_n}{2M} - A_n \right)c^{-0.8n} + c'c^{0.2n}
\]

Therefore

(5.19)

\[
S_n^2 f(x) \leq K \exp \left\{ \frac{c^n}{T} M^2 + g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right)c^{-0.8n} + q^{n+1} - \varepsilon'c^{0.2n} \right\}
\]

\[
\leq L_n \exp \left\{ g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right)c^{-0.8n} - \frac{\varepsilon'}{2}c^{0.2n} \right\}.
\]

This is estimate (4.5) (with \( \frac{\varepsilon'}{T} \) instead of \( \varepsilon' \)). For \( x \in \Omega^1_n \)

\[
g_n(x^{(1)} - M) + A_n x^{(2)2} = g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right)x^{(2)2} + O(c^{-0.2n}),
\]

hence (5.19) implies that for \( x \in \Omega^1_n \)

\[
S_n^2 f(x) \leq L_n \exp \left\{ g_n(x^{(1)} - M) + A_n x^{(2)2} - \frac{\varepsilon'}{4}c^{0.2n} \right\},
\]

and this is relation (4.2) (with \( \frac{\varepsilon'}{T} \)).

d) The estimation of \( S_n^3 f(x) \) for \( x \in \Omega^1_n \) and \( x \in \Omega^2_n \).

Clearly

\[
S_n^3 f(x) = S_n^2 f(x) + \int_{\{y, \frac{x+y}{2} \in \Omega^3_{n+1} \cap V^c_n(x)\}} \exp \left( \frac{c^n}{T} xy \right)f \left( \frac{x+y}{2} \right)p_n(y) dy,
\]

where \( S_n^2 f(x) \) and \( V^c_n \) are defined in (4.8) and (4.8'). The term \( S_n^2 f(x) \) is bounded in (5.13). On the other hand, we claim that there is some \( \varepsilon_0 = \varepsilon_0(c) \) such that if \( 0 < \varepsilon < \varepsilon_0 \) and \( x \in \Omega^1_n \cup \Omega^2_n \) then the set \( \{y, \frac{x+y}{2} \in \Omega^3_{n+1} \cap V^c_n(x)\} \) is empty, hence the last integral is zero. We have to show that if \( ||x| - M| < c^{-0.4n}, y \in V^c_n \), i.e. \( ||y| - M| < \varepsilon c^{-0.4n} \) and \( xy > |x||y| - \varepsilon c^{-0.4n} \) then \( \frac{x+y}{2} \notin \Omega^3_{n+1} \), i.e. \( c^{-0.4(n+1)} \leq \frac{x+y}{2} \)

Estimation from above:

\[
\left| \frac{x+y}{2} \right| - M \leq \frac{|x| - M}{2} + \frac{|y| - M}{2} \leq \frac{1 + \varepsilon}{2}c^{-0.4n} \leq c^{-0.4(n+1)}
\]
if $\frac{1+\varepsilon}{2} \leq c^{-0.4}$, what holds for sufficiently small $\varepsilon$.

Estimation from below:

$$\left| \frac{x+y}{2} \right|^2 - M^2 = \left| \frac{x}{2} \right|^2 + \left| \frac{y}{2} \right|^2 + 2 \frac{xy}{4} - M^2$$

$$\geq \frac{|x|^2 + |y|^2 + 2 |x||y| - 2\varepsilon c^{-0.4n}}{4} - M^2 = \left( \frac{|x| + |y|}{2} \right)^2 - M^2 - \frac{\varepsilon}{2} c^{-0.4n}$$

$$\geq \left( \frac{M - c^{-0.4n} + M - \varepsilon c^{-0.4n}}{2} \right)^2 - M^2 - \frac{\varepsilon}{2} c^{-0.4n}$$

$$\geq -(1 + \varepsilon) Mc^{-0.4n} - \frac{\varepsilon}{2} c^{-0.4n} = -M \left( 1 + \varepsilon + \frac{\varepsilon}{2M} \right) c^{-0.4n}.$$ 

Hence

$$\left| \frac{x+y}{2} \right| \geq \left( M^2 - M \left( 1 + \varepsilon + \frac{\varepsilon}{2M} \right) c^{-0.4n} \right)^{1/2}$$

$$\geq M - \left( 1 + \varepsilon + \frac{\varepsilon}{2M} \right) c^{-0.4n} - Kc^{-0.8n}$$

$$= M - \left[ \left( \frac{1}{2} + \frac{\varepsilon}{2} \left( 1 + \frac{1}{2M} \right) \right) + Kc^{-0.4n} \right] c^{-0.4n} \geq M - c^{-0.4(n+1)}.$$ 

if $\frac{1}{2} + \frac{\varepsilon}{2} (1 + \frac{1}{2M}) + K c^{-0.4n} \leq c^{-0.4}$, what we had to show.

Hence $S_n^3 f(x) = S_n^\varepsilon f(x)$, and (5.13) implies (4.6). To prove (4.3) we still have to remark that for $x \in \Omega_n^1$

$$g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) c^{-0.8n} - \frac{1}{2} c^{n/2}$$

$$\leq g_n(|x| - M) - \left( \frac{g_n}{2M} - A_n \right) x^{(2)}^2 - \frac{1}{2} c^{n/2} \leq g_n x^{(1)} - M + A_n x^{(2)}^2 - \frac{1}{6} c^{n/2}.$$ 

e) The estimation of $S_n f(x)$ for $x \in \Omega_n^3$.

We get from (5.12) and Part a) of Lemma 2 that

$$0 \leq S_n f(x) \leq 2 c^n \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) - \frac{c^n}{3T} (|x|^2 - M^2) \right\}.$$ 

For $x \in \Omega_n^3$ $\frac{c^n}{3T} (|x|^2 - M^2) \geq \frac{c^{0.2n}}{3T}$, hence (4.1’’) implies that

$$0 \leq S_n f(x) \leq 2 c^n \exp \left\{ \frac{c^n}{T} M^2 + \frac{g_n}{2M} (|x|^2 - M^2) - \frac{c^{0.2n}}{3T} \right\}$$

$$\leq L_n \exp \left\{ \frac{g_n}{2M} (|x|^2 - M^2) \right\},$$

i.e. relation (4.7) holds, as we have claimed. Proposition 1’ is proved.
6. The proof of Theorem 1 and Proposition 2. Existence of the thermodinamical limit. First we show with the help of Proposition 1, Lemma 1 and Theorem A that for all \( q, c^{-0.2} < q < 1 \), there exist some thresholds \( n_0 \) and \( N_0(n, q) \) for \( n \geq n_0 \) such that if \( n \geq n_0 \) and \( N \geq N_0(n, q) \) then

\[
\frac{d\mu_{n,N}^{h_n}}{d\mu_n}(x_1, \ldots, x_{2^n}) = f_{n,N}^{h_n}(2^{-n} \sum_{j=1}^{2^n} x_j)
\]

with

\[
f_{n,N}^{h_n}(x) = L_n \exp \left\{ \bar{g}c^n(x(1) - M) + \bar{A}c^n x(2)^2 + \varepsilon_n(x) \right\} \quad \text{for } x \in \Omega_n^1,
\]

where

\[
\sup_{x \in \Omega_n^1} |\varepsilon_n(x)| \leq q^n,
\]

\[
f_{n,N}^{h_n}(x) \leq L_n \exp \left\{ \frac{\bar{g}c^n}{M} (|x|^2 - M^2) \right\} \quad \text{if } x > M + c^{-0.4n}
\]

\[
f_{n,N}^{h_n}(x) \leq L_n \exp \left\{ \frac{\bar{g}c^n}{2M} (|x|^2 - M^2) \right\} \quad \text{if } 0 < x < M - c^{-0.4n}
\]

with some appropriate norming constant \( L_n \) which satisfies the relation

\[
C_1 < c^{-n/2}L_n < C_2 \quad \text{with some } 0 < C_1 < C_2 < \infty.
\]

Indeed, Proposition 1 and Lemma 1 imply (6.1)–(6.4') with some norming constant \( L_n = L_n(N, h_N) \). (In the domain \( \Omega_n^2 \) we have divided the cases \( |x|^2 - M^2 > 0 \) and \( |x|^2 - M^2 < 0 \), since here we apply that \( \bar{g}c^n < g_n < 2\bar{g}c^n \).) It remains to prove (6.5) and to show that \( L_n \) can be chosen independently of \( N \) and \( h_N \). For this aim we observe that

\[
1 = \mu_{n,N}^{h_n}((R^2)^{2^n}) = \int f_{n,N}^{h_n}(x)p_n(x) \, dx = \int_{\Omega_n^1} + \int_{\Omega_n^2} + \int_{\Omega_n^3}.
\]

By applying the change of variables \( r = M + c^{-n}t \) and by using the function \( f_n(t) \) defined in (4.11) we get that

\[
\int_{\Omega_n^1} f_{n,N}^{h_n}(x)p_n(x) \, dx \leq \int_{0}^{M - c^{-0.4n}} L_n(N, h_N) \exp \left\{ \frac{\bar{g}c^n}{2M} (r^2 - M^2) \right\} r\bar{p}_n(r) \, dr
\]

\[
+ \int_{M + c^{-0.4n}}^{\infty} L_n(N, h_N) \exp \left\{ \frac{\bar{g}c^n}{M} (r^2 - M^2) \right\} r\bar{p}_n(r) \, dr
\]

\[
= L_n(N, h_N) \left[ \int_{-c^{-0.6n}}^{c^{-0.6n}} \exp \left\{ \frac{\bar{g}}{2M} \left( 2Mt + \frac{t^2}{c^n} \right) \right\} (M + c^{-n}t) f_n(t) \, dt + \int_{-c^{-0.6n}}^{c^{-0.6n}} \exp \left\{ \frac{\bar{g}}{M} \left( 2Mt + \frac{t^2}{c^n} \right) \right\} (M + c^{-n}t) f_n(t) \, dt \right].
\]
Relations (4.11') and (4.11'') imply that

\[(6.7) \quad \int_{\Omega_n^0} = L_n(N, h_N) O \left(\exp(-Kc^{0.6n})\right) \quad \text{with some } K > 0.\]

Similarly,

\[(6.7') \quad \int_{\Omega_n^0} \leq L_n(N, h_N) \exp \left\{ -\frac{1}{2} \left( \frac{\bar{g}}{2M} - \bar{A} \right) c^{0.2n} \right\}.\]

(Observe that $\frac{\bar{g}}{2M} - \bar{A} > 0.$)

Define the number $T_n,$

\[(6.8) \quad T_n = \int_{\Omega_n^0} \exp \left\{ \bar{g}c^n(x^{(1)} - M) + \bar{A}c^n x^{(2)} \right\} p_n(x) \, dx.\]

It follows from Theorem A that

\[(6.9) \quad C_1 c^{-n/2} < T_n < C_2 c^{-n/2} \quad \text{with some } 0 < C_1 < C_2 < \infty.\]

Indeed, since the expression in the exponent of (6.8) can be written in the form

\[\bar{g}c^n(x^{(1)} - M) + \bar{A}c^n x^{(2)} = \bar{g}c^n(|x| - M) - \left( \frac{\bar{g}}{2M} - \bar{A} \right) c^n x^{(2)} + O(c^{-0.2n}),\]

we get (6.9) by integrating (6.8) first by the variable $x^{(2)}$. Some calculation with the help of (6.2) and (6.2') shows that

\[(6.10) \quad \left| \int_{\Omega_n^0} \int_{A(N)}^N \, dx - L_n(N, h_N)T_n \right| \leq L_n(N, h_N)T_n q^n.\]

Relations (6.6), (6.7), (6.7') and (6.10) imply that

\[1 = L_n(N, h_N)T_n \left( 1 + \varepsilon_n + O(\exp(-c^{0.1n})) \right), \quad \text{and } \varepsilon_n \leq q^n.\]

The last relation implies that relations (6.1)--(6.4') remain valid if we choose $L_n = T_n^{-1}$, and this $L_n$ satisfies (6.5) by (6.9).

We prove Theorem 1 with the help of (6.1)--(6.5). Fix some integer $k \geq 0$, and define for all $n \geq k$ and measurable sets $A \subset (R^2)^{2k}$ the cylindrical set $A(n) = A \times (R^2)^{2n-2k} \subset (R^2)^{2n}$. Put

\[\mu_n(A) = L_n \int_{\tilde{A}(n)} \exp \left\{ \bar{g}c^n 2^{-n} \sum_{j=1}^{2^n} (x^{(1)}_j - M) + \bar{A}c^n 4^{-n} \left( \sum_{j=1}^{2^n} x^{(2)}_j \right)^2 \right\} \mu_n(dx_1 \ldots, dx_{2^n})\]
with \( \bar{A}(n) = A(n) \cap \{(x_1, \ldots, x_{2^n}) \}, \ 2^{-n} \sum_{j=1}^{2^n} x_j \in \Omega^1_n \}. \) We claim that if \( n > n_0 \) and \( N > N_0(n, q) \) then

\[
(6.11) \quad |\tilde{\mu}_n(A(n)) - \mu_{h,N}^{A}(A(n))| \leq Kq^n
\]

with some \( K > 0 \) independent of the set \( A \). Indeed,

\[
|\tilde{\mu}_n(A(n)) - \mu_{h,N}^{A}(A(n))| \leq \int_{\Omega^1 \cup \Omega^2} f_{h,N}^{A}(x)p_n(x) \, dx
\]

\[
+ \int_{\bar{A}(n)} \mu_n(dx_1, \ldots, dx_{2^n}) f_{h,N}^{A}(2^{-n} \sum_{j=1}^{2^n} x_j)
\]

\[
- L_n \exp \left\{ \bar{g}c^n 2^{-n} \sum_{j=1}^{2^n} (x_j^{(1)} - M) + \bar{A}c^n 4^{-n} \left( \sum_{j=1}^{2^n} x_j^{(2)} \right)^2 \right\} = I_1 + I_2.
\]

It follows from (6.5), (6.7) and (6.7') that

\[
(6.12) \quad I_1 \leq \exp(-c^{0.1n}).
\]

On the other hand we increase the term \( I_2 \) by enlarging the domain of integration to the set \( \{(x_1, \ldots, x_{2^n}) \}, \ 2^{-n} \sum_{j=1}^{2^n} x_j \in \Omega^1_n \}. \) Hence

\[
I_2 \leq \int_{\Omega_n} \left| f_{h,N}^{A}(x) - L_n \exp \left\{ \bar{g}c^n (x^{(1)} - M) + \bar{A}c^n x^{(2)} \right\} \right| p_n(x) \, dx.
\]

The last inequality together with (6.2) and (6.2') imply that

\[
I_2 \leq 2q^n f_{h,N}^{A}((R^2)^{2^n}) = 2q^n.
\]

The last inequality together with (6.12) imply (6.11). Since for all \( k \geq 0 \) and measurable sets \( A \in (R^2)^{2^n} \), \( k \leq n \leq N \) we have \( \mu_{h,N}^{A}(A(n)) = \mu_{h,N}^{A}(A) \), relation (6.11) implies that for all \( \varepsilon > 0 \) there is some \( N_0(\varepsilon) \) such that for \( N > N_0(\varepsilon) \) and \( N' > N_0(\varepsilon) \) the relation \( |\mu_{h,N}^{A}(A) - \mu_{h,N'}^{A}(A)| \leq \varepsilon \) holds true. Let us emphasize that the threshold \( N(\varepsilon) \) does not depend on the set \( A \). Hence the last relation means that the limit \( \bar{\mu}_k(A) = \lim_{N \to \infty} \mu_{h,N}^{A}(A) \) exists, and the convergence is uniform in \( A \). This implies that \( \mu_{h,N}^{A} \to \bar{\mu}_k \) in variational metric. To complete the proof of Theorem 1 we have to show that the measure \( \bar{\mu}_k \) does not depend on the sequence \( h_N \). But it is not difficult to see with the help of (6.11) that this statement holds, since \( \bar{\mu}_k(A) = \lim_{n \to \infty} \bar{\mu}(A(n)) \), and the right hand side of the last expression does not depend on \( h_N \).

**Proof of Proposition 2.** Let \( n > n_0 \) and \( N > N_0(n, q) \). Relations (6.1)–(6.5) hold for such pairs \( n \) and \( N \). By Theorem 1 the measures \( \mu_{h,N}^{A} \) converge in variational metric to the projection \( \bar{\mu}_n \) of the measure \( \bar{\mu} \) to \((R^2)^{2^n}\) as \( N \to \infty \). Since all measures are absolute continuous with respect to the measure \( \mu_n \) the above convergence is equivalent to the convergence of the Radon–Nikodym derivatives \( \frac{d\mu_{h,N}^{A}}{d\mu_n} = f_{h,N}^{A} \) to \( \frac{d\bar{\mu}_n}{d\mu_n} = \bar{f} \) in \( L_1 \) norm in the space \( ((R^2)^{2^n}, \mu_n) \) as \( N \to \infty \). Since all functions \( f_{h,N}^{A}(x) \) satisfy (6.1)–(6.5) for \( N > N_0(n, q) \), their limit \( \bar{f} \) also has this property. Hence Proposition 2 holds true.
7. The proof of Theorem 2. Existence of the large-scale limit. First we need some results about the transformation \( Q_n = Q_n(k) \) of probability measures on \((R^2)^{2^{n+k}}\) to probability measures on \((R^2)^{2^k}\) to be defined below. First we define a transformation \( Q_n = Q_n(k), \) \( Q_n: (R^2)^{2^{n+k}} \to (R^2)^{2^k} \) in the following way: For all \((x_1, \ldots, x_{2^{n+k}}), \) \( x_j \in R^2, \) \( j = 1, \ldots, 2^{n+k} \)

\[
Q_n(x_1, \ldots, x_{2^{n+k}}) = (y_1, \ldots, y_{2^k}), \quad y_j = 2^{-n} \sum_{l=(j-1)2^{n+1}}^{j2^n} x_l, \quad j = 1, \ldots, 2^k.
\]

This transformation induces the transformation \( Q_n \) of probability measures on \((R^2)^{2^{n+k}}\) to probability measures on \((R^2)^{2^k}\) in a natural way. Namely, if \( \nu \) is a probability measure on \((R^2)^{2^{n+k}}\) and \( (\eta(1), \ldots, \eta(2^{n+k})) \) is a \( \nu \) distributed vector then \( Q_n \nu \) is the distribution of the random vector \( Q_n(\eta(1), \ldots, \eta(2^{n+k})) \). In Theorem 2 we have to study an appropriately rescaled version of the measure \( Q_n \mu_{n+k} \).

It is not difficult to see that relation (2.10) implies that

\[
(7.1) \quad \frac{dQ_{n+1}}{dQ_n \mu_{n+k}}(x_1, \ldots, x_{2^k}) = f_{n+k} \left( 2^{-k} \sum_{j=1}^{2^k} x_j \right).
\]

We formulate below Theorem C which follows from the relatively simple Theorem 1 in [4]. For the sake of completeness we present its proof in Appendix B.

**Theorem C.** The above defined measure \( Q_n \mu_{n+k} = Q_n \mu_{n+k}(T, t) \) has a density function \( h_k(x_1, \ldots, x_{2^k}) \) of the form

\[
h_k(x_1, \ldots, x_{2^k}) = L(T, t, n, k) \exp \left\{ -\frac{1}{T} \mathcal{H}_k \left( e^{n/2}x_1, \ldots, e^{n/2}x_{2^k} \right) \right\} \prod_{j=1}^{2^k} p_n(x_j)
\]

\[
x_j \in R^2, \quad j = 1, \ldots, 2^k.
\]

Here \( \mathcal{H}_k \) is the Hamiltonian function defined in (1.2\') of Part I, \( p_n(x) \) is the function appearing in Theorem A, and \( L \) is an appropriate norming constant.

Formula (7.1) and Theorem C enable us to express the density function of the random vector \( \{\mathcal{R}_n \sigma_{(1)}(j), \mathcal{R}_n \sigma_{(2)}(j), 1 \leq j \leq 2^k \} \) with the help of the functions \( p_n(x) \) and \( f_n(x) \), where the sequence \( \{\sigma(j), j \in Z\} \) is \( \mu \) distributed, and \( \mathcal{R}_n \sigma_{(1)}, \mathcal{R}_n \sigma_{(2)} \) are defined in (1.2) and (1.3) of Part II. This density equals to

\[
(7.2) \quad h_{n,k}(x_1, \ldots, x_{2^k})
\]

\[
= L_{n,k} f_{n+k} \left( 2^{-k} \sum_{j=1}^{2^k} \bar{x}_j \right) \exp \left\{ -\frac{1}{T} \mathcal{H}_k \left( e^{n/2} \bar{x}_1, \ldots, e^{n/2} \bar{x}_{2^k} \right) \right\} \prod_{j=1}^{2^k} p_n(\bar{x}_j)
\]

with

\[
(7.2') \quad \bar{x} = \bar{x}(x) = \left( M + e^{-n}x_{(1)}, e^{-n/2}x_{(2)} \right) \quad \text{for} \quad x = \left( x_{(1)}, x_{(2)} \right).
\]
Let us define the sets \( W_n \subset R^2 \) and \( \tilde{W}_n \subset R^2 \) by the formulas
\[
\tilde{W}_n = \left\{ (x^{(1)}, x^{(2)}), \ M - \frac{\eta \gamma}{\epsilon^n} < |x| < M + \frac{\eta \gamma^{1/\alpha}}{\epsilon^n}, \ |x^{(2)}| < c^{-0.45n}, \ x^{(1)} > 0 \right\}
\]
\[
W_n = \left\{ (x^{(1)}, x^{(2)}), \ \tilde{x}(x) \in \tilde{W} \right\},
\]
where \( \eta \) and \( \alpha \) are the same constants as in Theorem A, and \( \tilde{x}(x) \) is defined in (7.2'). We shall show that there is some \( n_0 > 0 \) and \( 0 < q < 1 \) such that
\[
P \left( (\mathcal{R}_n \sigma^{(1)}(j), \mathcal{R}_n \sigma^{(2)}(j)) \notin W_n \right) \leq q^n \quad \text{if} \ n \geq n_0
\]
for a \( \tilde{\mu} \) distributed random field \( \sigma(j), \ j \in \mathbb{Z} \), and give a good asymptotic formula for the expression \( h_{n,k}(x_1, \ldots, x_{2^k}) \) defined in (7.2) if \( x_j \in W_n \) for all \( 1 \leq j \leq 2^k \). First we prove (7.3).

\[
P \left( (\mathcal{R}_n \sigma^{(1)}(j), \mathcal{R}_n \sigma^{(2)}(j)) \notin W_n \right) = P \left( 2^{-n} \sum_{l=1}^{2^k} \sigma(k) \notin \tilde{W}_n \right)
= \int_{R^2-\tilde{W}_n} f_n(x)p_n(x)dx = \int_{\Omega^1_n-W_n} + \int_{\Omega^2_n} + \int_{\Omega^3_n} = I_1 + I_2 + I_3.
\]
We get similarly to the estimates (6.7) and (6.7') that
\[
I_3 \leq \exp \left( -Ce^{0.6n} \right),
\]
\[
I_2 \leq \text{const.} \exp \left\{ - \left( \frac{\bar{g}}{2M} - \bar{A} \right)e^{0.2n} \right\} \leq \exp \left( -Ce^{0.2n} \right)
\]
The term \( I_1 \) has to be estimated a little more carefully.
Define
\[
\tilde{\Omega}^1_n = \left\{ (x^{(1)}, x^{(2)}), \ | |x| - M| < c^{-0.4n}, \ |x^{(2)}| < c^{-0.45n}, \ x^{(1)} > 0 \right\},
\]
and write
\[
I_1 = I_{1,1} + I_{1,2} = \int_{\Omega^1_n-\tilde{\Omega}^1_n} + \int_{\tilde{\Omega}^1_n-W_n}.
\]
We get, similarly to the estimation of \( I_3 \) and \( I_2 \) that \( I_{1,1} \leq \exp(-Kc^{0.1n}) \), and we can write by (2.11) and (2.11') that
\[
I_{1,2} \leq 2L_n \int_{\tilde{\Omega}^1_n-W_n} \exp \left\{ \bar{g}c^n(x^{(1)} - M) + \bar{A}c^n(x^{(2)})^2 \right\} p_n(x)dx
\]
\[
\leq 3L_n \int_{\tilde{\Omega}^1_n-W_n} \exp \left\{ \bar{g}c^n(|x| - M) - \left( \frac{\bar{g}}{2M} - \bar{A} \right)e^{x^{(2)}} \right\} p_n(x)dx.
\]
Then integrating first by \( x^{(2)} \) we get that
\[
I_{1,2} \leq KL_n e^{-n/2} \left[ \int_{M-c^{-0.4n}}^{M-\eta \epsilon e^{-n}} + \int_{M+c^{-0.4n}}^{M+e^{-0.4n}} \right] \exp \left( \bar{g}c^n(r-M) \right) \tilde{p}_n(r)dr \leq q^n
\]
with the help of relations (1.8) and (1.10) in Theorem A. Let us emphasize that it was the multiplying term \( q^n \) in (1.8) that enabled us to give an exponentially small bound for the second term in the last integral. The above estimates imply (7.3).

To estimate the expression (7.2) in the case \( x_j \in W_n, \ j = 1, 2, \ldots, 2^k \), we make some preparatory remarks. Put

\[
\ell_n(x) = c^n(|\tilde{x}(x)| - M) = c^n \left\{ \left[ (M + c^{-n}x(1))^2 + c^{-n}x(2)^2 \right]^{1/2} - M \right\}
\]

We have

(7.4)

\[
\ell_n(x) = x(1) + \frac{x(2)^2}{2M} + O \left( c^{-n} \left( x(2)^2 + |x(1)|x(2)^2 \right) \right) = x(1) + \frac{x(2)^2}{2M} + O(c^{-0.8n}) \quad \text{if} \ x \in W_n,
\]

because, as it is not difficult to see, \( c^{-n/2}|x(2)| < c^{-0.45n} \), and \( M - 2c^{-0.9n} < M + c^{-n}x(1) < M + 2c^{-0.9n} \) if \( x \in W_n \). We show with the help of Theorem A and (7.4) that

(7.5)

\[
p_n(\tilde{x}) = \exp \left\{ -\frac{a_0}{T} \left( Mx(1) + \frac{x(2)^2}{2} \right) \right\} g \left( \frac{a_1}{T} \left( Mx(1) + \frac{x(2)^2}{2} \right) \right) \left( 1 + O(q^n) \right)
\]

with some \( 0 < q < 1 \) for \( x \in W_n \) if \( \eta > 0 \) is chosen sufficiently small in Theorem A. Indeed, by Theorem A

\[
p_n(\tilde{x}) = \exp \left\{ -\frac{a_0}{T} \ell_n(x) \right\} g \left( \frac{a_1}{T} \ell_n(x) \right) \left( 1 + O(q^n) \right),
\]

\[
\exp \left\{ -\frac{a_0M}{T} \ell_n(x) \right\} = \exp \left\{ -\frac{a_0}{T} \left( Mx(1) + \frac{x(2)^2}{2} \right) + O(c^{-0.8n}) \right\},
\]

and

(7.6)

\[
\left| g \left( \frac{a_1}{T} M \ell_n(x) \right) - g \left( \frac{a_1}{T} \left( Mx(1) + \frac{x(2)^2}{2} \right) \right) \right| = O(c^{-0.8n})
\]

by (7.4) and the boundedness of the function \( \frac{d}{dx} g(x) \). (See Lemma 13 in Part I.) On the other hand, by Lemma 17 of Part I the relation

\[
-2\eta \left( x(1) + \frac{x(2)^2}{2M} \right) < 2\eta n \quad \text{if} \ x \in W_n,
\]

which holds because of the definition of \( W_n \), and the inequality

\[
\left| c^n(|\tilde{x}(x)| - M) - \left( x(1) + \frac{x(2)^2}{2M} \right) \right| \leq K c^{-0.8n}
\]

we have

\[
g \left( \frac{a_1}{T} \left( Mx(1) + \frac{x(2)^2}{2} \right) \right) > c^{-0.3n}
\]
if \( x \in W_n \), and \( \eta \) is chosen in Theorem A sufficiently small. Hence (7.6) can be rewritten as

\[
g\left( \frac{a_1}{T}M_{\ell}(x) \right) = g\left( \frac{a_1}{T} \left( Mx^{(1)} + \frac{x^{(2)}_2}{2} \right) \right) \left( 1 + O(c^{-n/2}) \right).
\]

The above relations imply (7.5).

We also claim that

(7.7)

\[
\mathcal{H}_k \left( c^{n/2}x_1, \ldots, c^{n/2}x_{2^k} \right)
\]

\[
= - \sum_{i=1}^{2^k-1} \sum_{j=i+1}^{2^k} U(i, j) \left[ M(x_i^{(1)} + x_j^{(1)}) + x_i^{(2)} x_j^{(2)} + c^n M^2 \right] + O(q^n)
\]

and

(7.8)

\[
\bar{f}_{n+k} \left( 2^{-k} \sum_{j=1}^{2^k} \bar{x}_j \right) = L_{n+k} \exp \left\{ \frac{c^k}{2\tilde{g}} \sum_{j=1}^{2^k} x_j^{(1)} + \frac{c^k}{4\tilde{A}} \left( \sum_{j=1}^{2^k} x_j^{(2)} \right)^2 + O(q^n) \right\}
\]

if \( x_j \in W_n, j = 1, \ldots, 2^k \).

Indeed,

\[
\mathcal{H}_k \left( c^{n/2}x_1, \ldots, c^{n/2}x_{2^k} \right)
\]

\[
= - \sum_{i=1}^{2^k-1} \sum_{j=i+1}^{2^k} U(i, j) \left[ c^n (M + c^{-n}x_i^{(1)})(M + c^{-n}x_j^{(1)}) + x_i^{(2)} x_j^{(2)} \right],
\]

hence to prove (7.7) it is enough to remark that in the last expression the terms \( c^{-n}x_i^{(1)} x_j^{(1)} \) are negligibly small, since \( c^{-n}x_i^{(1)} x_j^{(1)} = O(c^{-0.8n}) \) if \( x_i \in W_n \) and \( x_j \in W_n \).

To prove (7.8) we have to show that \( 2^{-k} \sum_{j=1}^{2^k} \bar{x}_j \in \Omega_{n+k}^1 \) if \( \bar{x}_j \in \bar{W}_n \) for all \( j = 1, \ldots, 2^k \) and then apply Proposition 2. We can write with the notation \( \bar{x}_j = (\bar{x}_j^{(1)}, \bar{x}_j^{(2)}) \) that \( 2^{-k} \sum_{j=1}^{2^k} \bar{x}_j = M + c^{-n}x_i^{(1)}(M + c^{-n}x_j^{(1)}) + x_i^{(2)} x_j^{(2)} \) if \( n \) is sufficiently large (\( k \) is fixed, \( n \to \infty \)) and \( \bar{x}_j \in W_n \) for \( j = 1, \ldots, 2^k \). These relations imply that \( 2^{-k} \sum_{j=1}^{2^k} \bar{x}_j \in \Omega_{n+k}^1 \). We get, by putting (7.5), (7.7) and (7.8) into (7.2) that

\[
h_{n,k}(x_1, \ldots, x_{2^k}) = L_{n,k} \exp \left\{ \frac{c^k}{2\tilde{g}} \sum_{j=1}^{2^k} x_j^{(1)} + \frac{c^k}{4\tilde{A}} \left( \sum_{j=1}^{2^k} x_j^{(2)} \right)^2 \right\}
\]

\[
- \frac{1}{T} \sum_{i=1}^{2^k-1} \sum_{j=i+1}^{2^k} U(i, j) \left[ Mx_i^{(1)} + Mx_j^{(1)} + x_i^{(2)} x_j^{(2)} \right] - \sum_{j=1}^{2^k} \frac{a_0}{T} \left( Mx_j^{(1)} + \frac{x_j^{(2)}^2}{2} \right) - \prod_{j=1}^{2^k} g\left( \frac{a_1}{T} \left( Mx_j^{(1)} + \frac{x_j^{(2)}^2}{2} \right) \right) \left( 1 + O(q^n) \right)
\]
with some $0 < q < 1$ if $x_j \in W_n$, $j = 1, 2, \ldots, 2^k$.

Simple calculation shows that $-\sum_{i=1}^{2^k} U(i, j) = \frac{1-(c/2)^k}{1-c/2}$, hence the coefficient of $x_j^{(1)}$, $(c/2)^k g + \frac{M}{T} \sum_{j=1}^{2^k} U(i, j) - \frac{a_n M}{T}$ equals zero, and

$$h_{n,k}(x_1, \ldots, x_{2^k}) = h_k(x_1, \ldots, x_{2^k})(1 + O(q^n))$$

if $x_j \in W_n$, $j = 1, 2, \ldots, 2^k$, where $h_k$ is defined in (1.12) (with $p = 2$). It is not difficult to see that (7.3) also holds with a random vector with density function (1.12). Hence (7.3) and (7.9) imply Theorem 2.

8. Some open problems and conjectures. Dyson [12] has defined a more general class of models than that considered in this work. He defined, with the help of a real function $\varphi: \mathbb{Z} \to \mathbb{R}^1$, models with the Hamiltonian function

$$\mathcal{H}(\sigma) = -\sum_{i \in \mathbb{Z}} \sum_{j \in \mathbb{Z}, j > i} \varphi(d(i, j))S_n f(x(i))\sigma(j), \quad \sigma = \{\sigma(i), i \in \mathbb{Z}\},$$

where $d(\cdot, \cdot)$ denotes the hierarchical distance on $\mathbb{Z}$ given in formula (1.1) of Part I. In this work we have considered models in the special case $\varphi(x) = |x|^{-a}$ with $a = 2 - \frac{\log c}{\log 2}$. One question we are going to discuss here is that which are the functions $\varphi$ for which Dyson’s model with the Hamiltonian (8.1) has a phase transition at low temperatures. In the boundary case some more delicate phenomena appear which we also want to discuss. The behaviour of vector and scalar-valued models is different. First we discuss the vector-valued case.

The quantities $M_n = M_n(T)$ considered in Part I can be defined in a natural way in the general case. The arguments of Part I suggest that the relation

$$M_{n+1} = M_n - \frac{1}{4^n M_n \varphi(2^n)}$$

holds true. The existence or non-existence of phase transition depends on whether $M = \lim_{n \to \infty} M_n$ equals to zero or not if $T$ is small, i.e. if $M_0$ is large. Hence formula (8.2) suggests that a phase transition at low temperatures occurs if and only if $\sum_{i=1}^{n} \frac{1}{\varphi(2^n)}$ is convergent. Dyson has formulated the same conjecture in [13] and proved its convergent part in the special case when $\sigma(i) \in R^3$. He has also solved the problem for scalar-valued models. He proved that there is a phase transition at low temperatures if $\varphi(n) \geq C \log \log n \log n$ with some $C > 0$, and there is none if $\varphi(n) \log \log n \nrightarrow 0$. Moreover, in the boundary case $\varphi(n) = C \log \log n$ the following Thouless effect occurs: There is some critical parameter $T_{cr}$, such that $M(T) = \lim_{n \to \infty} M_n(T) > 0$ for $T \leq T_{cr}$, and $M(T) = 0$ for $T > T_{cr}$. The quantity $M(T)$ has a physical content, it is called the spontaneous magnetization. The interesting feature of the above result is that it states that the function $M(T)$ has a discontinuity at $T = T_{cr}$. This particular behaviour of the spontaneous magnetization appears only in the boundary case $\varphi(n) = C \log \log n$. On the other hand, the Thouless effect occurs in some other models too, like in the one-dimensional Ising model with $\frac{1}{|x-y|^2}$ interaction, in one-dimensional percolation models if the probability of the event that the points $i$ and $j$ are connected has the order $C(T)|i-j|^{-2}$,
e.t.c.. In recent time several interesting papers appeared on this subject, (see e.g. [1], [2], [17]). On the other hand, there are some other interesting phenomena connected with the Thouless effect, like the irregular behaviour of the correlation function, whose investigation requires essentially new ideas.

The appearance of phase transitions and the Thouless effect in scalar-valued models are connected with the behaviour of the sequence $M_n$. The quantity $M_{n+1}$ can be expressed asymptotically in a simple way with the help of $M_n$ and the function $\varphi$ in scalar-valued models too. But this formula is essentially different from his vector-valued counterpart, namely

$$M_{n+1} \sim M_n \left[ 1 - \exp \left\{ -\frac{1}{T} M_n^2 4^n \varphi(2^n) \right\} \right].$$

In the particular case $\varphi(n) = \frac{\log \log n}{n^2}$ we have

$$n+1 \sim M_n \left[ 1 - \exp \left\{ -\frac{1}{T} M_n^2 \log n \right\} \right].$$

Formula (8.3') may help us to understand the cause of the Thouless effect, at least at a heuristic level. If $M_n(T) < \sqrt{T}$ for some $n$ then relation (8.3) implies that $M(T) = \lim_{n \to \infty} M_n(T) = 0$, hence either $M(T) \geq \sqrt{T}$ or $M(T) = 0$. Since $M(T) \neq 0$ for small $T$, this relation implies the discontinuity of the function $M(T)$. In vector-valued models relation (8.2) does not suggest such a behaviour. We expect however that some delicate effects appear in this case too, and we are going to study them in the future.

Let us remark that the study of existence or non-existence of phase transitions at low temperatures seems to be an essentially simpler problem than the study of the Thouless effect and related questions. In the first problem it is enough to consider sufficiently low temperatures, and in the case of vector-valued models with Hamiltonian function of the form (8.1) for instance the method of the present paper works without any essential changes. In the second problem however, one has to study the behaviour of the model near the critical temperature, and this requires more work and new ideas.

Another problem we are going to discuss here is the description of the large-scale limit of vector-valued equilibrium states with translation invariant Hamiltonian function. We have discussed its scalar-valued counterpart in Section 8 of our paper [6], and formulated our conjectures about it.

Let us consider vector-valued models on the $d$-dimensional integer lattice with Hamiltonian function

$$\mathcal{H}(\sigma) = - \sum_{|i-j|=1} \sigma(i)\sigma(j) - \sum_{i \in \mathbb{Z}^d} p(\sigma(i))$$

with

$$p(x) = -\frac{t}{4} |x|^4 - \frac{|x|^2}{2}, \quad t > 0,$$
and the Lebesgue measure on $R^p$ with some $p \geq 2$ as the free measure of the model. The expression $\sigma(i)\sigma(j)$ in the above formulas denotes scalar product.

Let $\{X(k) = (X^{(1)}_k, \ldots, X^{(p)}_k), k \in \mathbb{Z}^d\}$ be a random field with the distribution of a (pure) equilibrium state with the above Hamiltonian function at a certain temperature $T$. If $d \geq 3$ then there exists a spontaneous magnetization at sufficiently low temperatures. This is proved with the help of the infrared bounds (see e.g. [14]). In case of phase transition we consider that pure state for which the direction of the spontaneous magnetization is $e_1 = (1, 0, \ldots, 0)$, i.e. $E X^{(1)}_k = M > 0$, and $E X^{(s)}_k = 0$ for $s = 2, \ldots, p$.

Define the “renormalized” random fields $\{Y_k(N) = (Y_k(N)^1, \ldots, Y_k(N)^p)\}$, $N = 1, 2, \ldots$ by the formulas

\[
Y_k(N)^1 = A(N)^{-1} \sum_{j \in D_k(N)} (X_j^{(1)} - EX_j^{(1)})
\]

(8.4)

\[
Y_k(N)^s = B(N)^{-1} \sum_{j \in D_k(N)} X_j^{(s)}, \quad s = 2, \ldots, p,
\]

(8.4')

with

\[
D_k(N) = \{j = (j^{(1)}, \ldots, j^{(d)}) \in \mathbb{Z}^d, \quad k^{(r)} N + 1 \leq j^{(r)} \leq (k^{(r)} + 1) N, \quad r = 1, \ldots, d\}.
\]

We are interested in the question that for which choice of $A(N)$ and $B(N)$ the fields $Y_k(N)$ have a non-trivial limit, i.e. a limit which is not concentrated on a single configuration. We also want to describe the distribution of the limit field.

Dyson’s hierarchical model with parameter $c$, $1 < c < 2$ can be considered as an approximation of translation invariant models with nearest neighbour interaction on the $d$-dimensional lattice $\mathbb{Z}^d$ with $d = \frac{2}{1 - \log_2 c}$. (See paper [21] for a discussion of this approximation.) It must be admitted that the above approximation is made only at a heuristic level, but it helps us to get a better understanding about the behaviour of the large-scale limit. On the basis of the present work and [5] we can formulate the following conjectures about the large-scale limit of translation invariant models at low temperatures.

The expression $\exp(i k x)$ in the above formulas denotes the complex exponential. It must be admitted that the above approximation is made only at a heuristic level, but it helps us to get a better understanding about the behaviour of the large-scale limit. On the basis of the present work and [5] we can formulate the following conjectures about the large-scale limit of translation invariant models at low temperatures.

The behaviour of the large-scale limit is different in the cases $d > 4$, $d = 4$ and $d < 4$, and they correspond to the cases $c > \sqrt{2}$, $c = \sqrt{2}$ and $c < \sqrt{2}$ in Dyson’s hierarchical model. In accordance with [5] we expect that for $d > 4$ the large-scale limit exists at low temperatures with $A(N) = N^{d/2}$ and $B(N) = N^{(d+2)/2}$. The limit field $\{Y_k = (Y_k^{(1)}, \ldots, Y_k^{(p)}), k \in \mathbb{Z}^d\}$ is such that the fields $\{Y_k^{(j)}, k \in \mathbb{Z}^d\}$, $j = 1, 2, \ldots, p$ are independent, $\{Y_k^{(1)}, k \in \mathbb{Z}^d\}$ consists of independent identically distributed Gaussian random variables with zero mean, $\{Y_k^{(s)}, k \in \mathbb{Z}^d\}$, $s = 2, \ldots, p$, are massless free Gaussian fields, i.e. they have the same distribution as the field

\[
Y_k = C \int \frac{\exp(i k x)}{|x|} \prod_{j=1}^d \left\{ \frac{\exp(i x^{(j)}) - 1}{ix^{(j)}} \right\} W(dx), \quad k \in \mathbb{Z}^d,
\]
with some \( C > 0 \), where \( W(dx) \) is a complex-valued white noise field on \( R^d \) with the conjugation property \( W(x) = \overline{W(-x)} \). (For the definition of \( W(x) \) see e.g. [16].) The situation is similar in the case \( d = 4 \), i.e. the investigation of Dyson’s model suggests a similar behaviour with the only difference that a logarithmic factor appears in the normalizing term \( A(n) \). More precisely, for \( d = 4 \) the fields \( Y^{(N)}_k \) defined in (8.4) and (8.4') with \( A(N) = N^2 \sqrt{\log N} \) and \( B(N) = N^{(d+2)/2} = N^3 \) have a Gaussian limit as \( N \to \infty \) which consists of independent components \( s = 1, \ldots, p \), similarly to the case \( d > 4 \). The limit of the fields \( Y^{(s)}_k, s = 2, \ldots, p \), is a massless free field.

The result of the present work motivates the following conjecture for \( d = 3 \).

**Conjecture.** For \( d = 3 \) the large-scale limit exists at low temperatures with the normalizations \( A(N) = N^2 \) and \( B(N) = N^{5/2} \). The large-scale limit has the same distribution as the random field \( \{Y_k = (Y^{(1)}_k, \ldots, Y^{(p)}_k), k \in Z^3\} \) defined by the formulas

\[
(8.5) \quad Y^{(s)}_k = C \int \frac{\exp(ikx)}{|x|} \prod_{j=1}^3 \frac{\exp(ix(j))-1}{i x(j)} W_s(dx), \quad k \in Z^3, \ s = 2, \ldots, p
\]

and

\[
(8.6) \quad Y^{(s)}_1 = -\frac{C^2}{2M} \sum_{s=2}^p \iint \frac{\exp(ik(x+y))}{|x||y|} \prod_{j=1}^3 \left\{ \frac{\exp[i(x(j)+y(j))]-1}{i(x(j)+y(j))} \right\} W_s(dx)W_s(dy), \quad k \in Z^3,
\]

where \( C > 0 \) is an appropriate positive constant, and \( W_s(dx), s = 1, \ldots, p \) are independent complex valued white noise fields on \( R^3 \) with the conjugation property \( W(x) = \overline{W(-x)} \). (For the definition of two-fold stochastic integrals with respect to a Gaussian field see e.g. [16]. Such an integral appears in the definition of \( Y^{(s)}_1 \).)

The fields \( Y^{(s)}_s, s = 2, \ldots, p \) defined in (8.5) are massless free fields, the field defined in (8.6) belongs to the class of self-similar fields constructed in Dobrushin’s paper [11]. It is a quadratic functional of a Gaussian field, just as the corresponding field in Dyson’s model for \( 1 < c < \sqrt{2} \). The large-scale limit of the equilibrium state in Dyson’s model described in Theorem 2 of Part II has the following independence property: The random variables \( Y^{(1)}_k + \frac{1}{2M} \sum_{s=2}^p Y^{(s)}_k \) are independent for different \( k \). This independence property does not hold for their translation invariant counterpart defined in (8.5) and (8.6). It cannot be preserved, because translation invariant models have less symmetry. Nevertheless, the following non-rigorous argument shows some analogy between the behaviour of the fields defined in (8.5) and (8.6) and the above mentioned independence property. In this non-rigorous argument we consider the limit field appearing in the Conjecture as the discretization of a generalized field.
Let $\delta(t)$ denote the Dirac-delta function in the point $t$, and consider the generalized field which at $\delta(t)$ takes the value $Y(\delta(t)) = (Y^{(1)}(\delta(t)), \ldots, Y^{(s)}(\delta(t)))$,

$$Y^{(s)}(\delta(t)) = C \int \frac{\exp(itx)}{|x|} W^{(s)}(dx), \quad s = 2, \ldots, p,$$
$$Y^{(1)}(\delta(t)) = -\frac{C^2}{2M} \sum_{s=2}^{p} \int \int \frac{\exp(it(x+y))}{|x||y|} W^{(s)}(dx) W^{(s)}(dy).$$

Actually this definition is not correct, since the above stochactic integrals are meaningless because of the divergence of the integrals $\int_{R^3} \frac{dx}{|x|^2}$ and $\int_{R^3} \frac{dx\,dy}{|x|^2|y|^2}$. But the integral

$$Y^{(1)}(\varphi) = \int Y^{(1)}(\delta(t))\varphi(t) dt = -\frac{C^2}{2M} \sum_{s=2}^{p} \int \int \frac{\hat{\varphi}(x+y)}{|x||y|} W^{(s)}(dx) W^{(s)}(dy),$$

and

$$Y^{(s)}(\varphi) = \int Y^{(s)}(\delta(t))\varphi(t) dt = C \int \frac{\hat{\varphi}(x)}{|x|} W^{(s)}(dx), \quad 2 \leq s \leq p,$$

are meaningful for nice functions $\varphi$. In particular, they are meaningful for the indicator functions of the unit cubes $I_{k} = [k_i, k_i + 1)$ which we denote by $\varphi_k$ if $k = (k_1, k_2, k_3)$. The random field appearing in the Conjecture can be considered as the discretization of the above defined generalized field if we identify $Y_k$ with $Y(\varphi(k))$.

A formal application of Itô’s formula (see e.g. [16]) would supply the relation $Y^{(1)}(\delta(t)) - \frac{1}{2M} \sum_{s=2}^{p} Y^{(s)}(\delta(t))^2 = const.$, and this can be considered as the analogue of the independence property of the large-scale limit of Dyson’s model for the above defined generalized field $Y(\cdot)$. On the other hand by our Conjecture the discretization of this generalized field is the large-scale limit of the three-dimensional translation invariant vector-valued model at low temperatures.

Let us finally discuss the cases $d = 1$ and $d = 2$. The case $d = 1$ is rather simple. In this case there is no phase transition, and if $\{X_k, k \in \mathbb{Z}^1\}$ is a random field with the distribution function of an equilibrium state at any temperature then it satisfies the central limit theorem with the usual normalization. The case $d = 2$ is more delicate. In this case the dimension $p$, $(\sigma(i) \in R^p)$, also plays an important role. In this case there is no symmetry breaking, but for $d = 2, \ p = 2$ a more delicate phenomenon, the so-called Kosterlitz–Thouless effect occurs. (See [15]). This means that at low temperatures the correlation function decreases rather slowly, only power-like. Hence a non-trivial large-scale limit should appear in this case. For $d = 2, \ p \geq 3$ it is expected that the Kosterlitz–Thouless effect does not occur, but for the time being it is proved only at a physical level (see[18]). Hence, it is expected that for $d = 2, \ p > 2$ the large-scale limit has the same (trivial) behaviour as for $d = 1$.

When typing the final version of this work the authors learned about some recent results about the Kosterlitz–Thouless effect (see [22], [23]). The arguments of these works, also supported by computer simulation, suggest that the situation in two-dimensional translation invariant models is essentially different from what we had expected. In particular, the difference between the cases $p = 2$ and $p > 2$ in the models we have discussed at the end of this Section does nevertheless not occur.
Appendix

Appendix A. The proof of the basic recursive relations (2.1) and (2.1') in Part I. Formula (2.1') immediately follows from (1.4) in Part I with \( n = 0 \). To prove (2.1) let us first observe that the recursive relation

\[
H_{n+1}(x_1, \ldots, x_{2n+1}) = H_n(x_1, \ldots, x_{2n}) + H_n(x_{2n+1}, \ldots, x_{2n+1}) - c^n \left( 2^{-n} \sum_{i=1}^{2n} x_i \right) \left( 2^{-n} \sum_{j=2^{n+1}+1}^{2^{n+1}} x_j \right)
\]

(A1)

holds for \( n \geq 0 \), where

\[
H_n(x_1, \ldots, x_{2n}) = - \sum_{i=1}^{2n} \sum_{j=i+1}^{2n} U(i, j) x_i x_j ,
\]

and \( U(i, j) \) is defined by (1.1) and (1.2') in Part I. By relation (1.4) in Part I

\[
p_{n+1}(x, T) = \frac{1}{Z_{n+1}(T, t)} \int \exp \left\{ -\frac{1}{T} H_n(x_1, \ldots, x_{2n+1}) \right\} \delta \left( 2^{-(n+1)} \sum_{i=1}^{2^{n+1}} x_i - x \right) \prod_{i=1}^{2^{n+1}} p(x_i) \, dx_i ,
\]

(A2)

where \( Z_{n+1}(T, t) \) is an appropriate norming constant, and \( \delta (2^{-(n+1)} \sum_{i=1}^{2^{n+1}} x_i - x) \) means that integration in (A2) is taken on the hyperplane \( 2^{-(n+1)} \sum_{i=1}^{2^{n+1}} x_i = x \) with respect to the Lebesgue measure. Let us fix some number \( u \), and calculate the integral on the right-hand side of (A2) by integrating first on the hyperplane defined by the relations \( 2^{-n} \sum_{i=1}^{2^n} x_i = x + u \) and \( 2^{-n} \sum_{i=2^n+1}^{2^{n+1}} x_i = x - u \) and then by integrating by \( u \). We get with the help of relations (A1) and (A2) that

\[
p_n(x, T) = \frac{1}{Z_{n+1}(T, t)} \int \exp \left\{ \frac{c^n}{T} (x + u)(x - u) \right\} \left[ \int \exp \left\{ -\frac{1}{T} H_n(x_1, \ldots, x_{2n}) \right\} \delta \left( 2^{-n} \sum_{i=1}^{2^n} x_i - (x + u) \right) \prod_{i=1}^{2^n} p(x_i) \, dx_i \right] \left[ \int \exp \left\{ -\frac{1}{T} H_n(x_{2n+1}, \ldots, x_{2n+1}) \right\} \delta \left( 2^{-n} \sum_{i=2^n+1}^{2^{n+1}} x_i - (x - u) \right) \prod_{i=2^n+1}^{2^{n+1}} p(x_i) \, dx_i \right] \, du
\]

\[= C_n \int \exp \left\{ \frac{c^n}{T} (x^2 - u^2) \right\} p_n(x + u)p_n(x - u) \, du ,\]

as we have claimed.
Appendix B. The proof of Theorem C. Since the measure $\mu_{n+k}$ has the density function

$$\frac{1}{Z_{n+k}} \exp \left\{ -\frac{1}{T} H_{n+k}(z_1, \ldots, z_{2^{n+k}}) \right\} \prod_{i=1}^{2^{n+k}} p(z_i),$$

the density function of the measure $Q_n \mu_{n+k}$, the function $h_k(x_1, \ldots, x_{2^k})$ equals to

\begin{equation}
\tag{B1}
h_k(x_1, \ldots, x_{2^k}) = \frac{1}{Z_{n+k}} \int \exp \left\{ -\frac{1}{T} H_{n+k}(z_1, \ldots, z_{2^{n+k}}) \right\} \prod_{l=1}^{2^k} \delta \left( 2^{-n} \sum_{j=(l-1)2^{n+1}}^{l2^n} z_j - x_l \right) \prod_{i=1}^{2^{n+k}} p(z_i) \, dz_i,
\end{equation}

where $\prod_{l=1}^{2^k} \delta (2^{-n} \sum_{j=(l-1)2^{n+1}}^{l2^n} z_j - x_l)$ in the integral (B1) means that integration is taken on the hyperplane defined by the relations $2^{-n} \sum_{j=(l-1)2^{n+1}}^{l2^n} z_j = x_l$, $l = 1, \ldots, 2^k$, with respect to the Lebesgue measure. The special structure of the hierarchical distance implies that

$$H_{n+k}(z_1, \ldots, z_{2^{n+k}}) = \sum_{l=1}^{2^k} H_n(z_{(l-1)2^{n+1}+1}, \ldots, z_{l2^n}) - \sum_{i=1}^{2^k} \sum_{j=i+1}^{2^k} c^n U(i, j) \bar{z}_i \bar{z}_j,$$

with $\bar{z}_i = 2^{-n} \sum_{p=(i-1)2^{n+1}}^{i2^n} z_p$, $i = 1, \ldots, 2^k$.

Hence relation (B1) can be rewritten as

$$h_k(x_1, \ldots, x_{2^k}) = \frac{1}{Z_{n+k}} \exp \left\{ \frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^{2^k} c^n U(i, j) x_i x_j \right\} \prod_{l=1}^{2^k} \int \exp \left\{ -\frac{1}{T} H_n(z_{(l-1)2^{n+1}+1}, \ldots, z_{l2^n}) \right\} \delta \left( 2^{-n} \sum_{j=(l-1)2^{n+1}}^{l2^n} z_j - x_l \right) \prod_{j=(l-1)2^{n+1}}^{l2^n} p(z_j) \, dz_j$$

$$= C_{k, n} \exp \left\{ -\frac{1}{T} H_k(c^{n/2} x_1, \ldots, c^{n/2} x_{2^k}) \right\} \prod_{l=1}^{2^k} p_n(x_l, T),$$

as we have claimed.

Appendix C. The calculation of the Radon–Nikodym derivatives. The proof of formulas (2.1)—(2.3') in Part II. For $n = N$ relations (2.1) and (2.2)
of Part II immediately follow from formula (1.4) in Part II. Hence it is enough to prove our relations by induction from \( n + 1 \) to \( n \). Clearly,

\[
P_{n+1}(x_1, \ldots, x_{2^{n+1}}) = C_n P_n(x_1, \ldots, x_{2^n}) P_n(x_{2^n+1}, \ldots, x_{2^{n+1}}) \exp \left\{ \frac{e^n}{T} \left( 2^{-n} \sum_{j=1}^{2^n} x_j \right) \left( 2^{-n} \sum_{j=2^n+1}^{2^{n+1}} x_j \right) \right\}
\]

with some norming constant \( C_n \). Given some measurable set \( A \subset (\mathbb{R}^p)^{2^n} \) define the cylindrical set \( \tilde{A} \subset (\mathbb{R}^p)^{2^{n+1}} \) as \( \tilde{A} = A \times (\mathbb{R}^p)^{2^n} \). By our inductive hypothesis for \( n + 1 \)

\[
\mu^{h, N}_{n,N}(A) = \int_{\tilde{A}} f^{h, N}_{n+1,N} \left( 2^{-(n+1)} \sum_{j=1}^{2^{n+1}} x_j \right) P_{n+1}(x_1, \ldots, x_{2^{n+1}}) \, dx_1 \ldots dx_{2^{n+1}}
\]

\[
= C_n \int_{\tilde{A}} f^{h, N}_{n+1,N} \left( 2^{-n} \left( \sum_{j=1}^{2^n} \frac{x_j}{2} + \sum_{j=2^n+1}^{2^{n+1}} \frac{x_j}{2} \right) \right) P_n(x_1, \ldots, x_{2^n}) P_n(x_{2^n+1}, \ldots, x_{2^{n+1}}) \exp \left\{ \frac{e^n}{T} \left( 2^{-n} \sum_{j=1}^{2^n} x_j \right) \left( 2^{-n} \sum_{j=2^n+1}^{2^{n+1}} x_j \right) \right\} \, dx_1 \ldots dx_{2^{n+1}}.
\]

Let us calculate the last integral by first integrating on the hyperplanes where \( x_1, \ldots, x_{2^n} \) and \( y = \frac{1}{2^n} \sum_{j=2^n+1}^{2^{n+1}} x_j \) are fixed. Since \( P_n(x_{2^n+1}, \ldots, x_{2^{n+1}}) \) is the only term in the integrand which is not constant on such a hyperplane, and its integral equals \( p_n(y) \) on it, we get that

\[
\mu^{h, N}_{n,N}(A) = C_n \int_{A \times \mathbb{R}^p} f^{h, N}_{n+1,N} \left( 2^{-n} \left( \sum_{j=1}^{2^n} \frac{x_j}{2} + \frac{y}{2} \right) \right) P_n(x_1, \ldots, x_{2^n}) \exp \left\{ \frac{e^n}{T} \left( 2^{-n} \sum_{j=1}^{2^n} x_j \right) y \right\} \, dx_1 \ldots dx_{2^n} \, dy.
\]

Hence we get, by integrating first by the variable \( y \) that

\[
\mu^{h, N}_{n,N}(A) = C_n' \int_A S_n f^{h, N}_{n+1,N} \left( 2^{-n} \sum_{j=1}^{2^n} x_j \right) P_n(x_1, \ldots, x_{2^n}) \, dx_1 \ldots dx_{2^n}.
\]

Since this relation holds for all measurable sets \( A \subset (\mathbb{R}^p)^{2^n} \), it implies our inductive hypothesis for \( n \).

**Appendix D. On limit Gibbs states.** Here we briefly describe the definition of limit Gibbs states (also called equilibrium states in the literature,) and discuss some important questions related to this definition. Limit Gibbs states are defined
with the help of a Hamiltonian (often called energy) function, a free measure and a physical parameter, the temperature $T$. The Hamiltonian function is a formal series. Let us have a subset $Z \subseteq Z^d$ of the $d$-dimensional integer lattice and a closed set $K \subseteq \mathbb{R}^p$ in the $p$-dimensional Euclidean space. We consider a Hamiltonian function $H(\sigma)$ of the form

$$H(\sigma) = -\sum_{i,j \in Z} U(i,j)\sigma(i)\sigma(j), \quad \sigma = \{\sigma(j), \sigma(j) \in K, j \in Z\},$$

where $U(\cdot, \cdot)$ is a given function, $U : \mathbb{Z} \times \mathbb{Z} \to \mathbb{R}^1$, and $\sigma(i)\sigma(j)$ denotes scalar product. (There is a more general definition of Hamiltonian functions, but this special class is sufficiently large for our purposes.) Given some finite set $V \subseteq \mathbb{Z}$, we define the energy function $H_V(\sigma)$ as

$$(D1) \quad H_V(\sigma) = -\sum_{i,j \in V} U(i,j)\sigma(i)\sigma(j), \quad \sigma = \{\sigma(i), i \in V\},$$

and the conditional energy in $V$ with respect to a configuration $\bar{\sigma}$ in $\mathbb{Z} - V$ as

$$(D2) \quad H_V(\sigma|\bar{\sigma}) = H_V(\sigma) - \sum_{i \in V} \sum_{j \in \mathbb{Z} - V} U(i,j)\sigma(i)\bar{\sigma}(j),$$

$$\sigma = \{\sigma(i), i \in V\}, \quad \bar{\sigma} = \{\bar{\sigma}(i), i \in \mathbb{Z} - V\},$$

provided that the last sum is convergent. Given some $h \in \mathbb{R}^p$, we also introduce the energy of a configuration $\sigma$ in the volume $V$ with respect to the external field $h$ as

$$(D3) \quad H_V^h(\sigma) = H_V(\sigma) - h \sum_{i \in V} \sigma(i).$$

Given some finite set $V \subseteq \mathbb{Z}$, a configuration $\bar{\sigma} = \{\bar{\sigma}(j), j \in \mathbb{Z} - V\}$ outside $V$, a Hamiltonian function $H(\sigma)$ and a free measure $P(dx)$ on $K$, we define the Gibbs measure in volume $V$ with respect to the external field $\bar{\sigma}$ at temperature $T$ as the probability measure $\mu_{V,T}(\cdot|\bar{\sigma})$ on $K^V$ given by the formula

$$(D4) \quad \mu_{V,T}(\sigma \in A|\bar{\sigma}) = \frac{1}{Z(V,T,\bar{\sigma})} \int_A \exp\left\{-\frac{1}{T}H_V(\sigma|\bar{\sigma})\right\} \prod_{j \in V} P(d\sigma(j)),$$

where $A \subseteq K^V$ is an arbitrary measurable set, $\sigma = \{\sigma(j), j \in V\}$ and $Z(V,T,\bar{\sigma})$ is an appropriate norming constant, provided that the above expression is meaningful. Now we formulate the following

**Definition of Gibbs states.** A probability measure $\tilde{\mu}$ is a Gibbs state with Hamiltonian function $H$ and free measure $P$ at temperature $T$ if a $\tilde{\mu}$ distributed random field $\sigma(j), j \in \mathbb{Z}$ satisfies the following relation: For any finite set $V$ and measurable set $A \subseteq K^V$ the conditional probability of the event $\sigma \in A$, $\sigma = \{\sigma(j), j \in V\}$ with
respect to the condition \( \{ \sigma(j) = \bar{\sigma}, j \in \mathbb{Z} - V \} \) with a configuration \( \bar{\sigma} = \{ \bar{\sigma}(j), j \in \mathbb{Z} - V \} \) equals to

\[
\bar{\mu}(A \mid \{ \sigma(j), j \in \mathbb{Z} - V \} = \bar{\sigma}) = \mu_{V,T}(A \mid \bar{\sigma})
\]

with \( \bar{\mu} \) probability one, where \( \mu_{V,T} \) is defined in (D4).

The question arises whether Gibbs states on \( K^\mathbb{Z} \) exist, and whether they are unique. A natural way to construct Gibbs states is to carry out the following procedure. Choose an increasing family of sets \( V_n \subset \mathbb{Z}, \cup V_n = \mathbb{Z} \), fix a configuration \( \bar{\sigma} = \bar{\sigma}(n) = \{ \bar{\sigma}(j), j \in \mathbb{Z} - V_n \} \) for each \( V_n \), and consider the measures \( \mu_{V_n,T}(\cdot \mid \bar{\sigma}) \) defined in (D4). Prove that under some mild restrictions there is a convergent subsequence of this sequence, and the limit of this subsequence is a Gibbs state.

The problem about the uniqueness of Gibbs states is closely related to the question whether, in dependence of the choice of the external configuration \( \bar{\sigma}(n) \), different limits can appear in the above construction. A slightly different, and often useful approach is to choose a sequence \( h_n \in \mathbb{R}^n, h_n \to 0 \), and try to construct Gibbs states as the limit of a sequence of measures of the form \( \mu_{V_n,T}^h \), where we define the probability measure \( \mu_{V,T}^h \) as

\[
\mu_{V,T}^h(A) = \frac{1}{Z(V,T,h)} \int_A \exp \left\{ -\frac{1}{T} \left( \sum_{i \in V} \sum_{j \in V} U(i,j)x_i x_j - h \sum_{i \in V} x_i \right) \right\} \prod_{i \in V} P(dx_i).
\]

If \( K \) is a compact subset of \( \mathbb{R}^p \) then standard results in probability theory imply the compactness of the measures \( \mu_{V_n,T}(\cdot \mid \bar{\sigma}) \) or of \( \mu_{V_n,T}^h \) in weak topology, i.e. the existence of a convergent subsequence in this topology. (See e.g. [3].) Nevertheless, there are many interesting models, where the set \( K \) is non-compact (e.g. \( K = \mathbb{R}^p \)) and in such cases a hard analysis is needed to prove the existence of such a convergent subsequence. (See e.g. [10] or [20] as an example.) In order to prove that the limit of the sequence of measures \( \mu_{V_n,T}(\cdot \mid \bar{\sigma}) \) (or \( \mu_{V_n,T}^h \)) is really a Gibbs state it is worth while to rewrite the definition of Gibbs states in an equivalent integral form. Let \( f = f(x_1, \ldots, x_{l_k}) \) and \( g = g(x_1, \ldots, x_{l_k}) \), \( x_i \in \mathbb{R}^p \), be two bounded and continuous functions with finitely many arguments, \( V = \{ j_1, \ldots, j_k \} \), \( W = \{ l_1, \ldots, l_k \} \), \( V \subset \mathbb{Z}, W \subset \mathbb{Z} \) such that \( V \cap W = \emptyset \).

The measure \( \bar{\mu} \) on \( K^\mathbb{Z} \) is a Gibbs state if and only if

\[
\int_{K^\mathbb{Z}} fg d\bar{\mu} = \int_{K^{\mathbb{Z} - V}} \mu(f)g d\bar{\mu}
\]

for all functions \( f \) and \( g \) with the above properties, where

\[
\mu(f) = \mu(f)(\bar{\sigma}) = \int_{K^V} f(\sigma(j_1), \ldots, \sigma(j_k)) \mu_{V,T}(d\sigma | \bar{\sigma})
\]

\[
\sigma = \{ \sigma(j), j \in V \}, \quad \bar{\sigma} = \{ \sigma(j), j \in \mathbb{Z} - V \},
\]

and \( \mu_{V,T} \) is defined in (D4).
Let us consider an arbitrary sequence of sets \( V_n \subset \mathbb{Z} \), \( \cup V_n = \mathbb{Z} \), and numbers \( h_n \in \mathbb{R}^p \), \( h_n \to 0 \). Some calculation shows that for sufficiently large \( n \) (if \( V \subset V_n \), \( W \subset V_n \))

\[
\int_{K_{V_n}} fg \, d\mu^{h_n}_{V_n,T} = \int_{K_{V_n} - V} \mu^{h_n}(f) g \, d\mu^{h_n}_{V_n,T}
\]

with

\[
\mu^{h_n}(f) = \mu^{h_n}_{V_n,T}(f; \sigma(j), j \in V_n - V)
\]

\[
\begin{align*}
\int f(\sigma(j_1), \ldots, \sigma(j_k)) \exp\left\{-\frac{1}{T}(H_{V_n}(\sigma) - h_n \sum_{i \in V} \sigma(i))\right\} \prod_{i \in V} P(\, d\sigma(i) )
\end{align*}
\]

\[
\int \exp\left\{-\frac{1}{T}(H_{V_n}(\sigma) - h_n \sum_{i \in V} \sigma(i))\right\} \prod_{i \in V} P(\, d\sigma(i) )
\]

where

\[
H_{V_n}(\sigma) = \sum_{i \in V} \sum_{j \in V_n} \sigma(i) \sigma(j).
\]

If the sequence \( \mu^{h_n}_{V_n,T} \) tends weakly to the measure \( \bar{\mu} \) then the left-hand side of (D8) converges to that of (D6). Hence to prove that the limit measure \( \bar{\mu} \) is a Gibbs state it suffices to establish the convergence of the right-hand side of (D8) to that of (D6). If the Hamiltonian function has a finite range interaction, i.e. there is some number \( r \) such that \( U(i,j) = 0 \) if \( |i - j| \geq r \) then it is not difficult to see that \( \mu^{h_n}(f)(\bar{\sigma}) \to \mu(f)(\bar{\sigma}) \), and the required convergence can be proved with the help of this relation. In case of infinite range interaction one must be more careful, especially if the state space \( K \) is non-compact. Dyson’s model which we are investigating is such a model. In Theorem 1 of Part II we have proved the weak convergence of the measures \( \mu^{h_n}_N \) to \( \bar{\mu} \). (Actually, we have proved a stronger form of convergence.) In Appendix E we prove Theorem B, i.e. we show that the limit measure is a Gibbs state. In the proof we approximate Dyson’s model with a model with finite range interaction, and this enables us to carry out the required limiting procedure. In Appendix E we restrict ourselves to Dyson’s model, although the argument also works in more general cases.

**Appendix E. The proof of Theorem B.** We apply the argument of Appendix D. The proof of Theorem B can be completed by showing that also in the case of Dyson’s model the right-hand side of (D8) tends to that of (D6). We formulate this statement in more detail.

It suffices to consider the case when \( V = \{1, 2, \ldots, 2^k\} \), \( W = \{2^k + 1, 2, \ldots, 2^m\} \) with some \( 0 \leq k < m \), i.e. \( f = f(x_1, \ldots, x_{2^k}) \), \( g = g(x_{2^k+1}, \ldots, x_{2^m}) \) and \( V_N = \{1, 2, \ldots, 2^N\} \). (We apply the notation of Appendix D.) Introduce the functions

\[
\begin{align*}
p_{k,N}(x_1, \ldots, x_{2^k}) &= \exp\left\{-\frac{1}{T} \left( \sum_{i=1}^{2^k} \sum_{j=i+1}^{2^N} U(i,j)x_i x_j - h \sum_{i=1}^{2^k} x_i \right) \right\}
\end{align*}
\]

\[
\int \exp\left\{-\frac{1}{T}(\sum_{i=1}^{2^k} \sum_{j=i+1}^{2^N} U(i,j)x_i x_j - h \sum_{i=1}^{2^k} x_i) \right\} \prod_{i=1}^{2^k} p(x_i) \, dx_i,
\]

\( x_j \in \mathbb{R}^p, \ j = 1, \ldots, 2^N \)
and

\[(E1') \quad p_k(x_1, \ldots, x_{2^k} | x_{2^k+1}, \ldots) = \frac{\exp\left\{ -\frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^\infty U(i, j) x_i x_j \right\}}{\int \exp\left\{ -\frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^\infty U(i, j) x_i x_j \right\} \prod_{i=1}^{2^k} p(x_i) \, dx_i}, \]

where the function $U(\cdot, \cdot)$ is defined in (1.2) and $p(x)$ in (1.3) of Part I. Put

\[(E2) \quad \mu_{k,N}^h(f)(x_{2^k+1}, \ldots, x_{2^N}) = \int f(x_1, \ldots, x_{2^k}) p_k(x_1, \ldots, x_{2^k} | x_{2^k+1}, \ldots, x_{2^N}) \prod_{i=1}^{2^k} p(x_i) \, dx_i \]

and

\[(E2') \quad \mu_k(f)(x_{2^k+1}, \ldots) = \int f(x_1, \ldots, x_{2^k}) p_k(x_1, \ldots, x_{2^k} | x_{2^k+1}, \ldots) \prod_{i=1}^{2^k} p(x_i) \, dx_i.\]

The convergence of the right hand side of (D8) to that of (D6) is equivalent to the relation

\[(E3) \quad \lim_{N \to \infty} \int g \mu_{k,N}^h(f) \, d\mu_{k,N}^h = \int g \mu_k(f) \, d\bar{\mu}\]

in our case, where $\mu_{k,N}^h$ and $\bar{\mu}$ are the same probability measures on $(R^p)^{2^N}$ and $(R^p)^Z$ as in Theorem 1 of Part I.

To prove this relation let us introduce the sets $A(K, k, n, N)$ and $A(K, k, n)$, where $K \in R^1$, $K > 0$, $k, n, N \in Z$ and $k < n < N$, defined by the formulas

\[A(K, k, n, N) = \{(x_{2^k+1}, \ldots, x_{2^N}), x_j \in R^p, j = 2^k + 1, \ldots 2^N, |x_j| < K\]

if $2^k < j \leq 2^n$, and $|x_j| < 2^{l\alpha}$ if $2^l < j \leq 2^{l+1}$, $l = n, \ldots, N - 1\}

and

\[A(K, k, n) = \{(x_{2^k+1}, \ldots), x_j \in R^p, j = 2^k + 1, \ldots, |x_j| < K\]

if $2^k < j \leq 2^n$, and $|x_j| < 2^{l\alpha}$ if $2^l < j \leq 2^{l+1}$, $l = n, n + 1, \ldots\},

where $\alpha = \frac{3}{4} - \frac{1}{2} \log e$. We claim that for all $\varepsilon > 0$ some $n = n(\varepsilon)$ and $K = K(\varepsilon, n)$ can be chosen in such a way that

\[(E4) \quad \mu_{k,N}^h((x_{2^k+1}, \ldots, x_{2^N}) \notin A(K, k, n, N)) < \varepsilon\]
and

(E4′) \[ \bar{\mu}(x_{2k+1}, \ldots) \not\in A(K,k,n) < \varepsilon. \]

To prove relations (E4) and (E4′) let us first observe that there is a universal constant \( C, C > 0 \), such that if \( S_n f(x)j \) is the \( j \)-th coordinate of a \( \mu_{N}^{hN} \) or \( \bar{\mu} \) distributed random variable then the inequality \( E\sigma^{2}(j) < C \) holds for all \( j \in \mathbb{Z} \) and measures \( \mu_{N}^{hN} \) and \( \bar{\mu} \). This can be seen by observing that the argument of Section 6 in Part II actually implies that all moments of the \( j \)-th coordinate \( \sigma(j) \) of a \( \mu_{N}^{hN} \) distributed random vector converge to the corresponding moment of the \( j \)-th coordinate of a \( \bar{\mu} \) distributed random vector as \( N \to \infty \). Then we get, by exploiting that \( 1 - 2\alpha > 0 \) that

\[ \mu_{N}^{hN} \{ (x_{2k+1}, \ldots, x_{2N}) \not\in A(K,k,n) \} \leq C \left( \frac{2^{n} - 2^{k}}{K^{2}} + \sum_{j=n}^{N} 2^{j(1 - 2\alpha)} \right) < \varepsilon \]

if first \( n \) and then \( K \) is chosen sufficiently large. The proof of (E4′) is the same. (We remark that relations (E4) and (E4′) hold with arbitrary \( \alpha > 0 \) in the definition of the sets \( A(\cdot, \cdot, \cdot) \). To prove it we have to apply the stronger statement \( E\sigma^{2k} < C_{k} \) for all \( k \geq 1 \). This observation is needed if we want to prove Theorem B in the case \( \sqrt{2} < c < 2 \) too.)

We claim that for all \( \varepsilon > 0 \) there is some \( N_{0} = N_{0}(\varepsilon, K, n) \) such that for \( N > N_{0} \)

(E5)

\[ \left| p_{k,N}^{hN}(x_1, \ldots, x_{2k} | x_{2k+1} \ldots, x_{2N}) - p_{k,N_0}^{hN}(x_1, \ldots, x_{2k} | x_{2k+1} \ldots, x_{2N_0}) \right| < \varepsilon \left( \sum_{i=1}^{2^k} |x_i| \right) \exp \left\{ (2Kn + 1) \sum_{i=1}^{2^k} |x_i| \right\} \]

and

(E5′)

\[ \left| p_{k}(x_1, \ldots, x_{2k} | x_{2k+1} \ldots) - p_{k,N_0}^{0}(x_1, \ldots, x_{2k} | x_{2k+1} \ldots, x_{2N_0}) \right| < \varepsilon \left( \sum_{i=1}^{2^k} |x_i| \right) \exp \left\{ (2Kn + 1) \sum_{i=1}^{2^k} |x_i| \right\} \]

if \( (x_{2k+1}, \ldots, x_{2N}) \in A(K,k,n,N) \) and \( (x_{2k+1}, \ldots) \in A(K,k,n) \). (The constants \( K \) and \( n \) in formulas (E5) and (E5′) are the same as in the definition of the sets \( A(K,k,n,N) \).) First we show that (E5) and (E5′) together with (E4) and (E4′) imply (E3), hence also Theorem B. Indeed, since \( p_{n}(x) \) decreases at infinity faster than \( \exp(-x^2/2) \) hence (E5) and (E5′) imply that

\[ \left| \mu_{k,N}^{hN}(f)(x_{2k+1}, \ldots, x_{2N}) - \mu_{k,N_0}^{0}(f)(x_{2k+1}, \ldots, x_{2N_0}) \right| < \text{const.} \varepsilon \]
and
\[ | \mu_k(f)(x_{2k+1}, \ldots) - \mu_{k,N_0}^0(f)(x_{2k+1}, \ldots, x_{2N_0}) | < \text{const.} \varepsilon. \]

if \((x_{2k+1}, \ldots) \in A(K, k, n)\). This implies that \((x_{2k+1}, \ldots, x_{2N}) \in A(K, k, n, N)\). Since this relation holds on a set of \(1 - \varepsilon\) \(\mu_N^h\) resp. \(\bar{\mu}\) probability by (E4) and (E4'), the functions \(f, g, \mu_{k,N}^h(f)\) and \(\mu(f)\) are bounded, hence an error less than \(\text{const.} \varepsilon\) is committed if \(\mu_{k,N}^h\) and \(\mu_k(f)\) is replaced by \(\mu_N^0\) and \(\mu(f)\) in formula (E3). After this replacement relation (E3) holds, because the projections of the measures \(\mu_{k,N}^h\) to \((R^n)^{2N}\) converge to the projection of \(\bar{\mu}\) to the same subspace. Since \(\varepsilon > 0\) can be chosen arbitrary small, relation (E3) holds in its original form.

We prove only (E5) the proof of (E5') being the same. Let us first observe that for any \(\eta > 0\) there is some \(N_0 = N_0(K, k, n, \eta)\) such that for \(N \geq N_0\) and \((x_{2k+1}, \ldots, x_{2N}) \in A(K, k, n, N)\)

(E6)

\[
\exp \left\{ -\frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^{2^N} U(i, j)x_i x_j - h_N \sum_{i=1}^{2^k} x_i \right\} - \exp \left\{ -\frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^{2^{N_0}} U(i, j)x_i x_j \right\} - 1 \leq \eta \sum_{i=1}^{2^k} \frac{x_i}{T} \exp \left\{ \frac{2Kn + \eta}{T} \sum_{i=1}^{2^k} |x_i| \right\}.
\]

In the last relation we have applied the inequality \(|e^x - 1| \leq |x|e^{|x|}\) together with the relations \(h_N < \eta/2\), \(\sum_{j=2^{N_0}+1}^{2^N} U(i, j)x_j \leq \sum_{j=N_0}^{\infty} 2^{j(\alpha - 1)} (\frac{\eta}{T})^j < \eta/2\) and \(\sum_{j=i+1}^{2^N} U(i, j)x_j \leq 2Kn\) if \(N > N_0\), \(j > 2^k\), \((x_{2k+1}, \ldots, x_{2N}) \in A(K, k, n, N)\) and \(N_0\) is sufficiently large. Integrating inequality (E6) with respect to the measure \(\prod_{i=1}^{2^k} p(x_i)\ dx_i\) we get that

(E7)

\[
\int \exp \left\{ -\frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^{2^N} U(i, j)x_i x_j - h_N \sum_{i=1}^{2^k} x_i \right\} \prod_{i=1}^{2^k} p(x_i)\ dx_i
\]

\[ - \int \exp \left\{ -\frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^{2^{N_0}} U(i, j)x_i x_j \right\} \prod_{i=1}^{2^k} p(x_i)\ dx_i < \text{const.} \eta,\]

where the \(\text{const.}\) may depend on \(K\) and \(n\). In formulas (E6) and (E7) we have shown that both the numerators and the denominators of the functions \(p_{k,N}^h\) and
$p_{k,N_0}^0$ defined in (E1) are close to each other. The number $\eta$ can be chosen arbitrary small in these estimates by fixing first $n$ then $K = K(n)$ and finally $N_0 = N_0(K,n)$ in an appropriate way. Moreover, given some appropriately chosen $n$ and $K$ the number $\eta > 0$ can be taken arbitrary small if $N_0 = N_0(K,n,\eta)$ is sufficiently large. Hence we prove (E5) by showing that

$$\int \exp\left\{-\frac{1}{T} \sum_{i=1}^{2^k} \sum_{j=i+1}^{2N_0} U(i,j)x_ix_j \right\} \prod_{i=1}^{2^k} p(x_i) \, dx_i > D$$

with some $D > 0$ on the set $A(K,k,n,N_0)$, i.e. the integral in (E8) is separated from zero. Here the constant $D$ may depend on $K$ and $n$ but not on $N_0$. Relation (E8) holds, since if $|x_j| < 1$, $j = 1, 2, \ldots, 2^k$ and $(x_{2^k+1}, \ldots, x_{2N}) \in A(K,k,n,N_0)$ then the integrand in (E8) is separated from zero.

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4. Some preparatory remarks to the proof of Proposition 1$'$.
5. The proof of Proposition 1$'$.
6. The proof of Theorem 1 and Proposition 2. Existence of the thermodynamical limit.
7. The proof of Theorem 2. Existence of the large-scale limit.
8. Some open problems and conjectures.
Appendix.
A. The proof of the basic recursive relations (2.1) and (2.1′) in Part I.
B. The proof of Theorem C.
C. The calculation of the Radon–Nikodym derivatives. The proof of formulas (2.1)—(2.3′) in Part II.
D. On limit Gibbs states.
E. The proof of Theorem B.