

Metastability for the dilute Curie–Weiss model with Glauber dynamics*

Anton Bovier[†] Saeda Marello[‡] Elena Pulvirenti[§]

Abstract

We analyse the metastable behaviour of the dilute Curie–Weiss model subject to a Glauber dynamics. The model is a random version of a mean-field Ising model, where the coupling coefficients are Bernoulli random variables with mean $p \in (0, 1)$. This model can be also viewed as an Ising model on the Erdős–Rényi random graph with edge probability p . The system is a Markov chain where spins flip according to a Metropolis dynamics at inverse temperature β . We compute the average time the system takes to reach the stable phase when it starts from a certain probability distribution on the metastable state (called the last-exit biased distribution), in the regime where $N \rightarrow \infty$, $\beta > \beta_c = 1$ and h is positive and small enough. We obtain asymptotic bounds on the probability of the event that the mean metastable hitting time is approximated by that of the Curie–Weiss model. The proof uses the potential theoretic approach to metastability and concentration of measure inequalities.

Keywords: metastability; Glauber dynamics; randomly dilute Curie–Weiss model; Erdős–Rényi random graph.

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1 Introduction and main results

The randomly dilute Curie–Weiss model (RDCW) is a classical model of a disordered ferromagnet and was studied, e.g. in Bovier and Gayraud [6]. It generalises the standard Curie–Weiss model (CW) in that the fixed interactions between each pair of spins is replaced by independent, identically distributed, random ferromagnetic couplings between

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[†]Institut für Angewandte Mathematik, Rheinische Friedrich-Wilhelms-Universität Bonn, Germany.
E-mail: bovier@uni-bonn.de

[‡]Institut für Angewandte Mathematik, Rheinische Friedrich-Wilhelms-Universität Bonn, Germany.
E-mail: marello@iam.uni-bonn.de

[§]TU Delft, Netherlands. E-mail: e.pulvirenti@tudelft.nl

any pair of spins. In Bovier and Gayrard [6] it is proven that the RDCW free energy converges, in the thermodynamic limit, to that of the CW model, under some assumptions on the coupling distribution. Their result relies on the fact that the RDCW Hamiltonian can be approximated by that of the CW model up to a small perturbation which can be uniformly bounded in high probability. In the last decade the RDCW model have gained again some attention and various results at equilibrium have been proven, both in the annealed and quenched case. De Sanctis and Guerra [9] give an exact expression of the free energy first in the high temperature and low connectivity regime, and then at zero temperature. The control of the fluctuations of the magnetisation in the high temperature limit is addressed by De Sanctis [8], while recently Kabluchko, Löwe and Schubert [15] prove a quenched Central Limit Theorem for the magnetisation in the high temperature regime.

One of the features which make these random systems with “bond disorder” very appealing is their deep connection with the theory of *random graphs*, which attracted great interest in the last years due to their application to real-world networks. Indeed, if the random couplings are chosen as i.i.d. Bernoulli random variables with mean p , one can view the model as a spin system on an Erdős–Rényi random graph with *fixed* edge probability p , which makes it a dense graph. There has been an extensive study of the Ising model at equilibrium on different kinds of random graphs, e.g. in Dembo, Montanari [10] and Dommers, Giardinà, van der Hofstad [14], where several thermodynamic quantities were analysed when the graph size tends to infinity. These results were all obtained for sparse graphs which have a locally tree-like structure. We refer to van der Hofstad [17] for a general overview of these results.

In contrast to the substantial body of literature on the equilibrium properties of the RDCW model, much less is known about its dynamical properties. The present paper focuses on the phenomenon of *metastability* for the RDCW model where, for simplicity, the couplings are Bernoulli distributed with *fixed* parameter $p \in (0, 1)$, independent of the number of vertices N , and the system evolves according to a Glauber dynamics. In particular, we give a precise estimate of the mean transition time from a certain probability distribution on the *metastable state* (called the last-exit biased distribution) to the *stable state*, when the external magnetic field is small enough and positive and when N tends to infinity. We obtain asymptotic bounds on the probability of the event that the average time is close to the CW one times some constants of order 1 which depend on the parameters of the system.

In the context of metastability for interacting particle systems on random graphs, progress has been made for the case of the random regular graph, analysed by Dommers [13] and for the configuration model, studied by Dommers, den Hollander, Jovanovski, and Nardi [12], both subject to Glauber dynamics, in the limit as the temperature tends to zero and the number of vertices is fixed. Both are dealing with sparse random graphs. In [11] den Hollander and Jovanovski investigate the same model considered in the present paper and obtain estimates on the average crossover time for fixed temperature in the thermodynamic limit. They show that, with high probability, the exponential term is the same as in the CW model, while the multiplicative term is polynomial in N . Their analysis relies on coupling arguments and on the *pathwise approach* to metastability. This method uses large deviations techniques in path space and focuses on properties of typical paths in the spirit of Freidlin-Wentzell theory. We refer to the classical book by Olivieri and Vares [18] for an overview on this method.

In contrast, in the present paper, we use the *potential theoretic approach* initiated by Bovier, Eckhoff, Gayrard and Klein in a series of papers [3, 4, 5] (see the monograph of Bovier and den Hollander [2] for an in-depth review of this as well as other approaches). This method gives less information on the evolution of the system, but leads to more

precise estimates of the metastable transition time. It has been successfully applied to a large variety of systems such as the random field CW model, where the external magnetic field is given by i.i.d. random variables, first by Bovier, Eckhoff, Gayrard and Klein in [3] and later by Bianchi, Bovier and Ioffe in [1]. Furthermore, inspired by the results of Bovier and Gayrard [6], namely that the equilibrium properties of the RDCW model are very close to those of the CW model, we observe that, using Talagrand’s concentration inequality, the mesoscopic measure can be expressed in terms of that of CW.

Before stating our results we give a precise definition of the model.

1.1 Glauber dynamics for the RDCW model

Let $[N] = \{1, \dots, N\}$, $N \in \mathbb{N}$, be a set of vertices. To each vertex $i \in [N]$ an Ising spin σ_i with values in $\{-1, +1\}$ is associated. We denote by $\sigma = \{\sigma_i : i \in [N], \sigma_i \in \{-1, +1\}\}$ a spin configuration and we define the state space $\mathcal{S}_N = \{-1, +1\}^N$ to be the set of all such configurations σ . We fix a probability $p \in (0, 1)$. Then the *randomly dilute Curie–Weiss* model (RDCW) has the following *random Hamiltonian* $H_N : \mathcal{S}_N \rightarrow \mathbb{R}$

$$H_N(\sigma) = -\frac{1}{Np} \sum_{1 \leq i < j \leq N} J_{ij} \sigma_i \sigma_j - h \sum_{i \in [N]} \sigma_i, \tag{1.1}$$

where $h \in \mathbb{R}$ represents an external constant magnetic field, while J_{ij}/Np is a ferromagnetic random coupling. In particular, $\{J_{ij}\}_{i,j \in [N]}$ is a sequence of i.i.d. random variables with $J_{ij} \sim \text{Ber}(p)$ and $J_{ij} = J_{ji}$.

Let us denote by \mathbb{P}_J the joint probability distribution of the the random couplings J_{ij} with $i, j \in [N]$ and by \mathbb{E} the corresponding mean value.

The RDCW model can be seen as the Ising model on the Erdős–Rényi random graph with vertex set $[N]$, edge set E and edge probability $p \in (0, 1)$ (see van der Hofstad [16] for a general overview on random graphs). In this picture the Hamiltonian can also be written as

$$H_N(\sigma) = -\frac{1}{Np} \sum_{\{i,j\} \in E} \sigma_i \sigma_j - h \sum_{i \in [N]} \sigma_i. \tag{1.2}$$

The Gibbs measure associated to the random Hamiltonian H_N is

$$\mu_{\beta,N}(\sigma) = \frac{e^{-\beta H_N(\sigma)}}{Z_{\beta,N}}, \quad \sigma \in \mathcal{S}_N, \tag{1.3}$$

where $\beta \in (0, \infty)$ is the inverse temperature and the partition function is defined as

$$Z_{\beta,N} = \sum_{\sigma \in \mathcal{S}_N} e^{-\beta H_N(\sigma)}. \tag{1.4}$$

The Gibbs measure $\mu_{\beta,N}$ is the unique invariant (and reversible) measure for the (discrete time) Glauber dynamics on \mathcal{S}_N with Metropolis transition probabilities

$$p_N(\sigma, \sigma') = \begin{cases} \frac{1}{N} \exp(-\beta[H_N(\sigma') - H_N(\sigma)]_+), & \text{if } \sigma \sim \sigma', \\ 1 - \sum_{\eta \neq \sigma} p_N(\sigma, \eta), & \text{if } \sigma = \sigma', \\ 0, & \text{else,} \end{cases} \tag{1.5}$$

where $\sigma \sim \sigma'$ means $\|\sigma - \sigma'\| = 2$ with $\|\cdot\|$ the ℓ_1 -norm on \mathcal{S}_N , i.e. $\sigma \sim \sigma'$ if and only if σ' is obtained from σ by a single spin flip. We denote this Markov chain by $\{\sigma(t)\}_{t \geq 0}$ and write \mathbb{P}_ν for the law of the process $\sigma(t)$ with initial distribution ν conditioned on the realisation of the random couplings. Analogously, \mathbb{E}_ν is the quenched expectation with

respect to the Markov chain with initial distribution ν . Moreover, we set $\mathbb{P}_\sigma = \mathbb{P}_{\delta_\sigma}$. For any subset $A \subset \mathcal{S}_N$ we define the hitting time of A as

$$\tau_A = \inf\{t > 0 : \sigma_t \in A\}. \tag{1.6}$$

Notice that H_N , $\mu_{\beta,N}$ and p_N are random variables, with respect to the random realisation of the random variables $\{J_{ij}\}_{i,j \in [N]}$. In this paper the results involving these random variables hold pointwise, namely for every realisation of $\{J_{ij}\}_{i,j \in [N]}$, unless we specify it differently, as in our main theorems.

1.2 The Curie–Weiss model

Before stating the main results, we recall some results for the mean-field Curie–Weiss (CW) model (see e.g. Bovier and den Hollander [2, Section 13] and Bovier, Eckhoff, Gaynard and Klein [3]). The CW Hamiltonian \tilde{H}_N can be obtained taking the mean value of (1.1) (namely, the first equality in (1.8) below). A simplifying feature of the CW model is that its Hamiltonian depends on the configuration $\sigma \in \mathcal{S}_N$ only through the empirical magnetisation $m_N : \mathcal{S}_N \rightarrow \Gamma_N$ defined as

$$m_N(\sigma) = \frac{1}{N} \sum_{i=1}^N \sigma_i \in \Gamma_N = \left\{ -1, -1 + \frac{2}{N}, \dots, 1 - \frac{2}{N}, 1 \right\}. \tag{1.7}$$

From now on we will drop the dependency on N from the magnetisation. Then we can write

$$\tilde{H}_N(\sigma) = -\frac{1}{N} \sum_{1 \leq i < j \leq N} \sigma_i \sigma_j - h \sum_{i \in [N]} \sigma_i = -N \left(\frac{1}{2} m(\sigma)^2 + hm(\sigma) \right) \tag{1.8}$$

and we can define, for any $m \in \Gamma_N$,

$$E(m) = -\frac{1}{2} m^2 - hm, \tag{1.9}$$

obtaining

$$\tilde{H}_N(\sigma) = NE(m(\sigma)). \tag{1.10}$$

The associated Gibbs measure is

$$\tilde{\mu}_{\beta,N}(\sigma) = \frac{e^{-\beta NE(m(\sigma))}}{\tilde{Z}_{\beta,N}}, \quad \sigma \in \mathcal{S}_N, \tag{1.11}$$

where $\tilde{Z}_{\beta,N} = \sum_{\sigma \in \mathcal{S}_N} e^{-\beta \tilde{H}_N(\sigma)}$ is the normalising partition function.

We denote the law of $m(\sigma)$ under the Gibbs measure by

$$\tilde{Q}_{\beta,N} = \tilde{\mu}_{\beta,N} \circ m^{-1}. \tag{1.12}$$

Then

$$\tilde{Q}_{\beta,N}(m) = \frac{e^{-\beta NE(m)}}{\tilde{Z}_{\beta,N}} \sum_{\sigma \in \mathcal{S}_N} \mathbb{1}_{m(\sigma)=m} = \frac{e^{-\beta NE(m)}}{\tilde{Z}_{\beta,N}} \binom{N}{\frac{1+m}{2}N} = \frac{e^{-\beta N f_{\beta,N}(m)}}{\tilde{Z}_{\beta,N}}, \tag{1.13}$$

where

$$f_{\beta,N}(m) = E(m) + \beta^{-1} I_N(m) = -\frac{m^2}{2} - hm + \beta^{-1} I_N(m) \tag{1.14}$$

is the finite volume *free energy*, while the *entropy* of the system is given by the following combinatorial coefficient

$$I_N(m) = -\frac{1}{N} \log \left(\binom{N}{\frac{1+m}{2}N} \right) \tag{1.15}$$

and it has the following properties: as $N \rightarrow \infty$,

$$I_N(m) \rightarrow I(m) \equiv \frac{1-m}{2} \log \frac{1-m}{2} + \frac{1+m}{2} \log \frac{1+m}{2}, \tag{1.16}$$

more precisely,

$$I_N(m) - I(m) = \frac{1}{2N} \ln \frac{1-m^2}{4} + \frac{\ln N + \ln(2\pi)}{2N} + O\left(\frac{1}{N^2}\right). \tag{1.17}$$

As reference see for example Bovier, Eckhoff, Gayrard and Klein [3, (7.18)].

Notice that the previous definitions imply

$$\tilde{\mu}_{\beta,N}(\sigma) = \tilde{Q}_{\beta,N}(m(\sigma)) e^{NI_N(m(\sigma))}. \tag{1.18}$$

We use the notation $f_\beta(m) = \lim_{N \rightarrow \infty} f_{\beta,N}(m)$. We refer to Bovier and den Hollander [2, (13.2.6)] for more details on the following result.

Lemma 1.1. For $m \in (-1, 1)$,

$$e^{-\beta N f_{\beta,N}(m)} = e^{-\beta N f_\beta(m)} (1 + o(1)) \sqrt{\frac{2}{\pi N (1-m^2)}} \tag{1.19}$$

and for $m \in \{1, -1\}$, $f_{\beta,N}(m) = f_\beta(m)$.

Remark 1.2. Comparing our definitions and the literature (e.g. Bovier and den Hollander [2, Section 13.1]), one notices that the Gibbs measure is often defined with an additional factor 2^{-N} , corresponding to the reference measure. More precisely, the Gibbs measure would be $\tilde{\mu}_{\beta,N}(\sigma) = \frac{1}{Z_{\beta,N}} e^{-\beta N E(m(\sigma))} 2^{-N}$, where the partition function would be defined by $\sum_{\sigma \in \mathcal{S}_N} e^{-\beta \tilde{H}_N(\sigma)} 2^{-N}$. We preferred to discard the 2^{-N} from our definitions. Therefore, for consistency, our definition of I_N differs from the classical one by a factor 2^{-N} inside the logarithm, yielding a difference of $\log(2)$ in the limit in (1.16) with respect to Bovier and den Hollander [2, (13.1.14)] or Bovier, Eckhoff, Gayrard and Klein [3, (7.17)].

We consider the Glauber dynamics associated to the CW Hamiltonian in analogy with (1.5) and with transition probabilities $\tilde{p}_N(\sigma, \sigma')$. A particular feature of this model is that the image process $m(t) \equiv m(\sigma(t))$ of the Markov process $\sigma(t)$ under the map m is again a Markov process on Γ_N , with transition probabilities

$$\tilde{r}_N(m, m') = \begin{cases} \exp(-\beta N [E(m') - E(m)]_+) \frac{(1-m)}{2} & \text{if } m' = m + \frac{2}{N}, \\ \exp(-\beta N [E(m') - E(m)]_+) \frac{(1+m)}{2} & \text{if } m' = m - \frac{2}{N}, \\ 0 & \text{else.} \end{cases} \tag{1.20}$$

The equilibrium CW model displays a phase transition. Namely, there is a critical value of the inverse temperature $\beta_c = 1$ such that, in the regime $\beta > \beta_c$, $h > 0$ and small, the free energy $f_\beta(m)$ is a double-well function with local minimisers m_-, m_+ and saddle point m^* . They are the solutions of equation $m = \tanh(\beta(m+h))$. Since $f_\beta(m_-) > f_\beta(m_+)$, the phase with m_- represents the metastable state, while m_+ represents the stable state for the system. Define $m_-(N), m^*(N), m_+(N)$ as the closest points in Γ_N to m_-, m^*, m_+ respectively, with respect to the Euclidean distance on \mathbb{R} . $\{m_-(N), m_+(N)\}$ form a metastable set in the sense of Definition 8.2 of Bovier and den Hollander [2]. Let $\mathbb{E}_{m_-(N)}^{\text{CW}}$ be the expectation with respect to the Markov process $m(t)$ with transition probabilities \tilde{r}_N and starting at $m_-(N)$. Then the following theorem holds.

Theorem 1.3. For $\beta > 1$ and $h > 0$ small enough, as $N \rightarrow \infty$,

$$\begin{aligned} \mathbb{E}_{m_-(N)}^{\text{CW}}[\tau_{m_+(N)}] &= \exp\left(\beta N [f_\beta(m^*) - f_\beta(m_-)]\right) \\ &\times \frac{\pi}{1+m^*} \sqrt{\frac{1-m^{*2}}{1-m_-^2}} \frac{N(1+o(1))}{\beta \sqrt{f''_\beta(m_-) (-f''_\beta(m^*))}}. \end{aligned} \quad (1.21)$$

As a reference see Bovier and den Hollander [2, Theorem 13.1]. The difference of sign in the denominator with respect to our statement is due to the fact that their result holds for $h < 0$, while ours for $h > 0$.

We conclude this section by giving the explicit formula of the capacity for the CW model. The definition of *capacity* is given in (1.31), while its relation with the mean hitting time is given by the key relation (1.30). Let us denote, for any subset U of Γ_N , the set of configurations with magnetisation in U by

$$\mathcal{S}_N[U] = \{\sigma \in \mathcal{S}_N : m(\sigma) \in U\} \quad (1.22)$$

and for simplicity, for any $m \in \Gamma_N$, the set of configurations with given magnetisation m by $\mathcal{S}_N[m]$. Notice that $\mathcal{S}_N[m]$ has cardinality $e^{-NI_N(m)}$, where $I_N(m)$ is defined in (1.15).

Then, the following formula,

$$\text{cap}^{\text{CW}}(\mathcal{S}_N[m_-(N)], \mathcal{S}_N[m_+(N)]) = \frac{1}{\tilde{Z}_{\beta,N}} e^{-\beta N f_\beta(m^*)} \frac{\sqrt{\beta (-f''_\beta(m^*))}}{\pi N} \sqrt{\frac{1+m^*}{1-m^*}} (1+o(1)), \quad (1.23)$$

follows from standard arguments (see e.g. techniques used in the proof of Bovier and den Hollander [2, Theorem 13.1]).

1.3 Main results

For any $A, B \subset \mathcal{S}_N$ disjoint, we define the so-called *last-exit biased distribution* on A for the transition from A to B as

$$\nu_{A,B}(\sigma) = \frac{\mu_{\beta,N}(\sigma) \mathbb{P}_\sigma(\tau_B < \tau_A)}{\sum_{\sigma \in A} \mu_{\beta,N}(\sigma) \mathbb{P}_\sigma(\tau_B < \tau_A)}, \quad \sigma \in A. \quad (1.24)$$

Since we are going to use $\nu_{A,B}$ on the sets $\mathcal{S}_N[m_-(N)], \mathcal{S}_N[m_+(N)]$ defined above, we introduce the following simplified notation

$$\nu_{m_-,m_+}^N = \nu_{\mathcal{S}_N[m_-(N)], \mathcal{S}_N[m_+(N)]}. \quad (1.25)$$

The following theorem gives a description of the dynamical properties of the RDCW model in the *metastable regime* where h is positive and small enough, $\beta > \beta_c = 1$ (β_c is the critical inverse temperature for the RDCW model) and N is going to infinity. We provide an estimate on the mean time it takes to the system, starting with initial distribution ν_{m_-,m_+}^N , to reach $\mathcal{S}_N[m_+(N)]$. More precisely, we estimate, in the limit as $N \rightarrow \infty$, its ratio with the mean metastable exit time for the CW model to go from $m_-(N)$ to $m_+(N)$, providing constant upper and lower bounds independent of N . Because of the random interaction, the result is given in the form of tail bounds.

After recalling that notation \mathbb{P}_J and \mathbb{E}_ν was introduced in Section 1.1, while $\mathbb{E}_{m_-(N)}^{\text{CW}}$ was introduced in Section 1.2, we are ready to formulate our main theorem.

Theorem 1.4 (Mean metastable exit time). For $\beta > 1$, $h > 0$ small enough and for $s > 0$, there exist absolute constants $k_1, k_2 > 0$ and $C_1(p, \beta) < C_2(p, \beta, h)$ independent of N ,

such that

$$\lim_{N \uparrow \infty} \mathbb{P}_J \left(C_1 e^{-s} (1 + o(1)) \leq \frac{\mathbb{E}_{\nu_{m_-, m_+}^N} [\tau_{\mathcal{S}_N[m_+(N)]}]}{\mathbb{E}_{m_-(N)}^{CW} [\tau_{m_+(N)}]} \leq C_2 e^s (1 + o(1)) \right) \geq 1 - k_1 e^{-k_2 s^2}. \tag{1.26}$$

The quantities C_1 and C_2 in the previous theorem can be explicitly written. Set

$$\alpha = \frac{\beta^2(1-p)}{4p}, \quad \kappa = \alpha + \max_{\eta \in (0,1)} \left\{ \log \eta - \frac{\beta \sqrt{2\alpha + \log \left(\frac{c_1}{(1-\eta)^2} \right)}}{p\sqrt{2c_2}} \right\}, \tag{1.27}$$

where $c_1, c_2 > 0$ are absolute constants coming from Theorem 2.8. It is easy to see that $\kappa < \alpha$. With this notation

$$C_1 = C_1(\beta, h, p) = e^{-2\beta(1+h)-\alpha+\kappa}, \tag{1.28}$$

$$C_2 = C_2(\beta, h, p) = e^{2\beta(1+h)+2\alpha}. \tag{1.29}$$

1.4 Proof of the main theorem

The proof of Theorem 1.4 is based on the *potential theoretic approach* to metastability, which turns out to be a rather powerful tool to analyse the main object we are interested in, i.e. the mean hitting time of $\mathcal{S}_N[m_+(N)]$ for the system with initial distribution ν_{m_-, m_+}^N . The general ideas of this approach were first introduced in a series of papers by Bovier, Eckhoff, Gayraud and Klein [3, 4, 5]. We refer to Bovier and den Hollander [2] for an overview on this method.

The crucial formula in the study of metastability is given by the following relation linking mean hitting time and *capacity* of two sets $A, B \in \mathcal{S}_N$, which can be found in Bovier and den Hollander [2, Eq. (7.1.41)]

$$\mathbb{E}_{\nu_{A,B}}[\tau_B] = \sum_{\sigma \in A} \nu_{A,B}(\sigma) \mathbb{E}_\sigma[\tau_B] = \frac{1}{\text{cap}(A, B)} \sum_{\sigma' \in \mathcal{S}_N} \mu_{\beta, N}(\sigma') h_{AB}(\sigma'), \tag{1.30}$$

where the capacity, as in Bovier and den Hollander [2, (7.1.39)], is defined by

$$\text{cap}(A, B) = \sum_{\sigma \in A} \mu_{\beta, N}(\sigma) \mathbb{P}_\sigma(\tau_B < \tau_A). \tag{1.31}$$

The function h_{AB} is called *harmonic function* and has the following probabilistic interpretation

$$h_{AB}(\sigma) = \begin{cases} \mathbb{P}_\sigma(\tau_A < \tau_B) & \sigma \in \mathcal{S}_N \setminus (A \cup B), \\ \mathbb{1}_A(\sigma) & \sigma \in A \cup B. \end{cases} \tag{1.32}$$

We refer to Bovier and den Hollander [2, Section 7.1.2] for further details on the latter quantities.

By (1.30), in order to estimate mean hitting times one needs estimates both on the capacity and on the harmonic function.

We prove bounds on the capacity of two sets $\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]$, stated in the two following theorems.

Theorem 1.5. *For any $m_1 \neq m_2 \in \Gamma_N$ and any $s > 0$, there exist absolute constants $k_1, k_2 > 0$ such that*

$$\mathbb{P}_J \left(\frac{Z_{\beta, N} \text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2])}{\tilde{Z}_{\beta, N} \text{cap}^{CW}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2])} \leq e^{s+2\beta(1+h)+\alpha} (1 + o(1)) \right) \geq 1 - k_1 e^{-k_2 s^2}, \tag{1.33}$$

asymptotically as $N \rightarrow \infty$, where α is defined in (1.27).

Theorem 1.6. For any $m_1 \neq m_2 \in \Gamma_N$ and any $s > 0$, there exist absolute constants $k_1, k_2 > 0$ such that

$$\mathbb{P}_J \left(\frac{Z_{\beta,N} \text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2])}{\tilde{Z}_{\beta,N} \text{cap}^{\text{CW}}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2])} \geq e^{-(s+2\beta(1+h)+\alpha)}(1+o(1)) \right) \geq 1 - k_1 e^{-k_2 s^2}, \quad (1.34)$$

asymptotically as $N \rightarrow \infty$, where α is defined in (1.27).

We state asymptotic upper and lower bounds on the sum over the harmonic function in the numerator of (1.30) in the following proposition. We used the simplified notation

$$h_{m_-,m_+}^N = h_{\mathcal{S}_N[m_-(N)], \mathcal{S}_N[m_+(N)]}. \quad (1.35)$$

Theorem 1.7. For any $s > 0$, there exist absolute constants $k_1, k_2 > 0$ such that

$$\mathbb{P}_J \left(\sum_{\sigma \in \mathcal{S}_N} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \leq e^{\alpha+s} \frac{\exp(-\beta N f_\beta(m_-)) (1+o(1))}{Z_{\beta,N} \sqrt{(1-m_-^2) \beta f_\beta''(m_-)}} \right) \geq 1 - k_1 e^{-k_2 s^2}, \quad (1.36)$$

$$\mathbb{P}_J \left(\sum_{\sigma \in \mathcal{S}_N} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \geq e^{\kappa-s} \frac{\exp(-\beta N f_\beta(m_-)) (1+o(1))}{Z_{\beta,N} \sqrt{(1-m_-^2) \beta f_\beta''(m_-)}} \right) \geq 1 - k_1 e^{-k_2 s^2}, \quad (1.37)$$

asymptotically as $N \rightarrow \infty$, and where α and κ are defined in (1.27).

We conclude this section using Theorems 1.5-1.7, to prove the main theorem. First, we introduce the following notation which will be extensively used:

$$A \stackrel{P(s)}{\geq} B \quad \text{is equivalent to} \quad \mathbb{P}_J(A \geq B) \geq 1 - k_1 e^{-k_2 s^2}, \quad (1.38)$$

for all $s > 0$ and for some absolute constants $k_1, k_2 > 0$, whose values might change along the paper.

Proof of Theorem 1.4 . We prove here only the upper bound, as the lower bound follows similarly. More precisely, we prove

$$\frac{\mathbb{E}_{\nu_{m_-,m_+}^N} [\tau_{\mathcal{S}_N[m_+(N)]}]}{\mathbb{E}_{m_-(N)}^{\text{CW}} [\tau_{m_+(N)}]} \stackrel{P(s)}{\leq} C_2 e^s. \quad (1.39)$$

We start from (1.30), which in our case reads

$$\mathbb{E}_{\nu_{m_-,m_+}^N} [\tau_{\mathcal{S}_N[m_+(N)]}] = \frac{\sum_{\sigma \in \mathcal{S}_N} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma)}{\text{cap}(\mathcal{S}_N[m_-(N)], \mathcal{S}_N[m_+(N)])}. \quad (1.40)$$

From (1.36) we obtain

$$\mathbb{E}_{\nu_{m_-,m_+}^N} [\tau_{\mathcal{S}_N[m_+(N)]}] \stackrel{P(s)}{\leq} \frac{e^{\alpha+s} \exp(-\beta N f_\beta(m_-)) (1+o(1))}{Z_{\beta,N} \text{cap}(\mathcal{S}_N[m_-(N)], \mathcal{S}_N[m_+(N)]) \sqrt{(1-m_-^2) \beta f_\beta''(m_-)}}. \quad (1.41)$$

Via the lower bound on the capacity from Theorem 1.6, we obtain

$$\begin{aligned} & \mathbb{E}_{\nu_{m_-,m_+}^N} [\tau_{\mathcal{S}_N[m_+(N)]}] \\ & \leq e^{2s+2\beta(1+h)+2\alpha} \sqrt{\frac{1-m^*}{1+m^*}} \frac{\pi N \exp(\beta N [f_\beta(m^*) - f_\beta(m_-)])}{\beta \sqrt{(1-m_-^2) f_\beta''(m_-)} (-f_\beta''(m^*))} (1+o(1)) \\ & = e^{2s+2\beta(1+h)+2\alpha} \mathbb{E}_{m_-(N)}^{\text{CW}} [\tau_{m_+(N)}], \end{aligned} \quad (1.42)$$

where we used (1.23) and Theorem 1.3. □

1.5 Outline

The remainder of this paper is organised as follows. In Section 2 we use the powerful *Talagrand's concentration inequality* to obtain bounds on the equilibrium measure of the RDCW model. These bounds allow us to write the RDCW *mesoscopic* measure in terms of the deterministic CW one, times a random factor which is the exponential of a sub-Gaussian random variable. In Section 3 we give the proof of Theorems 1.5 and 1.6 via two dual variational principles, the Dirichlet and the Thomson principles, which are the building blocks of the potential theoretic approach to metastability. In obtaining upper and lower bounds on the capacity, the main strategy is to use the results of Section 2 in order to recover the capacity of the CW model. In Section 4 we prove Theorem 1.7, i.e. we compute the asymptotics of the numerator in the formula for the mean hitting time using estimates on the harmonic function.

2 Equilibrium analysis via Talagrand's concentration inequality

In this section we prove that the equilibrium mesoscopic measure of the RDCW model is in fact very close to that of the CW model. This is done in two steps. First, we prove that the difference between the *random free energy* at fixed magnetisation and its average can be controlled via *Talagrand's concentration inequality*. Second, we find upper and lower bounds on the aforementioned average by estimating first and second moments of the partition function of the RDCW model at fixed magnetisation.

2.1 Mesoscopic measure and closeness to the CW model

We start by analysing the equilibrium measure of the RDCW model. The aim is to express the equilibrium measure $\mu_{\beta,N}$, defined in (1.3), in terms of the empirical magnetisation in order to obtain a *mesoscopic* description, as we did for the CW model in Section 1.2. Let us define the measure $\mathcal{Q}_{\beta,N}$ on Γ_N , and let the partition function be its normalisation

$$\mathcal{Q}_{\beta,N}(\cdot) = \mu_{\beta,N} \circ m^{-1}(\cdot) = \sum_{\sigma \in \mathcal{S}_N[\cdot]} \mu_{\beta,N}(\sigma), \quad Z_{\beta,N} = \sum_{m \in \Gamma_N} \mathcal{Q}_{\beta,N}(m). \quad (2.1)$$

A priori the Hamiltonian of the RDCW model is not only depending on m , but it depends of course on the whole spin configuration. Nonetheless, we will see later in this section that the mesoscopic measure $\mathcal{Q}_{\beta,N}$ can be written in terms of the mesoscopic measure $\tilde{\mathcal{Q}}_{\beta,N}$ of the standard CW model.

$$\mathbb{E}[H_N(\sigma)] = -\frac{1}{Np} \sum_{i < j} \mathbb{E}[J_{ij}] \sigma_i \sigma_j - h \sum_i \sigma_i = -\frac{p}{Np} \sum_{i < j} \sigma_i \sigma_j - h \sum_i \sigma_i = \tilde{H}_N(\sigma). \quad (2.2)$$

Therefore, we can split the Hamiltonian into the mean-field part and the remaining random part obtaining

$$H_N(\sigma) = \mathbb{E}[H_N(\sigma)] + \Delta_{N,p}(\sigma), \quad (2.3)$$

where, introducing the notation $\hat{J}_{ij} = J_{ij} - p$,

$$\Delta_{N,p}(\sigma) = H_N(\sigma) - \tilde{H}_N(\sigma) = -\frac{1}{Np} \sum_{i < j} \hat{J}_{ij} \sigma_i \sigma_j. \quad (2.4)$$

Note that $\Delta_{N,p}$ is a random variable with zero mean. In order to simplify the notation, we drop from now on the dependence on N and p , from $\Delta_{N,p}$. Next, we write the mesoscopic measure as

$$\mathcal{Q}_{\beta,N}(m) = \frac{1}{Z_{\beta,N}} e^{-\beta N E(m)} \cdot \sum_{\sigma \in \mathcal{S}_N[m]} e^{-\beta \Delta(\sigma)}, \quad (2.5)$$

where $E(m)$ is defined in (1.8).

We will now focus on proving bounds for functions of $\sum_{\sigma \in \mathcal{S}_N[m]} e^{-\beta \Delta(\sigma)}$ more general than $\mathcal{Q}_{\beta,N}(m)$. These results will be fundamental to prove our main theorem in the following sections. We will come back to $\mathcal{Q}_{\beta,N}$ at the end of this section, proving its closeness to the CW correspondent $\tilde{\mathcal{Q}}_{\beta,N}$ as a consequence of those general results.

Let us introduce the following notation, where we drop the dependence on β for simplicity

$$\mathcal{Z}_{N,g} = \sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} e^{-\beta \Delta(\sigma)} = \exp(N p_{N,g}) \exp(N [F_{N,g} - p_{N,g}]), \quad (2.6)$$

$$F_{N,g} = \frac{1}{N} \log \mathcal{Z}_{N,g}, \quad (2.7)$$

$$p_{N,g} = \mathbb{E}(F_{N,g}), \quad (2.8)$$

where $g : \Gamma_N \rightarrow [0, \infty)$ is a function which may depend on N .

We are interested in finding precise estimates on $\mathcal{Z}_{N,g}$ by writing it in terms of the entropic exponential term $e^{-NI_N(m)}$ times some random factor which takes into account the randomness of the couplings. We notice that $\mathcal{Z}_{N,g}$ is the product of a deterministic factor $e^{N p_{N,g}}$ and a random factor $e^{N(F_{N,g} - p_{N,g})}$.

We first characterise the random variable $N(F_{N,g} - p_{N,g})$ in the following Proposition.

Proposition 2.1. For any $\beta, t > 0$,

$$\mathbb{P}_J \left(|N(F_{N,g} - p_{N,g})| \geq t \right) \leq c_1 \exp \left(-\gamma t^2 \right), \quad (2.9)$$

where $\gamma \propto \frac{p^2}{\beta^2}$.

The previous result intuitively means that the random $F_{N,g}$ is in fact very well concentrated around its mean $p_{N,g}$.

As a second step we provide asymptotic bounds on the average of $F_{N,g}$, i.e. the deterministic term $p_{N,g}$.

Lemma 2.2. Asymptotically, as $N \rightarrow \infty$,

$$p_{N,g} \leq \frac{\alpha}{N} + \frac{1}{N} \log \left(\sum_{m \in \Gamma_N} g(m) \exp(-NI_N(m)) \right) + o\left(\frac{1}{N}\right), \quad (2.10)$$

where $I_N(m)$ is defined in (1.15) and α in (1.27).

Lemma 2.3. Asymptotically, as $N \rightarrow \infty$,

$$p_{N,g} \geq \frac{\kappa}{N} + \frac{1}{N} \log \left(\sum_{m \in \Gamma_N} g(m) \exp(-NI_N(m)) \right) + o\left(\frac{1}{N}\right), \quad (2.11)$$

where $I_N(m)$ is defined in (1.15) and κ in (1.27).

Proposition 2.1 together with Lemmas 2.2 and 2.3 imply the following result.

Proposition 2.4. Asymptotically, as $N \rightarrow \infty$, we have

$$\mathcal{Z}_{N,g} \leq e^\alpha \left(\sum_{m \in \Gamma_N} g(m) \exp(-NI_N(m)) \right) \exp[N(F_{N,g} - p_{N,g})] (1 + o(1)), \quad (2.12)$$

and

$$\mathcal{Z}_{N,g} \geq e^\kappa \left(\sum_{m \in \Gamma_N} g(m) \exp(-NI_N(m)) \right) \exp[N(F_{N,g} - p_{N,g})] (1 + o(1)), \quad (2.13)$$

where $Z_{N,g}$ is defined in (2.6), α and κ in (1.27), and $I_N(m)$ in (1.15). Moreover, $N(F_{N,g} - p_{N,g})$ is a sub-Gaussian random variable with variance

$$\text{Var}[N(F_{N,g} - p_{N,g})] \leq \frac{c\beta^2}{p^2}, \tag{2.14}$$

where c is a positive constant.

We prove Proposition 2.1 in Section 2.2, and Lemmas 2.2 and 2.3 in Section 2.3.

We are ready to state the main result of this section, as a corollary of Proposition 2.1 and Proposition 2.4.

Corollary 2.5. *Asymptotically, as $N \rightarrow \infty$, using notation (1.38), the following bounds hold for any $\beta > 0$ and any function $g : \Gamma_N \rightarrow [0, \infty)$*

$$\sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} e^{-\beta\Delta(\sigma)} \stackrel{P(s)}{\leq} e^{s+\alpha} \left(\sum_{m \in \Gamma_N} g(m) \exp(-NI_N(m)) \right) (1 + o(1)), \tag{2.15}$$

$$\sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} e^{-\beta\Delta(\sigma)} \stackrel{P(s)}{\geq} e^{-s+\kappa} \left(\sum_{m \in \Gamma_N} g(m) \exp(-NI_N(m)) \right) (1 + o(1)), \tag{2.16}$$

where α and κ are defined in (1.27), $I_N(m)$ in (1.15) and $\Delta(\sigma)$ in (2.4).

Proof. By Proposition 2.1 we obtain, for any fixed $s > 0$,

$$\exp[N(F_{N,g} - p_{N,g})] \stackrel{P(s)}{\leq} e^s \quad \text{and} \quad \exp[N(F_{N,g} - p_{N,g})] \stackrel{P(s)}{\geq} e^{-s}, \tag{2.17}$$

where $k_1, k_2 > 0$ are absolute constants.

To conclude the proof it is sufficient to use the definition of $Z_{N,g}$ (2.6) and Proposition 2.4. □

Remark 2.6. The exact same statement of Corollary 2.5 holds replacing $e^{-\beta\Delta(\sigma)}$ with $e^{\beta\Delta(\sigma)}$. The proof remains the same: the Lipschitz constant for the Talagrand concentration inequality (in Section 2.2) is the same and the change of sign, being squared, disappears from (2.30) onwards.

We conclude this section with an immediate application of Corollary 2.5 which states the closeness of the random mesoscopic measure $\mathcal{Q}_{\beta,N}$ to the correspondent deterministic CW quantity $\tilde{\mathcal{Q}}_{\beta,N}$. This result will be widely used in Section 4.

Corollary 2.7. *Asymptotically, as $N \rightarrow \infty$, using notation (1.38), the following bounds hold for any fixed $s > 0$ and any function $\bar{g} : \Gamma_N \rightarrow [0, \infty)$*

$$\sum_{m \in \Gamma_N} \bar{g}(m) \mathcal{Q}_{\beta,N}(m) \stackrel{P(s)}{\leq} e^{s+\alpha} \frac{\tilde{Z}_{\beta,N}}{Z_{\beta,N}} \left(\sum_{m \in \Gamma_N} \bar{g}(m) \tilde{\mathcal{Q}}_{\beta,N}(m) \right) (1 + o(1)), \tag{2.18}$$

$$\begin{aligned} & \sum_{m \in \Gamma_N} \bar{g}(m) \mathcal{Q}_{\beta,N}(m) \\ & \stackrel{P(s)}{\leq} e^{s+\alpha} \frac{1}{Z_{\beta,N}} \left(\sum_{m \in \Gamma_N \setminus \{1,-1\}} \bar{g}(m) \exp(-\beta N f_\beta(m)) \sqrt{\frac{2}{\pi N(1-m^2)}} \right) (1 + o(1)) \\ & + e^{s+\alpha} \frac{1}{Z_{\beta,N}} \left(\sum_{m \in \{1,-1\}} \bar{g}(m) \exp(-\beta N f_\beta(m)) \right) (1 + o(1)), \end{aligned} \tag{2.19}$$

where α and κ are defined in (1.27).

Proof. Using (2.5) we obtain

$$\sum_{m \in \Gamma_N} \bar{g}(m) \mathcal{Q}_{\beta,N}(m) = \frac{1}{Z_{\beta,N}} \sum_{m \in \Gamma_N} \bar{g}(m) e^{-\beta N E(m)} \sum_{\sigma \in \mathcal{S}_N[m]} e^{-\beta \Delta(\sigma)}. \quad (2.20)$$

Now we can apply the upper bound in Corollary 2.5, with $g(m) = \frac{1}{Z_{\beta,N}} \bar{g}(m) e^{-\beta N E(m)}$, to the right hand side of (2.20). We conclude the proof of (2.18) using the definition of $\tilde{\mathcal{Q}}_{\beta,N}$ (1.13) and (1.14).

(2.19) follows by (2.18) simply applying Lemma 1.1. □

2.2 Sub-Gaussian bounds on the random term

Proposition 2.1 follows from Talagrand’s concentration inequality, which we cite for completeness in the version of Tao [20, Theorem 2.1.13].

Theorem 2.8 (Talagrand concentration inequality). *Let $G : \mathbb{R}^M \rightarrow \mathbb{R}$ be a 1-Lipschitz and convex function. Let $M \in \mathbb{N}$, $X = (X_1, \dots, X_M)$, with X_i be independent r.v., uniformly bounded by $K > 0$, i.e. $|X_i| \leq K$, for every $1 \leq i \leq M$. Then, for any $t \geq 0$,*

$$\mathbb{P}\left(|G(X) - \mathbb{E} G(X)| \geq tK\right) \leq c_1 \exp(-c_2 t^2), \quad (2.21)$$

with positive absolute constants c_1, c_2 .

Proof of Proposition 2.1. We can apply Theorem 2.8 to the free energies $F_{N,g}$ as a function of the N^2 coupling constants \hat{J}_{ij} . Indeed it is standard to see that $F_{N,g}$ is convex and Lipschitz continuous with constant $\frac{\beta}{Np\sqrt{2}}$ (see e.g. Talagrand [19, Corollary 2.2.5]). Thus, applying Theorem 2.8 for $G = F_{N,g} \left(\frac{\beta}{Np\sqrt{2}}\right)^{-1}$ and $K = 1$, after defining $t' = t \frac{\beta}{Np\sqrt{2}}$ we obtain, for some positive constants c_1, c_2 and for any $t' \geq 0$,

$$\mathbb{P}_J\left(N|F_{N,g} - p_{N,g}| \geq t'\right) \leq c_1 \exp\left(-c_2 \frac{2p^2}{\beta^2} t'^2\right), \quad (2.22)$$

concluding the proof of (2.9) and hence Proposition 2.1. □

2.3 Asymptotic bounds on the deterministic term

In this section we prove first the upper bound on $p_{N,g}$ (Lemma 2.2) and then the lower bound (Lemma 2.3). The upper bound is obtained by estimates on the first moment of the random partition function $\mathcal{Z}_{N,g}$, while the lower bound is in the spirit of Talagrand [19, Theorem 2.2.1] and is more delicate. We will see that it involves also estimates on the second moment of the random partition function.

Proof of Lemma 2.2. Observing that $\{\hat{J}_{ij}\}_{i,j \in [N]}$ defined in (2.4) are i.i.d. random variables such that $\mathbb{E} \hat{J}_{ij} = 0$, we easily obtain

$$\begin{aligned} \mathbb{E}[\mathcal{Z}_{N,g}] &= \sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} \mathbb{E} \left(\exp \left[\frac{\beta}{Np} \sum_{i < j} \hat{J}_{ij} \sigma_i \sigma_j \right] \right) \\ &= \sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} \prod_{i < j} \mathbb{E} \left(\exp \left[\frac{\beta}{Np} \hat{J}_{ij} \sigma_i \sigma_j \right] \right). \end{aligned} \quad (2.23)$$

In order to find estimates for (2.23), we first define

$$\Phi(x) := \mathbb{E} [\exp(x \hat{J}_{ij})], \quad (2.24)$$

which is a function independent of i, j , being $\{\hat{J}_{ij}\}_{i,j}$ i.i.d., with first and second derivatives

$$\Phi'(0) = \mathbb{E} \hat{J}_{ij} = 0, \tag{2.25}$$

$$\Phi''(0) = \mathbb{E} \hat{J}_{ij}^2 = p(1 - p). \tag{2.26}$$

Performing a Taylor expansion of Φ we get

$$\Phi(x) = \Phi(0) + x \Phi'(0) + \frac{x^2}{2} \Phi''(0) + o(x^2) = 1 + \frac{x^2}{2} p(1 - p) + o(x^2). \tag{2.27}$$

Thus, we can exponentiate $\Phi(x)$ to obtain

$$\Phi(x) = \exp \left(\log (\Phi(x)) \right) = \exp \left(\frac{x^2}{2} p(1 - p) + o(x^2) \right), \tag{2.28}$$

where we used the expansion $\log(1 + x) = x + o(x)$. Therefore, for any sequence of coefficients x_{ij}^2 which are independent of i, j and σ , we have the following

$$\begin{aligned} & \sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} \prod_{i < j} \mathbb{E} \left[\exp(x_{ij} \hat{J}_{ij}) \right] = \sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} \prod_{i < j} \Phi(x_{ij}) \\ & = \sum_{m \in \Gamma_N} g(m) \sum_{\sigma \in \mathcal{S}_N[m]} \prod_{i < j} \exp \left(\frac{x_{ij}^2}{2} p(1 - p) + o(x_{ij}^2) \right) \\ & = \sum_{m \in \Gamma_N} g(m) e^{-NI_N(m)} \exp \left(\frac{x_{ij}^2}{2} p(1 - p) + o(x_{ij}^2) \right)^{N(N-1)/2} \\ & = \sum_{m \in \Gamma_N} g(m) e^{-NI_N(m)} \exp \left(x_{ij}^2 p(1 - p) \frac{N(N-1)}{4} + o(x_{ij}^2 N(N-1)) \right), \end{aligned} \tag{2.29}$$

asymptotically, for $x_{ij} \rightarrow 0$, where the third equality holds only if x_{ij}^2 is independent of i, j and σ . Moreover, we used that the cardinality of $\mathcal{S}_N[m]$ is $e^{-NI_N(m)}$, where $I_N(m)$ is defined in (1.15), and the cardinality of $\{(i, j) \in [N]^2 : i < j\}$ is $\frac{N(N-1)}{2}$.

We can apply (2.29) with $x_{ij} = \frac{\beta}{Np} \sigma_i \sigma_j$ because x_{ij}^2 is independent of i, j and σ . Indeed $x_{ij}^2 = \frac{\beta^2}{N^2 p^2}$, being $\sigma_i, \sigma_j \in \{-1, +1\}$ for any $i, j \in [N]$ and $\sigma \in \mathcal{S}_N$. Thus, we get, asymptotically as $N \rightarrow \infty$,

$$\begin{aligned} \mathbb{E} [\mathcal{Z}_{N,g}] & = \sum_{m \in \Gamma_N} g(m) e^{-NI_N(m)} \exp \left(\frac{\beta^2(1-p)}{4p} + o(1) \right) \\ & = \exp \left(\alpha + o(1) \right) \sum_{m \in \Gamma_N} g(m) \exp \left(-NI_N(m) \right), \end{aligned} \tag{2.30}$$

where α is defined in (1.27).

Therefore, by Jensen's inequality and (2.30), we have

$$\mathbb{E} \left[\log \mathcal{Z}_{N,g} \right] \leq \log \left(\mathbb{E} [\mathcal{Z}_{N,g}] \right) = \alpha + o(1) + \log \left(\sum_{m \in \Gamma_N} g(m) \exp \left(-NI_N(m) \right) \right) \tag{2.31}$$

which proves the upper bound. □

Proof of Lemma 2.3. A key ingredient in the proof is to control the upper bound on the second moment of $\mathcal{Z}_{N,g}$, i.e. prove that the following bound holds

$$\mathbb{E} [\mathcal{Z}_{N,g}^2] \leq e^{2\alpha} \mathbb{E} [\mathcal{Z}_{N,g}]^2 (1 + o(1)), \tag{2.32}$$

where α is defined in (1.27).

We estimate $\mathbb{E} [\mathcal{Z}_{N,g}^2]$ using the first two lines of (2.29) with

$$x_{ij} = \frac{\beta}{Np} \left(\sigma_i^{(1)} \sigma_j^{(1)} + \sigma_i^{(2)} \sigma_j^{(2)} \right), \tag{2.33}$$

which hold also when x_{ij}^2 is not independent on i, j and σ ,

$$\begin{aligned} \mathbb{E} [\mathcal{Z}_{N,g}^2] &= \mathbb{E} \left[\sum_{m,m' \in \Gamma_N} g(m) g(m') \sum_{\substack{\sigma^{(1)} \in \mathcal{S}_N[m], \\ \sigma^{(2)} \in \mathcal{S}_N[m']}} \exp \left(\sum_{i < j} \frac{\beta}{Np} \hat{J}_{ij} \left(\sigma_i^{(1)} \sigma_j^{(1)} + \sigma_i^{(2)} \sigma_j^{(2)} \right) \right) \right] \\ &= \sum_{m,m' \in \Gamma_N} g(m) g(m') \mathbb{E} \left[\sum_{\substack{\sigma^{(1)} \in \mathcal{S}_N[m], \\ \sigma^{(2)} \in \mathcal{S}_N[m']}} \exp \left(\sum_{i < j} \frac{\beta}{Np} \hat{J}_{ij} \left(\sigma_i^{(1)} \sigma_j^{(1)} + \sigma_i^{(2)} \sigma_j^{(2)} \right) \right) \right] \\ &= \sum_{m,m' \in \Gamma_N} g(m) g(m') \sum_{\substack{\sigma^{(1)} \in \mathcal{S}_N[m], \\ \sigma^{(2)} \in \mathcal{S}_N[m']}} \prod_{i < j} \exp \left(\frac{1}{2} \frac{\beta^2}{N^2 p^2} \left(\sigma_i^{(1)} \sigma_j^{(1)} + \sigma_i^{(2)} \sigma_j^{(2)} \right)^2 p(1-p) + o \left(\frac{\beta^2}{N^2} \right) \right) \\ &\leq \sum_{m,m' \in \Gamma_N} g(m) g(m') \sum_{\substack{\sigma^{(1)} \in \mathcal{S}_N[m], \\ \sigma^{(2)} \in \mathcal{S}_N[m']}} \prod_{i < j} \exp \left(\frac{\beta^2}{N^2 p} 2(1-p) + o \left(\frac{1}{N^2} \right) \right) \\ &= \sum_{m,m' \in \Gamma_N} g(m) g(m') e^{-NI_N(m)} e^{-NI_N(m')} \exp \left(\frac{N(N-1)}{2} \left[\frac{\beta^2}{N^2 p} 2(1-p) + o \left(\frac{1}{N^2} \right) \right] \right) \\ &= \sum_{m,m' \in \Gamma_N} g(m) g(m') e^{-NI_N(m)} e^{-NI_N(m')} \exp \left(\beta^2 \frac{(1-p)}{p} + o(1) \right) \\ &= \exp(4\alpha + o(1)) \sum_{m \in \Gamma_N} g(m) e^{-NI_N(m)} \sum_{m' \in \Gamma_N} g(m') e^{-NI_N(m')} \\ &= e^{2\alpha} \mathbb{E} [\mathcal{Z}_{N,g}]^2 (1 + o(1)), \end{aligned} \tag{2.34}$$

where, similarly to the last steps in (2.29), we used that the cardinality of $\mathcal{S}_N[m]$ is $e^{-NI_N(m)}$, the cardinality of $\{(i, j) \in [N]^2 : i < j\}$ is $\frac{N(N-1)}{2}$. Moreover, in the last line we used (2.30).

We recall the Paley-Zygmund inequality, which states that

$$\mathbb{P}_J(X \geq \eta \mathbb{E}X) \geq (1 - \eta)^2 \frac{(\mathbb{E}X)^2}{\mathbb{E}X^2}, \tag{2.35}$$

for any non negative random variable X and any $\eta \in (0, 1)$. Using (2.35) with $X = \mathcal{Z}_{N,g}$, (2.30) and (2.34) we get, asymptotically as $N \rightarrow \infty$,

$$\begin{aligned} &\mathbb{P}_J \left(\frac{1}{N} \log \mathcal{Z}_{N,g} \geq \frac{1}{N} \log (\eta \mathbb{E} \mathcal{Z}_{N,g}) \right) \\ &= \mathbb{P}_J \left(\frac{1}{N} \log \mathcal{Z}_{N,g} \geq \frac{1}{N} \log (\mathbb{E} \mathcal{Z}_{N,g}) + \frac{1}{N} \log \eta \right) \\ &\geq \frac{(1 - \eta)^2}{\exp(2\alpha + o(1))}. \end{aligned} \tag{2.36}$$

Moreover, using (2.22) together with (2.7) and the change of variables $t' = Nt''$, we obtain $\forall t'' > 0$,

$$\mathbb{P}_J \left(\left| \frac{1}{N} \log \mathcal{Z}_{N,g} - p_{N,g} \right| \geq t'' \right) \leq c_1 \exp \left(-\frac{2c_2 N^2 p^2 t''^2}{\beta^2} \right). \quad (2.37)$$

Thus, taking the complementary event, we get

$$\mathbb{P}_J \left(-t'' \leq \frac{1}{N} \log \mathcal{Z}_{N,g} - p_{N,g} \leq t'' \right) \geq 1 - c_1 \exp \left(-\frac{2c_2 N^2 p^2 t''^2}{\beta^2} \right). \quad (2.38)$$

Now, using

$$\mathbb{P}_J \left(\frac{1}{N} \log \mathcal{Z}_{N,g} - p_{N,g} \leq t'' \right) \geq \mathbb{P}_J \left(-t'' \leq \frac{1}{N} \log \mathcal{Z}_{N,g} - p_{N,g} \leq t'' \right) \quad (2.39)$$

and the change of variable $t = \frac{Np\sqrt{2c_2}}{\beta} t''$ we obtain

$$\mathbb{P}_J \left(\frac{1}{N} \log \mathcal{Z}_{N,g} \leq p_{N,g} + \frac{t\beta}{Np\sqrt{2c_2}} \right) \geq 1 - c_1 \exp(-t^2). \quad (2.40)$$

Next we prove that the intersection of the events in (2.36) and (2.40) is non empty. Assuming, for $\eta \in (0, 1)$, that

$$\mathbb{P}_J \left(\frac{1}{N} \log \mathcal{Z}_{N,g} \leq p_{N,g} + \frac{t\beta}{Np\sqrt{2c_2}} \right) > 1 - \frac{(1-\eta)^2}{\exp(2\alpha + o(1))} \quad (2.41)$$

and comparing (2.36) and (2.41), we notice that the sum of the probabilities of the two events

$$\left\{ \frac{1}{N} \log \mathcal{Z}_{N,g} \leq p_{N,g} + \frac{t\beta}{Np\sqrt{2c_2}} \right\}, \quad (2.42)$$

and

$$\left\{ \frac{1}{N} \log \mathcal{Z}_{N,g} \geq \frac{1}{N} \log (\mathbb{E} \mathcal{Z}_{N,g}) + \frac{1}{N} \log \eta \right\} \quad (2.43)$$

is strictly greater than 1. Therefore, they intersect in the not empty event

$$\left\{ \frac{1}{N} \log (\mathbb{E} \mathcal{Z}_{N,g}) + \frac{1}{N} \log \eta \leq \frac{1}{N} \log \mathcal{Z}_{N,g} \leq p_{N,g} + \frac{t\beta}{Np\sqrt{2c_2}} \right\} \quad (2.44)$$

which is contained in the deterministic set

$$\left\{ \frac{1}{N} \log (\mathbb{E} \mathcal{Z}_{N,g}) + \frac{1}{N} \log \eta \leq p_{N,g} + \frac{t\beta}{Np\sqrt{2c_2}} \right\}. \quad (2.45)$$

As a consequence, the latter set is non empty and, being deterministic,

$$p_{N,g} \geq \frac{1}{N} \log (\mathbb{E} \mathcal{Z}_{N,g}) + \frac{1}{N} \log \eta - \frac{t\beta}{Np\sqrt{2c_2}} \quad (2.46)$$

holds with probability 1.

It remains to choose a suitable $t > 0$ for assumption (2.41) to hold. A sufficient condition is, for every $\eta \in (0, 1)$,

$$c_1 \exp(-t^2) < \frac{(1-\eta)^2}{\exp(2\alpha + o(1))}, \quad (2.47)$$

namely

$$t^2 > 2\alpha + \log\left(\frac{c_1}{(1-\eta)^2}\right) + o(1). \tag{2.48}$$

Therefore, by (2.46) and (2.48), using (2.30) we obtain, for every $\eta \in (0, 1)$,

$$\begin{aligned} p_{N,g} &\geq \frac{1}{N} \log(\mathbb{E}\mathcal{Z}_{N,g}) + \frac{1}{N} \log \eta - \frac{\beta \sqrt{2\alpha + \log\left(\frac{c_1}{(1-\eta)^2}\right) + o(1)}}{Np\sqrt{2c_2}} \\ &= \frac{1}{N} \log\left(\sum_{m \in \Gamma_N} g(m) \exp(-NI_N(m))\right) + \frac{\kappa_\eta}{N} + o\left(\frac{1}{N}\right), \end{aligned} \tag{2.49}$$

where

$$\kappa_\eta = \alpha + \log \eta - \frac{\beta \sqrt{2\alpha + \log\left(\frac{c_1}{(1-\eta)^2}\right)}}{p\sqrt{2c_2}}. \tag{2.50}$$

Notice that $\kappa_\eta < \alpha$. In order to obtain the best lower bound, namely the closer to the upper bound proven in Lemma 2.2, we choose $\eta \in (0, 1)$ s.t. $\alpha - \kappa_\eta$ is minimised and we conclude the proof. This choice motivates the maximum in the definition of κ , in (1.27). \square

3 Capacity estimates

This section is entirely devoted to obtain upper and lower bounds on capacities between sets with a fixed magnetisation. These bounds are obtained via two dual variational principles, i.e. the *Dirichlet* and *Thomson principles* which are extensively discussed in Bovier and den Hollander [2, Sections 7.3.1, 7.3.2]. The result will be expressed in terms of the capacity for the Curie–Weiss model, see (1.23). In particular, we prove Theorem 1.5 in Section 3.1 and Theorem 1.6 in Section 3.2.

3.1 Asymptotics on capacity: upper bound

In this section we prove Theorem 1.5, obtaining the upper bound on the capacity of the RDCW model in terms of the capacity of the CW model.

Proof of Theorem 1.5. The main idea of the proof is to find an upper bound on the capacity via the following Dirichlet principle (see Bovier and den Hollander [2, Section 7.3.1 and (7.1.29)] for details)

$$\text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) = \min_{f \in \mathcal{H}} \sum_{\sigma, \sigma' \in \mathcal{S}_N} \mu_{\beta, N}(\sigma) p_N(\sigma, \sigma') [f(\sigma) - f(\sigma')]^2, \tag{3.1}$$

where

$$\mathcal{H} = \left\{ f : \mathcal{S}_N \rightarrow [0, 1] \text{ s.t. } f|_{\mathcal{S}_N[m_1]} = 1, f|_{\mathcal{S}_N[m_2]} = 0 \right\}. \tag{3.2}$$

Later it will be clear that we can restrict the previous variational principle over the functions on the space Γ_N , hence it is useful to define

$$\tilde{\mathcal{H}} = \left\{ v : \Gamma_N \rightarrow [0, 1] \text{ s.t. } v(m_1) = 1, v(m_2) = 0 \right\}. \tag{3.3}$$

In order to simplify the notation we will often neglect the dependency on m_1, m_2 when this will not generate confusion.

From (3.1), in view of (1.5) and since $[f(\sigma) - f(\sigma')]$ vanishes for $\sigma = \sigma'$, we are left only with the terms such that $\sigma \sim \sigma'$ and obtain the following first equality in (3.4).

The second equality in (3.4) follows by (1.10), (2.2), (2.3) and multiplying and dividing by $\exp(-\beta N [E(m(\sigma')) - E(m(\sigma))]_+)$. The inequality in (3.4) is obtained restricting the minimum on \mathcal{H} to the minimum on $\{f \in \mathcal{H} : f(\eta) = f(\eta') \forall \eta, \eta' \in \mathcal{S}_N \text{ s.t. } m(\eta) = m(\eta')\}$ and noticing that the latter is in bijection with \mathcal{H} .

$$\begin{aligned}
 & Z_{\beta, N} \text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) \\
 &= \min_{f \in \mathcal{H}} \frac{1}{N} \sum_{\sigma, \sigma' \in \mathcal{S}_N} \mathbb{1}_{\sigma \sim \sigma'} \exp(-\beta H_N(\sigma)) \exp(-\beta [H_N(\sigma') - H_N(\sigma)]_+) [f(\sigma) - f(\sigma')]^2 \\
 &= \min_{f \in \mathcal{H}} \tilde{Z}_{\beta, N} \sum_{m, m' \in \Gamma_N} \sum_{\substack{\sigma \in \mathcal{S}_N[m], \\ \sigma' \in \mathcal{S}_N[m']}} \mathbb{1}_{\sigma \sim \sigma'} \frac{\exp(-\beta N E(m(\sigma)))}{\tilde{Z}_{\beta, N} N} \exp(-\beta N [E(m(\sigma')) - E(m(\sigma))]_+) \\
 &\quad \times [f(\sigma) - f(\sigma')]^2 \exp(-\beta \Delta(\sigma)) \frac{\exp(-\beta [H_N(\sigma') - H_N(\sigma)]_+)}{\exp(-\beta N [E(m(\sigma')) - E(m(\sigma))]_+)} \\
 &\leq \min_{v \in \mathcal{H}} \tilde{Z}_{\beta, N} \sum_{m, m' \in \Gamma_N} \frac{\exp(-\beta N E(m))}{\tilde{Z}_{\beta, N} N} \exp(-\beta N [E(m') - E(m)]_+) [v(m) - v(m')]^2 \\
 &\quad \times \sum_{\sigma \in \mathcal{S}_N[m]} \exp(-\beta \Delta(\sigma)) \sum_{\sigma' \in \mathcal{S}_N[m']} \mathbb{1}_{\sigma \sim \sigma'} \frac{\exp(-\beta [H_N(\sigma') - H_N(\sigma)]_+)}{\exp(-\beta N [E(m') - E(m)]_+)}.
 \end{aligned} \tag{3.4}$$

We turn now to the last sum in (3.4) and call this quantity $G(\sigma, m')$. If $\sigma \sim \sigma'$, then σ and σ' differ on a single vertex, say $\ell \in [N]$, i.e. $\forall i \in [N] \setminus \{\ell\}, \sigma_i = \sigma'_i$ and $\sigma_\ell = -\sigma'_\ell$. Thus, setting $m = m(\sigma)$ and recalling (2.4) and (1.8), we can write

$$\Delta(\sigma') - \Delta(\sigma) = -\frac{2}{Np} \sum_{i:i \neq \ell} \hat{J}_{i\ell} \sigma'_i \sigma'_\ell = \frac{2}{Np} \sum_{i:i \neq \ell} \hat{J}_{i\ell} \sigma_i \sigma_\ell, \tag{3.5}$$

$$\tilde{H}_N(\sigma') - \tilde{H}_N(\sigma) = \sigma_\ell \left[\frac{2}{N} \sum_{i:i \neq \ell} \sigma_i + 2h \right] = \sigma_\ell \left[\frac{2}{N} (Nm - \sigma_\ell) + 2h \right]. \tag{3.6}$$

Moreover, using (2.2), (2.3), the definition of \hat{J}_{ij} below (2.3), the second equality in (3.5) and the first equality in (3.6) we can write

$$H_N(\sigma') - H_N(\sigma) = \tilde{H}_N(\sigma') - \tilde{H}_N(\sigma) + \Delta(\sigma') - \Delta(\sigma) = \sigma_\ell \left[\frac{2}{Np} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i + 2h \right]. \tag{3.7}$$

Due to the presence of the indicator function $\mathbb{1}_{\sigma \sim \sigma'}$, $G(\sigma, m')$ vanishes if $m' \notin \{m \pm \frac{2}{N}\}$. Moreover, we can rewrite the sum $\sum_{\sigma' \in \mathcal{S}_N[m']}$ in terms of the single vertex $\ell \in [N]$ on which σ and σ' differ. Notice that if $m(\sigma') = m + \frac{2}{N}$ then $\sigma_\ell = -1 = -\sigma'_\ell$ and if $m(\sigma') = m - \frac{2}{N}$ then $\sigma_\ell = 1 = -\sigma'_\ell$.

Therefore, calling $i^\pm(\sigma) := \{j \in [N] : \sigma_j = \pm 1\}$, and using (1.10), (3.6) and (3.7), we obtain

$$G(\sigma, m + \frac{2}{N}) = \sum_{\ell \in i^-(\sigma)} \frac{\exp\left(-\beta \left[-\frac{2}{Np} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i - 2h\right]_+\right)}{\exp\left(-\beta \left[-\frac{2}{N} (Nm + 1) - 2h\right]_+\right)} \leq N \frac{1-m}{2} e^{2\beta}, \tag{3.8}$$

$$G(\sigma, m - \frac{2}{N}) = \sum_{\ell \in i^+(\sigma)} \frac{\exp\left(-\beta \left[\frac{2}{Np} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i + 2h\right]_+\right)}{\exp\left(-\beta \left[\frac{2}{N} (Nm - 1) + 2h\right]_+\right)} \leq N \frac{1+m}{2} e^{2\beta(1+h)}. \tag{3.9}$$

To obtain the inequalities we used the fact that, for any σ in \mathcal{S}_N , the cardinalities of $i^-(\sigma)$ and $i^+(\sigma)$ are respectively $N \frac{1-m(\sigma)}{2}$ and $N \frac{1+m(\sigma)}{2}$. Moreover, for the inequality in (3.8) we used the following elementary facts holding asymptotically in N ,

$$\exp \left(-\beta \left[-\frac{2}{Np} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i - 2h \right]_+ \right) \leq 1, \tag{3.10}$$

$$\exp \left(\beta \left[-\frac{2}{N} (Nm + 1) - 2h \right]_+ \right) \leq \exp \left(\beta \left[-2m - \frac{2}{N} - 2h \right]_+ \right) \leq e^{2\beta}. \tag{3.11}$$

Similar inequalities were used to prove (3.9).

Thus, using (3.4), (3.8), (3.9) we obtain

$$\begin{aligned} & Z_{\beta,N} \text{cap} (\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) \\ & \leq \min_{v \in \mathcal{H}} \tilde{Z}_{\beta,N} \sum_{m,m' \in \Gamma_N} \frac{\exp(-\beta NE(m))}{\tilde{Z}_{\beta,N} N} \exp(-\beta N [E(m') - E(m)]_+) [v(m) - v(m')]^2 \\ & \quad \times e^{2\beta(1+h)} \sum_{\sigma \in \mathcal{S}_N[m]} \exp(-\beta \Delta(\sigma)) \left[N \frac{1+m}{2} \mathbb{1}_{m-\frac{2}{N}}(m') + N \frac{1-m}{2} \mathbb{1}_{m+\frac{2}{N}}(m') \right]. \end{aligned} \tag{3.12}$$

Using the upper bound in Corollary 2.5 with

$$\begin{aligned} g(m) &= \sum_{m' \in \Gamma_N} \frac{\exp(-\beta NE(m))}{\tilde{Z}_{\beta,N} N} \exp(-\beta N [E(m') - E(m)]_+) [v(m) - v(m')]^2 \\ & \quad \times e^{2\beta(1+h)} \left[N \frac{1+m}{2} \mathbb{1}_{m-\frac{2}{N}}(m') + N \frac{1-m}{2} \mathbb{1}_{m+\frac{2}{N}}(m') \right] \end{aligned} \tag{3.13}$$

we obtain

$$\begin{aligned} & Z_{\beta,N} \text{cap} (\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) \\ & \stackrel{P(s)}{\leq} e^{s+2\beta(1+h)+\alpha} \tilde{Z}_{\beta,N} \min_{v \in \mathcal{H}} \sum_{m,m' \in \Gamma_N} \frac{\exp(-\beta NE(m) - NI_N(m))}{\tilde{Z}_{\beta,N} N} \exp(-\beta N [E(m') - E(m)]_+) \\ & \quad \times [v(m) - v(m')]^2 \left[N \frac{1+m}{2} \mathbb{1}_{m-\frac{2}{N}}(m') + N \frac{1-m}{2} \mathbb{1}_{m+\frac{2}{N}}(m') \right] (1 + o(1)) \\ & = e^{s+2\beta(1+h)+\alpha} \tilde{Z}_{\beta,N} \min_{v \in \mathcal{H}} \sum_{m,m' \in \Gamma_N} \tilde{Q}(m) \tilde{r}(m, m') [v(m) - v(m')]^2 (1 + o(1)) \\ & = e^{s+2\beta(1+h)+\alpha} \tilde{Z}_{\beta,N} \text{cap}^{CW} (\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) (1 + o(1)), \end{aligned} \tag{3.14}$$

where we used notation (1.38) and in the middle step we used (1.13), the first equality in (1.14) and (1.20). Furthermore, we noticed that the variational form appearing in the previous formula is the Dirichlet principle (see Bovier and den Hollander [2, (7.1.29), (7.3.1)]) applied to the random walk performed by the projection of the CW model dynamics onto the magnetisation space. See Section 1.2 for the CW model.

We conclude that the minimum equals the capacity of the CW model using lumping techniques. More precisely, here we used Bovier and den Hollander [2, (9.3.6)], stating that the capacity for the dynamics projected onto the magnetisation space equals the capacity for the CW dynamics on the configuration space, which holds because of the CW model mean-field property. For reference on lumping see Bovier and den Hollander [2, Section 9.3]. \square

3.2 Asymptotics on capacity: lower bound

In this section we prove Theorem 1.6, obtaining the lower bound on the capacity of the RDCW model in terms of the capacity of the CW model. We will prove it without loss of generality, only for $m_1 < m_2 \in \Gamma_N$, because the capacity can be proven to be symmetric, using the reversibility of the dynamics.

The main idea of the proof is to find a lower bound on the capacity of the RDCW model via the Thomson principle (see e.g. Bovier and den Hollander [2, Theorem 7.37]). For $\text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2])$ it reads

$$\text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) = \sup \left\{ \frac{1}{\mathcal{D}(\bar{\Psi})} : \bar{\Psi} \in \mathcal{U}_{\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]} \right\}, \tag{3.15}$$

where we denote by $\mathcal{U}_{\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]}$ the space of all unitary antisymmetric flows from $\mathcal{S}_N[m_1]$ to $\mathcal{S}_N[m_2]$ and \mathcal{D} is defined by

$$\mathcal{D}(\psi) = \frac{1}{2} \sum_{\sigma, \sigma' \in \mathcal{S}_N} \mathbb{1}_{\sigma' \sim \sigma} \frac{\psi(\sigma, \sigma')^2}{\mu_{\beta, N}(\sigma) p_N(\sigma, \sigma')} \tag{3.16}$$

for any $\psi : \mathcal{S}_N^2 \rightarrow \mathbb{R}$ antisymmetric flow. Thus, in order to find a lower bound in terms of the capacity of the CW model we have to find a unitary flow from which we could reconstruct the CW capacity term.

For all $\sigma, \sigma' \in \mathcal{S}_N$, we define the candidate flow Ψ_N as follows

$$\Psi_N(\sigma, \sigma') = \phi_N(m(\sigma), m(\sigma')), \tag{3.17}$$

where, for all $m, m' \in \Gamma_N$,

$$\phi_N(m, m') = \begin{cases} \left[\frac{(1-m)N}{2} \exp(-NI_N(m)) \right]^{-1} & \text{if } m_1 \leq m \leq m_2 - \frac{2}{N}, m' = m + \frac{2}{N} \\ - \left[\frac{(1+m)N}{2} \exp(-NI_N(m)) \right]^{-1} & \text{if } m_1 + \frac{2}{N} \leq m \leq m_2, m' = m - \frac{2}{N} \\ 0 & \text{otherwise.} \end{cases} \tag{3.18}$$

The proof of Theorem 1.6 is postponed after two technical intermediate results which are essential for it. The following lemma allows us to use Ψ_N in the Thomson principle.

Lemma 3.1. *Let $m_1 < m_2 \in \Gamma_N$. The flow Ψ_N on \mathcal{S}_N , defined in (3.17) is a unitary antisymmetric flow from $\mathcal{S}_N[m_1]$ to $\mathcal{S}_N[m_2]$, i.e. $\Psi_N \in \mathcal{U}_{\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]}$.*

Proof. Ψ_N is antisymmetric because for all $m \in \Gamma_N$, i.e.

$$\frac{(1+m)}{2} N \exp(-NI_N(m)) = \frac{(1 - (m - \frac{2}{N}))}{2} N \exp(-NI_N(m - \frac{2}{N})). \tag{3.19}$$

Indeed, using (1.15), the right hand side of (3.19) writes

$$\begin{aligned} & \frac{(1 - (m - \frac{2}{N}))}{2} N \exp(-NI_N(m - \frac{2}{N})) = \frac{(1 - (m - \frac{2}{N}))^N}{2} \binom{N}{\frac{1 - (m - \frac{2}{N})}{2} N} \\ & = \frac{(1 - (m - \frac{2}{N}))^N}{2} \frac{N!}{\left[\frac{(1-m)N}{2} + 1 \right]! \left[\frac{(1+m)N}{2} - 1 \right]!} = \frac{N!}{\left[\frac{(1-m)N}{2} \right]! \left[\frac{(1+m)N}{2} - 1 \right]!} \\ & = \frac{(1+m)}{2} N \binom{N}{\frac{1+m}{2} N} = \frac{(1+m)}{2} N \exp(-NI_N(m)). \end{aligned} \tag{3.20}$$

Next we prove that the *Kirchhoff* law holds, i.e., for all $\sigma \in \mathcal{S}_N \setminus (\mathcal{S}_N[m_1] \cup \mathcal{S}_N[m_2])$

$$\sum_{\sigma' \in \mathcal{S}_N: \sigma \sim \sigma'} \Psi_N(\sigma, \sigma') = 0. \tag{3.21}$$

For all $\sigma \in \mathcal{S}_N$ such that $m(\sigma) \notin (m_1, m_2)$, (3.21) holds trivially being all terms zero, by (3.18). Now, for all $\sigma \in \mathcal{S}_N$ such that $m(\sigma) \in (m_1, m_2)$,

$$\begin{aligned} \sum_{\sigma' \in \mathcal{S}_N: \sigma \sim \sigma'} \Psi_N(\sigma, \sigma') &= \sum_{\substack{\sigma' \in \mathcal{S}_N: \sigma \sim \sigma', \\ m(\sigma') = m(\sigma) + \frac{2}{N}}} \phi_N(m(\sigma), m(\sigma')) + \sum_{\substack{\sigma' \in \mathcal{S}_N: \sigma \sim \sigma', \\ m(\sigma') = m(\sigma) - \frac{2}{N}}} \phi_N(m(\sigma), m(\sigma')) \\ &= \frac{(1 - m(\sigma))N}{2} \left[\frac{(1 - m(\sigma))N}{2} \exp(-NI_N(m(\sigma))) \right]^{-1} \\ &\quad - \frac{(1 + m(\sigma))N}{2} \left[\frac{(1 + m(\sigma))N}{2} \exp(-NI_N(m(\sigma))) \right]^{-1} \\ &= 0, \end{aligned} \tag{3.22}$$

where $\frac{(1 \mp m(\sigma))N}{2}$ in the second equality are the cardinalities of the set over which we were summing, namely the number of negative, respectively positive, spins in a configuration $\sigma \in \mathcal{S}_N$.

We are left to show that Ψ_N is *unitary* from $\mathcal{S}_N[m_1]$ to $\mathcal{S}_N[m_2]$, namely

$$\sum_{a \in \mathcal{S}_N[m_1]} \sum_{\sigma' \in \mathcal{S}_N: a \sim \sigma'} \Psi_N(a, \sigma') = 1 = \sum_{b \in \mathcal{S}_N[m_2]} \sum_{\sigma \in \mathcal{S}_N: \sigma \sim b} \Psi_N(\sigma, b). \tag{3.23}$$

The left hand side of (3.23) equals

$$\begin{aligned} &\sum_{a \in \mathcal{S}_N[m_1]} \sum_{\sigma' \in \mathcal{S}_N: a \sim \sigma'} \phi_N(m(a), m(\sigma')) \\ &= \sum_{a \in \mathcal{S}_N[m_1]} \sum_{\substack{\sigma' \in \mathcal{S}_N: a \sim \sigma', \\ m(\sigma') = m_1 + \frac{2}{N}}} \left[\frac{(1 - m_1)N}{2} \exp(-NI_N(m_1)) \right]^{-1} \\ &= \exp(-NI_N(m_1)) \frac{(1 - m_1)N}{2} \left[\frac{(1 - m_1)N}{2} \exp(-NI_N(m_1)) \right]^{-1} = 1. \end{aligned} \tag{3.24}$$

The right hand side of (3.23) equals

$$\begin{aligned} &\sum_{b \in \mathcal{S}_N[m_2]} \sum_{\sigma \in \mathcal{S}_N: \sigma \sim b} \phi_N(m(\sigma), m(b)) \\ &= \sum_{b \in \mathcal{S}_N[m_2]} \sum_{\substack{\sigma \in \mathcal{S}_N: \sigma \sim b, \\ m(\sigma) = m_2 - \frac{2}{N}}} \left[\frac{(1 - (m_2 - \frac{2}{N}))N}{2} \exp(-NI_N(m_2 - \frac{2}{N})) \right]^{-1} \\ &= \exp(-NI_N(m_2)) \frac{(1 + (m_2))N}{2} \left[\frac{(1 - (m_2 - \frac{2}{N}))N}{2} \exp(-NI_N(m_2 - \frac{2}{N})) \right]^{-1}. \end{aligned} \tag{3.25}$$

We use (3.19) to conclude the proof. □

Lemma 3.2. For all $\sigma \in \mathcal{S}_N$ and $m' \in \Gamma_N$, the following holds

$$\begin{aligned} \sum_{\sigma' \in \mathcal{S}_N[m']} \mathbb{1}_{\sigma' \sim \sigma} \frac{\exp\left(-\beta \left[\tilde{H}_N(\sigma') - \tilde{H}_N(\sigma)\right]_+\right)}{\exp\left(-\beta \left[H_N(\sigma') - H_N(\sigma)\right]_+\right)} \\ \leq e^{2\beta(1+h)} \left[N \frac{1+m(\sigma)}{2} \mathbb{1}_{m(\sigma)-\frac{2}{N}}(m') + N \frac{1-m(\sigma)}{2} \mathbb{1}_{m(\sigma)+\frac{2}{N}}(m') \right]. \end{aligned} \quad (3.26)$$

Proof. Let $m = m(\sigma)$. The left hand side is non-zero only if $m' \in \left\{m + \frac{2}{N}, m - \frac{2}{N}\right\}$. Recalling the definition $i^\pm(\sigma) = \{j \in [N] : \sigma_j = \pm 1\}$, if $m' = m + \frac{2}{N}$, we have

$$\begin{aligned} \sum_{\sigma' \in \mathcal{S}_N\left[m+\frac{2}{N}\right]} \mathbb{1}_{\sigma' \sim \sigma} \frac{\exp\left(-\beta \left[\tilde{H}_N(\sigma') - \tilde{H}_N(\sigma)\right]_+\right)}{\exp\left(-\beta \left[H_N(\sigma') - H_N(\sigma)\right]_+\right)} \\ = \sum_{\ell \in i^-(\sigma)} \frac{\exp\left(-\beta \left[-\frac{2p}{N}(Nm+1) - 2h\right]_+\right)}{\exp\left(-\beta \left[-\frac{2}{N} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i - 2h\right]_+\right)} \\ \leq \sum_{\ell \in i^-(\sigma)} \exp\left(\beta \left[-\frac{2}{N} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i - 2h\right]_+\right) \leq \sum_{\ell \in i^-(\sigma)} e^{2\beta} = N \frac{1-m}{2} e^{2\beta}, \end{aligned} \quad (3.27)$$

where we have used that, since $h > 0$,

$$-\frac{2}{N} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i - 2h \leq \frac{2}{N} \sum_{i:i \neq \ell} |J_{i\ell} \sigma_i| \leq \frac{2(N-1)}{N} \leq 2. \quad (3.28)$$

Similarly, if $m' = m - \frac{2}{N}$, we get

$$\begin{aligned} \sum_{\sigma' \in \mathcal{S}_N\left[m-\frac{2}{N}\right]} \mathbb{1}_{\sigma' \sim \sigma} \frac{\exp\left(-\beta \left[\tilde{H}_N(\sigma') - \tilde{H}_N(\sigma)\right]_+\right)}{\exp\left(-\beta \left[H_N(\sigma') - H_N(\sigma)\right]_+\right)} \\ = \sum_{\ell \in i^+(\sigma)} \frac{\exp\left(-\beta \left[\frac{2p}{N}(Nm-1) + 2h\right]_+\right)}{\exp\left(-\beta \left[\frac{2}{N} \sum_{i:i \neq \ell} J_{i\ell} \sigma_i + 2h\right]_+\right)} \leq N \frac{1+m}{2} e^{2\beta(1+h)}. \end{aligned} \quad (3.29)$$

□

With the previous lemmas at hand, we are now ready to prove the lower bound on the capacity.

Proof of Theorem 1.6. As we mentioned above, since the capacity is symmetric, we will prove the result only for $m_1 < m_2 \in \Gamma_N$.

Let Ψ_N be the test flow defined in (3.17), which by Lemma 3.1 is in $\mathcal{U}_{\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]}$. Thus, using the Thomson principle (3.15), we obtain the following bound

$$\text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) \geq \frac{1}{\mathcal{D}(\Psi_N)}. \quad (3.30)$$

Therefore, we are interested in upper bounds on $\mathcal{D}(\Psi_N)$ which, using (1.3), (1.5) and (2.4), can be written as follows

$$\mathcal{D}(\Psi_N) = \frac{1}{2}N \sum_{\sigma, \sigma' \in \mathcal{S}_N} \mathbb{1}_{\sigma' \sim \sigma} \frac{\phi_N(m(\sigma), m(\sigma'))^2}{\exp(-\beta(\tilde{H}_N(\sigma) + \Delta(\sigma)))} \frac{Z_{\beta, N}}{\exp(-\beta[H_N(\sigma') - H_N(\sigma)]_+)} \tag{3.31}$$

By multiplying and dividing by $\exp(-\beta[\tilde{H}_N(\sigma') - \tilde{H}_N(\sigma)]_+) \tilde{Z}_{\beta, N}$, and using (1.10), (1.11) and (1.18), we get

$$\begin{aligned} \mathcal{D}(\Psi_N) &= N \frac{Z_{\beta, N}}{2\tilde{Z}_{\beta, N}} \sum_{m, m' \in \Gamma_N} \frac{\phi_N(m, m')^2}{\tilde{Q}_{\beta, N}(m) \exp(NI_N(m)) \exp(-\beta N[E(m') - E(m)]_+)} \\ &\quad \times \sum_{\sigma \in \mathcal{S}_N[m]} \exp(\beta\Delta(\sigma)) \sum_{\sigma' \in \mathcal{S}_N[m']} \mathbb{1}_{\sigma' \sim \sigma} \frac{\exp(-\beta[\tilde{H}_N(\sigma') - \tilde{H}_N(\sigma)]_+)}{\exp(-\beta[H_N(\sigma') - H_N(\sigma)]_+)} \\ &\leq N \frac{Z_{\beta, N}}{2\tilde{Z}_{\beta, N}} e^{2\beta(1+h)} \sum_{m, m' \in \Gamma_N} \frac{\phi_N(m, m')^2}{\tilde{Q}_{\beta, N}(m) \exp(NI_N(m)) \exp(-\beta N[E(m') - E(m)]_+)} \\ &\quad \times \left[N \frac{1+m}{2} \mathbb{1}_{m-\frac{2}{N}}(m') + N \frac{1-m}{2} \mathbb{1}_{m+\frac{2}{N}}(m') \right] \sum_{\sigma \in \mathcal{S}_N[m]} \exp(\beta\Delta(\sigma)), \end{aligned} \tag{3.32}$$

where we used Lemma 3.2 to bound the sum over σ' , uniformly in $\sigma \in \mathcal{S}_N[m]$. Then, to bound the remaining sum over σ , we use the upper bound in Corollary 2.5 (in the version with $e^{\beta\Delta(\sigma)}$, motivated by the Remark therein) with

$$\begin{aligned} g(m) &= \sum_{m' \in \Gamma_N} \frac{\phi_N(m, m')^2}{\tilde{Q}_{\beta, N}(m) \exp(NI_N(m)) \exp(-\beta N[E(m') - E(m)]_+)} \\ &\quad \times \left[N \frac{1+m}{2} \mathbb{1}_{m-\frac{2}{N}}(m') + N \frac{1-m}{2} \mathbb{1}_{m+\frac{2}{N}}(m') \right], \end{aligned} \tag{3.33}$$

obtaining, with notation (1.38),

$$\begin{aligned} \mathcal{D}(\Psi_N) &\stackrel{P(s)}{\leq} N \frac{Z_{\beta, N}}{2\tilde{Z}_{\beta, N}} \sum_{m, m' \in \Gamma_N} \frac{e^{s+2\beta(1+h)+\alpha} \phi_N(m, m')^2 \exp(-NI_N(m))}{\tilde{Q}_{\beta, N}(m) \exp(NI_N(m)) \exp(-\beta N[E(m') - E(m)]_+)} \\ &\quad \times \left[N \frac{1+m}{2} \mathbb{1}_{m-\frac{2}{N}}(m') + N \frac{1-m}{2} \mathbb{1}_{m+\frac{2}{N}}(m') \right] (1 + o(1)) \\ &= \frac{Z_{\beta, N}}{2\tilde{Z}_{\beta, N}} e^{s+2\beta(1+h)+\alpha} \sum_{m, m' \in \Gamma_N} \frac{\phi_N(m, m')^2 \exp(-2NI_N(m))}{\tilde{Q}_{\beta, N}(m) \tilde{r}_N(m, m')} \\ &\quad \times \left[N \frac{1+m}{2} \mathbb{1}_{m-\frac{2}{N}}(m') + N \frac{1-m}{2} \mathbb{1}_{m+\frac{2}{N}}(m') \right]^2 (1 + o(1)), \end{aligned} \tag{3.34}$$

where in the equality we only used (1.20).

Now we first substitute ϕ_N defined in (3.18) into (3.34) and then use reversibility to

obtain

$$\begin{aligned}
 \mathcal{D}(\Psi_N) &\stackrel{P(s)}{\leq} \frac{Z_{\beta,N}}{2\tilde{Z}_{\beta,N}} e^{s+2\beta(1+h)+\alpha} \sum_{\substack{m_1 \leq m < m_2, \\ m \in \Gamma_N}} \frac{1}{\tilde{Q}_{\beta,N}(m) \tilde{r}_N(m, m + \frac{2}{N})} (1 + o(1)) \\
 &+ \frac{Z_{\beta,N}}{2\tilde{Z}_{\beta,N}} e^{s+2\beta(1+h)+\alpha} \sum_{\substack{m_1 < m \leq m_2, \\ m \in \Gamma_N}} \frac{1}{\tilde{Q}_{\beta,N}(m - \frac{2}{N}) \tilde{r}_N(m - \frac{2}{N}, m)} (1 + o(1)) \quad (3.35) \\
 &= \frac{Z_{\beta,N}}{\tilde{Z}_{\beta,N}} e^{s+2\beta(1+h)+\alpha} \sum_{m_1 \leq m < m_2} \frac{1}{\tilde{Q}_{\beta,N}(m) \tilde{r}_N(m, m + \frac{2}{N})} (1 + o(1)),
 \end{aligned}$$

where the last equality follows noticing that the two sums in the previous step are equal. Therefore, by (3.30) and (3.35), we obtain

$$\begin{aligned}
 Z_{\beta,N} \text{cap}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) &\geq \frac{Z_{\beta,N}}{\mathcal{D}(\Psi_N)} \\
 &\stackrel{P(s)}{\geq} \tilde{Z}_{\beta,N} e^{-s-2\beta(1+h)-\alpha} \left[\sum_{m_1 \leq m < m_2} \frac{1}{\tilde{Q}_{\beta,N}(m) \tilde{r}_N(m, m + \frac{2}{N})} \right]^{-1} (1 + o(1)) \quad (3.36) \\
 &= \tilde{Z}_{\beta,N} e^{-s-2\beta(1+h)-\alpha} \text{cap}^{\text{CW}}(\mathcal{S}_N[m_1], \mathcal{S}_N[m_2]) (1 + o(1)),
 \end{aligned}$$

where we used notation (1.38) and we noticed that the inverse of the expression appearing in brackets in (3.36) gives exactly the capacity for the CW model. Indeed, that expression gives exactly the capacity for the one-dimensional random walk in Γ_N which is the projection of the CW dynamics onto the magnetisation space Γ_N (see the formula for the capacity in Bovier and den Hollander [2, Section 7.1.4, (7.1.60)]). Using lumping techniques exactly as at the end of the proof of Theorem 1.5 (end of Section 3.1), we have that the aforementioned capacity equals the CW capacity. \square

4 Estimates on the harmonic function

As pointed out in Section 1.4, the proof of Theorem 1.4 relies on sharp estimates on capacities, carried out in Section 3, and estimates on the harmonic function. We entirely devote this section to obtain asymptotic upper and lower bounds on the numerator in (1.40), which is given by the following sum

$$\sum_{\sigma \in \mathcal{S}_N} \mu_{\beta,N}(\sigma) h_{m_-, m_+}^N(\sigma), \quad (4.1)$$

that is to give the proof of Theorem 1.7.

In order to control the sum (4.1), one generally uses a renewal argument which relies again on estimates over capacities. However, in our case this is not possible, due to the fact that capacities of single spins are too small.

We first prove the upper bound and then give some details about how to prove the lower bound, which is very similar and more straightforward. Our proof follows Bianchi, Bovier and Ioffe [1, Section 6].

4.1 Notation and decomposition of the space

Before starting with the proof, we introduce some notation. We refer to Figure 1 below for a better visual understanding of the objects we are defining.

Recall that we denote by m_+ the global minimum, by m_- the local minimum, and by m^* the local maximum of $f_\beta(\cdot)$ in $[-1, 1]$, where $f_\beta(\cdot) = \lim_{N \rightarrow \infty} f_{\beta,N}(\cdot)$, defined in (1.14). We want to decompose the space Γ_N (and eventually the set of spin configurations \mathcal{S}_N)

according to the values of f_β . The notation and the decomposition are organised in 4 steps.

Step 1. First, let $\delta > 0$ be small in a way which will become clear later, and define the set

$$U_\delta = \{m \in [-1, 1] : f_\beta(m) \leq f_\beta(m_-) + \delta\}. \quad (4.2)$$

We write $U_\delta^c = [-1, 1] \setminus U_\delta$ and we denote by $U_\delta(m)$ the connected component of U_δ containing m . Note that $\{m_-, m_+\} \in U_\delta$. In general, $U_\delta(m_-)$ and $U_\delta(m_+)$ may have non empty intersection, but we choose δ such that $m^* \notin U_\delta$, implying that U_δ is partitioned by the disjoint sets $U_\delta(m_-)$ and $U_\delta(m_+)$. For this to hold, it suffices to take $\delta < f_\beta(m^*) - f_\beta(m_-)$. Moreover, we choose δ also such that $-1 \notin U_\delta(m_-)$. For this to hold, it suffices to take $\delta < f_\beta(-1) - f_\beta(m_-)$. Thus, we choose $\delta < \min(f_\beta(-1), f_\beta(m^*)) - f_\beta(m_-)$. These conditions are needed to prove (4.11) below.

Let us denote by m_δ the unique point in (m^*, m_+) such that

$$f_\beta(m_\delta) = f_\beta(m_-) + \delta. \quad (4.3)$$

Step 2. With δ chosen as above, we define a sequence $(\delta_N)_{N \in \mathbb{N}}$, converging to δ from below, such that the left extreme of $U_{\delta_N}(m_+)$ is in Γ_N . Specifically, we define δ_N as follows:

$$\delta_N = \max \{\bar{\delta} \in (0, \delta] : \exists m \in U_{\delta_N}(m_+) \cap \Gamma_N \setminus [m_+, 1] \text{ s.t. } f_\beta(m) = f_\beta(m_-) + \bar{\delta}\}, \quad (4.4)$$

for N sufficiently large. Moreover, set

$$U_{\delta_N} = U_{\delta_N} \cap \Gamma_N, \quad U_{\delta_N}^c = \Gamma_N \setminus U_{\delta_N} \quad \text{and} \quad U_{\delta_N}(m) = U_{\delta_N}(m) \cap \Gamma_N, \quad (4.5)$$

for all $m \in [-1, 1]$. Thus, we have the partitions

$$\Gamma_N = U_{\delta_N}(m_-) \cup U_{\delta_N}(m_+) \cup U_{\delta_N}^c \quad (4.6)$$

and

$$\mathcal{S}_N = \mathcal{S}_N[U_{\delta_N}(m_-)] \cup \mathcal{S}_N[m_+(N)] \cup \mathcal{S}_N[U_{\delta_N}^c] \cup \mathcal{S}_N[U_{\delta_N}(m_+) \setminus \{m_+(N)\}]. \quad (4.7)$$

Remark 4.1. Notice that, for N sufficiently large, $U_{\delta_N}(m_-(N)) = U_{\delta_N}(m_-)$ and $U_{\delta_N}(m_+(N)) = U_{\delta_N}(m_+)$. Furthermore, with these definitions, $m_{\delta_N} \in U_{\delta_N}$ and it is the left extreme of $U_{\delta_N}(m_+)$.

Step 3. Let $\varepsilon > 0$ be arbitrarily small (the choice of ε will be relevant in Section 4.2). We denote by m_ε the only point in a small left neighbourhood of m_+ , more precisely in $U_\delta(m_+) \setminus [m_+, 1]$, such that

$$f_\beta(m_\varepsilon) = f_\beta(m_+) + \varepsilon. \quad (4.8)$$

Let us define an ε -dependent parameter $\theta > 0$ by

$$\theta = m_+ - m_\varepsilon. \quad (4.9)$$

Step 4. Similarly to Step 2, fixed $\varepsilon > 0$, we want to define a sequence $(\varepsilon_N)_{N \in \mathbb{N}}$ converging to ε from below such that m_{ε_N} is in Γ_N . More precisely, we define ε_N as follows

$$\varepsilon_N = \max \{\bar{\varepsilon} \in (0, \varepsilon] : \exists m \in U_{\delta_N}(m_+) \setminus [m_+, 1] \text{ s.t. } f_\beta(m) = f_\beta(m_+) + \bar{\varepsilon}\}. \quad (4.10)$$

We will use later that $m_{\varepsilon_N} \in U_{\delta_N}(m_+)$ and it satisfies $f_\beta(m_{\varepsilon_N}) = f_\beta(m_+) + \varepsilon_N$.

Moreover, given $\varepsilon > 0$, we define the sequence $(\theta_N)_{N \in \mathbb{N}}$, analogously to (4.9), by setting $\theta_N = m_+(N) - m_{\varepsilon_N}$. θ_N plays an important role in Lemma 4.4 below. Notice that $\lim_{N \rightarrow \infty} \theta_N = \theta$ and, if $m_+ \neq m_+(N)$, then $f(m_{\varepsilon_N}) - f(m_+(N)) \neq \varepsilon_N$.

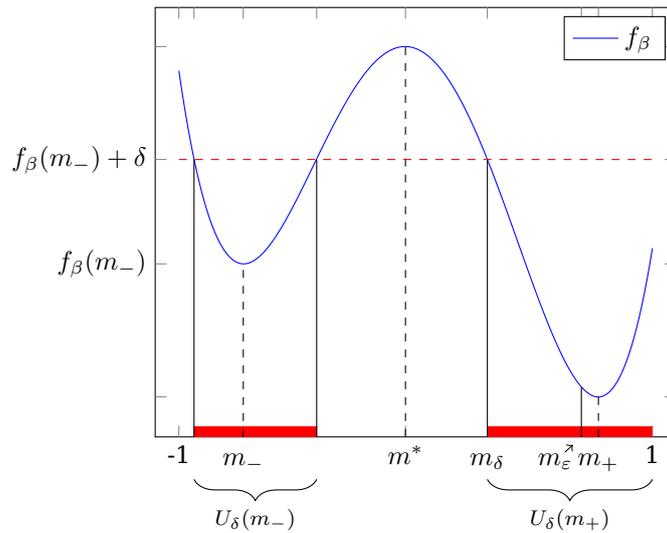


Figure 1: Graph of f_β and decomposition of the magnetisation space $[-1, 1]$: the two intervals $U_\delta(m_-)$ and $U_\delta(m_+)$ around the two minima are drawn, together with the special points m_δ, m_ϵ . U_δ is painted in red.

4.2 Upper bound on the harmonic sum

In this section we prove the first part of Theorem 1.7 by giving an upper bound on the harmonic sum in (4.1).

We will estimate the contribution of each set of the partition in (4.7) to the sum in (4.1). As one expects, the only relevant contribution will be given by the terms in $\mathcal{S}_N[U_{\delta,N}(m_-)]$. Indeed, $\mu_{\beta,N}$ is very small in $\mathcal{S}_N[U_{\delta,N}^c]$ while h_{m_-,m_+}^N is very small in $\mathcal{S}_N[U_{\delta,N}(m_+)]$ and we will see the two contributions on these two sets turn out to be irrelevant.

The main ingredients in the proof of the upper bound are Corollary 2.7 and Lemma 4.2 below. The proof of the latter result is quite technical and it is postponed to Section 4.3.

Proof of Theorem 1.7. Upper bound. We are ready to start estimating the contributions of each disjoint set of the partition in (4.7) to the sum in (4.1).

Part 1. Sum on $\mathcal{S}_N[U_{\delta,N}(m_-)]$. This will be the relevant part. Using first that $h_{m_-,m_+}^N(\sigma) \leq 1$, (2.1) and (2.19) of Corollary 2.7 with $\bar{g}(m) = \mathbb{1}_{m \in U_{\delta,N}(m_-)}(m)$ we obtain

$$\begin{aligned} \sum_{\sigma \in \mathcal{S}_N[U_{\delta,N}(m_-)]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) &\leq \sum_{m \in U_{\delta,N}(m_-)} \mathcal{Q}_{\beta,N}(m) \\ &\stackrel{P(s)}{\leq} \frac{e^{s+\alpha}(1+o(1))}{Z_{\beta,N}} \sum_{m \in U_{\delta,N}(m_-)} \exp(-\beta N f_\beta(m)) \sqrt{\frac{2}{\pi N(1-m^2)}} \\ &= \frac{e^{s+\alpha}(1+o(1)) \exp(-\beta N f_\beta(m_-))}{Z_{\beta,N} \sqrt{(1-m_-^2) \beta f_\beta''(m_-)}}. \end{aligned} \tag{4.11}$$

In the second line we used our assumption of δ_N being small enough such that $-1 \notin U_{\delta,N}(m_-)$ (see Section 4.1, Step 1). To obtain the last equality we first approximated, for N sufficiently large, the sum with an integral and then applied the saddle point method (see, for instance, de Bruijn [7, Chp 5.7]), where m_- is the maximum point of $-\beta f_\beta$ on

the considered domain. Notice that here we use the fact that $m^* \notin U_{\delta,N}(m_-)$, which holds again for δ_N small enough (see Section 4.1, Step 1). More precisely,

$$\begin{aligned} & \sum_{m \in U_{\delta,N}(m_-)} \exp\left(-\beta N f_\beta(m)\right) \frac{1}{\sqrt{(1-m^2)}} \\ & \approx \frac{N}{2} \int_a^b \exp\left(-\beta N f_\beta(x)\right) \frac{1}{\sqrt{(1-x^2)}} dx \\ & = \exp\left(-\beta N f_\beta(m_-)\right) \frac{1}{\sqrt{(1-m_-^2)}} \sqrt{\frac{\pi N}{2\beta f_\beta''(m_-)}} (1 + o(1)), \end{aligned} \tag{4.12}$$

where $-1 < a, b \in \Gamma_N$ are the left and right extremes of $U_{\delta,N}(m_-)$, respectively.

Part 2. Sum on $\mathcal{S}_N[m_+(N)]$. Being by definition $h_{m_-,m_+}^N(\sigma) = 0$ for all $\sigma \in \mathcal{S}_N[m_+(N)]$, we trivially have

$$\sum_{\sigma \in \mathcal{S}_N[m_+(N)]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) = 0. \tag{4.13}$$

Part 3. Sum on $\mathcal{S}_N[U_{\delta,N}^c]$.

Using $h_{m_-,m_+}^N \leq 1$ and (2.1), we have

$$\begin{aligned} & \sum_{\sigma \in \mathcal{S}_N[U_{\delta,N}^c]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \leq \sum_{\sigma \in \mathcal{S}_N[U_{\delta,N}^c]} \mu_{\beta,N}(\sigma) = \sum_{m \in U_{\delta,N}^c} \mathcal{Q}_{\beta,N}(m) \\ & = \sum_{m \in U_{\delta,N}^c \setminus \{1,-1\}} \mathcal{Q}_{\beta,N}(m) + \sum_{m \in U_{\delta,N}^c \cap \{1,-1\}} \mathcal{Q}_{\beta,N}(m). \end{aligned} \tag{4.14}$$

We bound the right hand side using (2.19) of Corollary 2.7 with $\bar{g}(m) = \mathbb{1}_{m \in U_{\delta,N}^c}(m)$ obtaining

$$\begin{aligned} & \sum_{\sigma \in \mathcal{S}_N[U_{\delta,N}^c]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \\ & \leq \frac{e^{s+\alpha} (1 + o(1))}{Z_{\beta,N}} \sum_{m \in U_{\delta,N}^c \setminus \{1,-1\}} \exp\left(-\beta N f_\beta(m)\right) \sqrt{\frac{2}{\pi N (1-m^2)}} \\ & \quad + \frac{e^{s+\alpha} (1 + o(1))}{Z_{\beta,N}} \sum_{m \in U_{\delta,N}^c \cap \{1,-1\}} \exp\left(-\beta N f_\beta(m)\right) \\ & \leq \frac{e^{s+\alpha} (1 + o(1))}{Z_{\beta,N}} \exp\left(-\beta N (f_\beta(m_-) + \delta_N)\right) \left(\sqrt{\frac{2}{\pi N}} \sum_{m \in U_{\delta,N}^c \setminus \{1,-1\}} \frac{1}{\sqrt{(1-m^2)}} + 2 \right), \end{aligned} \tag{4.15}$$

where in the last inequality we used the bound $f_\beta(m) \geq f_\beta(m_-) + \delta_N$ given by the definition of $U_{\delta,N}^c$ (see (4.2)).

Part 4. Sum on $\mathcal{S}_N[U_{\delta,N}(m_+) \setminus \{m_+(N)\}]$. Using (1.32) and the fact that, for any $\sigma \in \mathcal{S}_N$ such that $m(\sigma) > m_+(N)$, $\mathbb{P}_\sigma(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]}) = 0$, we get

$$\sum_{\sigma \in \mathcal{S}_N[U_{\delta,N}(m_+) \setminus \{m_+(N)\}]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) = \sum_{\sigma \in \mathcal{S}_N[[m_{\delta_N}, m_+(N)]]} \mu_{\beta,N}(\sigma) \mathbb{P}_\sigma(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]}). \tag{4.16}$$

Thus, applying Lemma 4.2 below, the following holds for any $\gamma \in (0, 1)$

$$\begin{aligned} & \sum_{\sigma \in \mathcal{S}_N[U_{\delta_N}(m_+) \setminus \{m_+(N)\}]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \\ & \leq \exp\left(-\beta N(1-\gamma)f_\beta(m_-)\right) \sum_{m \in [m_{\delta_N}, m_+(N)]} \mathcal{Q}_{\beta,N}(m) \left[\exp\left(\beta N(1-\gamma)f_\beta(m)\right) \right. \\ & \quad \left. + \exp\left(\beta N(1-\gamma)(f_\beta(m_+) + 3\varepsilon_N) + N\ell_N(\theta_N)\right) \right] \exp\left(-\beta N(1-\gamma)\delta_N\right)(1+o(1)). \end{aligned} \tag{4.17}$$

We use (2.19) of Corollary 2.7 with $\bar{g}(m)$ defined by

$$\bar{g}(m) = \left[\exp\left(\beta N(1-\gamma)f_\beta(m)\right) + \exp\left(\beta N(1-\gamma)(f_\beta(m_+) + 3\varepsilon_N) + N\ell_N(\theta_N)\right) \right], \tag{4.18}$$

for $m \in [m_{\delta_N}, m_+(N))$ and $\bar{g}(m) = 0$ for $m \in \Gamma_N \setminus [m_{\delta_N}, m_+(N))$. Thus, we obtain

$$\begin{aligned} & \sum_{\sigma \in \mathcal{S}_N[U_{\delta_N}(m_+) \setminus \{m_+(N)\}]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \\ & \stackrel{P(s)}{\leq} \frac{e^{s+\alpha}(1+o(1))}{Z_{\beta,N}} \exp\left(-\beta N(1-\gamma)(f_\beta(m_-) + \delta_N)\right) \sum_{m \in [m_{\delta_N}, m_+(N))} \exp\left(-\beta N f_\beta(m)\right) \\ & \quad \times \sqrt{\frac{2}{\pi N(1-m^2)}} \left[\exp\left(\beta N(1-\gamma)f_\beta(m)\right) + \exp\left(\beta N(1-\gamma)(f_\beta(m_+) + 3\varepsilon_N) + N\ell_N(\theta_N)\right) \right] \\ & \leq \frac{e^{s+\alpha}(1+o(1))}{Z_{\beta,N}} \exp\left(-\beta N(1-\gamma)(f_\beta(m_-) + \delta_N)\right) \sqrt{\frac{2N}{\pi(1-m_+^2)}} \\ & \quad \times \left[\exp\left(-\gamma\beta N f_\beta(m_+)\right) + \exp\left(\beta N(1-\gamma)(f_\beta(m_+) + 3\varepsilon_N) + N\ell_N(\theta_N) - \beta N f_\beta(m_+)\right) \right] \\ & = \frac{e^{s+\alpha}(1+o(1))}{Z_{\beta,N}} \exp\left(-\beta N f_\beta(m_-)\right) \sqrt{\frac{2N}{\pi(1-m_+^2)}} \exp\left(-\gamma\beta N[f_\beta(m_+) - f_\beta(m_-)]\right) \\ & \quad \times \exp\left(-\beta N(1-\gamma)(\delta_N - 3\varepsilon_N) + N\ell_N(\theta_N)\right) \left[\exp\left(-\beta N(1-\gamma)3\varepsilon_N - N\ell_N(\theta_N)\right) + 1 \right] \\ & \leq \frac{e^{s+\alpha}(1+o(1))}{Z_{\beta,N}} \exp\left(-\beta N f_\beta(m_-)\right) \sqrt{\frac{2}{\pi(1-m_+^2)}} \\ & \quad \times \exp\left[-\beta N\left(\gamma[f_\beta(m_+) - f_\beta(m_-)] + (1-\gamma)(\delta_N - 3\varepsilon_N) - \frac{1}{\beta}\ell_N(\theta_N) - \varepsilon_N\right)\right]. \end{aligned} \tag{4.19}$$

In the last step we embedded $[\exp(-\beta N(1-\gamma)3\varepsilon_N - N\ell_N(\theta_N)) + 1]$ in the already present $(1+o(1))$ and bounded \sqrt{N} by $\exp(-\beta N(-\varepsilon_N))$, because for N large enough $\frac{\log(N)}{2\beta N} \leq \varepsilon_N$ (which converges to $\varepsilon > 0$, see Step 4 in Section 4.1).

Now we prove that this part is not relevant compared to the right hand side of (4.11). In particular, we show that, for a certain choice of γ ,

$$c_N = \gamma[f_\beta(m_+) - f_\beta(m_-)] + (1-\gamma)(\delta_N - 3\varepsilon_N) - \frac{1}{\beta}\ell_N(\theta_N) - \varepsilon_N \tag{4.20}$$

is positive and its limit,

$$\lim_{N \rightarrow \infty} c_N = \gamma[f_\beta(m_+) - f_\beta(m_-)] + (1-\gamma)(\delta - 3\varepsilon) - \frac{\theta}{2\beta}(\log(2) + 3 - \log(1-m_+)) - \varepsilon, \tag{4.21}$$

is positive and finite. In order to achieve this, we choose $\gamma \in (0, 1)$ small enough, such that c_N and its limit are positive, definitely in N . In particular, we want to impose

$$0 < \gamma < \frac{\delta_N - 4\varepsilon_N - \frac{1}{\beta}\ell_N(\theta_N)}{f_\beta(m_-) - f_\beta(m_+) + \delta_N - 3\varepsilon_N} < 1, \tag{4.22}$$

definitely in N , and

$$0 < \gamma < \frac{\delta - 4\varepsilon - \frac{1}{\beta} \lim_{N \rightarrow \infty} \ell_N(\theta)}{f_\beta(m_-) - f_\beta(m_+) + \delta - 3\varepsilon} < 1. \tag{4.23}$$

First, we notice that it is easy to check that the previous quantities are strictly smaller than 1. Second, we want to show that a strictly positive γ satisfying (4.22)-(4.23) exists. Note that $\ell_N(\theta_N)$, defined in (4.37), has the following trivial upper bound for every N ,

$$\ell_N(\theta_N) \leq \theta_N (\beta + \log 2 + O(\theta_N)). \tag{4.24}$$

Thus, a sufficient condition is to choose, for N large enough, $\gamma \geq \gamma_0$, where

$$\gamma_0 = \frac{\delta - 4\varepsilon - \theta \left(1 + \frac{\log 2}{\beta} + O(\theta)\right)}{f_\beta(m_-) - f_\beta(m_+) + \delta} \tag{4.25}$$

is clearly strictly positive. Indeed, we can choose $\varepsilon > 0$ sufficiently small for the numerator on the left hand side of (4.25) to be positive, while θ is small accordingly to ε (see Section 4.1). We conclude by obtaining, for N sufficiently large,

$$\sum_{\sigma \in \mathcal{S}_N[U_{\delta,N}(m_+) \setminus \{m_+(N)\}]} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \stackrel{P(s)}{\leq} \frac{e^{s+\alpha+o(1)}}{Z_{\beta,N}} \exp(-\beta N(f_\beta(m_-) + c_N)) \sqrt{\frac{2}{\pi(1-m_+^2)}}, \tag{4.26}$$

where $0 < c_N = O(1)$.

Conclusion.

With the previous bounds at hand, we are now ready to conclude the proof of the upper bound. Decomposing the sum over \mathcal{S}_N using (4.7), and inserting the estimates we computed above into (4.1), we obtain

$$\begin{aligned} & \sum_{\sigma \in \mathcal{S}_N} \mu_{\beta,N}(\sigma) h_{m_-,m_+}^N(\sigma) \\ & \leq \frac{e^{s+\alpha} (1 + o(1))}{Z_{\beta,N}} \exp(-\beta N f_\beta(m_-)) \left[e^{-\beta N \delta_N} \left(\sqrt{\frac{2}{\pi N}} \sum_{m \in U_{\delta,N}^c \setminus \{1,-1\}} \frac{1}{\sqrt{(1-m^2)}} + 2 \right) \right. \\ & \quad \left. + \sqrt{\frac{2}{\pi(1-m_+^2)}} e^{-\beta N c_N} + \frac{1}{\sqrt{(1-m_-^2) \beta f_\beta''(m_-)}} \right] \\ & \leq \frac{e^{s+\alpha}}{Z_{\beta,N}} \exp(-\beta N f_\beta(m_-)) \frac{1}{\sqrt{(1-m_-^2) \beta f_\beta''(m_-)}} (1 + o(1)), \end{aligned} \tag{4.27}$$

concluding the proof. □

4.3 Some technical results

In this section we prove Lemma 4.2, which is pivotal in obtaining the upper bound in Theorem 1.7 (see (4.17)). The proof is quite involved, therefore we split it into subsequent technical results. Before starting the proof, we give a brief outline of this section. First, we state Lemma 4.2 and prove it via Lemmas 4.4, 4.5 and 4.6, which follow later on. Second, we give the proof of Lemmas 4.4 and 4.5. The latter relies on Lemma 4.6, which we subsequently prove using Lemma 4.7. We conclude the section proving Lemma 4.7. Throughout this section we will use the notation introduced in Section 4.1.

Lemma 4.2. For all $\sigma \in \mathcal{S}_N [[m_{\delta_N}, m_+(N)]]$, for all $\gamma \in (0, 1)$ and $\varepsilon > 0$,

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) &\leq \exp \left(-\beta N(1-\gamma)[f_\beta(m_-) + \delta_N] \right) (1 + o(1)) \\ &\times \left[\exp \left(\beta N(1-\gamma)f_\beta(m(\sigma)) \right) + \exp \left(\beta N(1-\gamma)[f_\beta(m_+) + 3\varepsilon_N] + N\ell_N(\theta_N) \right) \right], \end{aligned} \tag{4.28}$$

where $\ell_N(\cdot)$ is defined in (4.37).

Proof. For all $\sigma \in \mathcal{S}_N [[m_{\delta_N}, m_+(N)]]$, we have

$$\begin{aligned} &\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \\ &= \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]}, \tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_{\varepsilon_N}]} \right) \\ &\quad + \sum_{\eta \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]}, \tau_\eta < \tau_{\mathcal{S}_N[\{m_{\varepsilon_N}, m_-(N), m_+(N)\}] } \right) \\ &= \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}] } \right) \\ &\quad + \sum_{\eta \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \mid \tau_\eta < \tau_{\mathcal{S}_N[\{m_{\varepsilon_N}, m_-(N), m_+(N)\}] } \right) \\ &\quad \times \mathbb{P}_\sigma \left(\tau_\eta < \tau_{\mathcal{S}_N[\{m_{\varepsilon_N}, m_-(N), m_+(N)\}] } \right), \end{aligned} \tag{4.29}$$

where we notice that,

$$\begin{aligned} &\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \mid \tau_\eta < \tau_{\mathcal{S}_N[\{m_{\varepsilon_N}, m_-(N), m_+(N)\}] } \right) \\ &= \mathbb{P}_\eta \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right). \end{aligned} \tag{4.30}$$

Using the Markov property and taking the maximum of the first factor out of the sum, we have that, for all $\sigma \in \mathcal{S}_N [[m_{\delta_N}, m_+(N)]]$,

$$\begin{aligned} &\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \\ &\leq \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}] } \right) \\ &\quad + \left(\max_{\eta \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\eta \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \right) \sum_{\eta \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\sigma \left(\tau_\eta < \tau_{\mathcal{S}_N[\{m_{\varepsilon_N}, m_-(N), m_+(N)\}] } \right) \\ &= \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}] } \right) \\ &\quad + \left(\max_{\eta \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\eta \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \right) \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_{\varepsilon_N}]} < \tau_{\mathcal{S}_N[\{m_-(N), m_+(N)\}] } \right) \\ &\leq \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}] } \right) \\ &\quad + \left(\max_{\eta \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\eta \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \right) \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_{\varepsilon_N}]} < \tau_{\mathcal{S}_N[m_+(N)]} \right). \end{aligned} \tag{4.31}$$

We first consider the case $\sigma \in \mathcal{S}_N [m_{\varepsilon_N}]$. By Lemma 4.4, we get

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) &\leq \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}] } \right) \\ &\quad + \left(\max_{\eta \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\eta \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \right) \left(1 - e^{-N\ell_N(\theta_N)} (1 + o(1)) \right). \end{aligned} \tag{4.32}$$

Taking the maximum over σ and noticing that the same term appears in both right and left hand side of the inequality, we obtain

$$\begin{aligned} & \max_{\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \\ & \leq \max_{\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) e^{N\ell_N(\theta_N)} (1 + o(1)) \\ & \leq \exp \left(-\beta N (1 - \gamma) [f_\beta(m_-) + \delta_N - f_\beta(m_{\varepsilon_N} - \frac{2}{N}) - \varepsilon_N] - N\ell_N(\theta_N) \right) (1 + o(1)), \end{aligned} \tag{4.33}$$

where we used Lemma 4.5.

By Taylor expansion of $f_\beta(m_{\varepsilon_N} - \frac{2}{N})$ and definition of m_{ε_N} , we get

$$\begin{aligned} & \max_{\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) \\ & \leq \exp \left(-\beta N (1 - \gamma) [f_\beta(m_-) + \delta_N - 3\varepsilon_N - f_\beta(m_+)] - N\ell_N(\theta_N) \right) (1 + o(1)), \end{aligned} \tag{4.34}$$

where the last inequality holds for N sufficiently large. Here we bounded the Taylor expansion error $O(\frac{1}{N})$ with ε_N , which converges to $\varepsilon > 0$ (see Step 4 in Section 4.1).

Now we consider the case where $\sigma \in \mathcal{S}_N[[m_{\delta_N}, m_+(N)) \setminus \{m_{\varepsilon_N}\}]$. Going back to (4.31) and using again (4.34), we obtain

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right) & \leq \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) \\ & \quad + \exp \left(-\beta N (1 - \gamma) [f_\beta(m_-) + \delta_N - 3\varepsilon_N - f_\beta(m_+)] - N\ell_N(\theta_N) \right) (1 + o(1)) \\ & \leq \exp \left(-\beta N (1 - \gamma) [f_\beta(m_-) + \delta_N] \right) (1 + o(1)) \\ & \quad \times \left[\exp \left(\beta N (1 - \gamma) f_\beta(m(\sigma)) \right) + \exp \left(\beta N (1 - \gamma) [f_\beta(m_+) + 3\varepsilon_N] + N\ell_N(\theta_N) \right) \right]. \end{aligned} \tag{4.35}$$

In the last inequality we used Lemma 4.6, which holds for $\sigma \in \mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$, and that $\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) = 0$ for all $\sigma \in \mathcal{S}_N[(m_{\varepsilon_N}, m_+(N))]$. \square

Remark 4.3. In Lemma 4.2 one might try to further bound the right hand side of (4.28) using that $f_\beta(m(\sigma))$ is bounded by $f_\beta(m_{\delta_N}) = f_\beta(m_-) + \delta_N$. This would yield to the trivial upper bound 1 on $\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[m_+(N)]} \right)$, which is not sufficient for our purpose of proving that the second term in (4.27) is negligible with respect to the last one. The way to go is, therefore, to keep the dependence on $m(\sigma)$ in order to obtain later a more suitable bound, uniform in m , by exploiting the smallness of $\mathcal{Q}_{\beta, N}(m(\sigma))$ in (4.17) and (4.19).

In order for (4.32) to be true, we have to prove the following result.

Lemma 4.4. For all $\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]$, for ε sufficiently small and for N sufficiently large,

$$\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_+(N)]} < \tau_{\mathcal{S}_N[m_{\varepsilon_N}]} \right) \geq e^{-N\ell_N(\theta_N)} (1 + o(1)), \tag{4.36}$$

where $\ell_N : \mathbb{R} \rightarrow \mathbb{R}$ is defined by

$$\begin{aligned} \ell_N(x) = \frac{1}{2} \left[x \left(\log 2 + \beta |2 - 2h| + 1 \right) - (1 - m_+(N) + x) \log(1 - m_+(N) + x) \right. \\ \left. + (1 - m_+(N)) \log(1 - m_+(N)) \right]. \end{aligned} \tag{4.37}$$

Proof. Recall that $\{\sigma(t)\}_{t \geq 0}$ is the Markov chain with transition probabilities (1.5) and, for $\sigma \in \mathcal{S}_N$ with $m(\sigma) < m_+(N)$, let

$$A_N(\sigma) = \left\{ (\sigma(0), \sigma(1), \sigma(2), \dots) : \sigma(0) = \sigma, \forall i \in \mathbb{N}, \sigma(i) \in \mathcal{S}_N, \sigma(i) \sim \sigma(i+1), \right. \\ \left. \exists k \in \mathbb{N} \text{ s.t. } \sigma(k) \in \mathcal{S}_N[m_+(N)], \text{ and } \forall i \leq k-1, m(\sigma(i+1)) = m(\sigma(i)) + \frac{2}{N} \right\} \quad (4.38)$$

be the set of infinite paths starting in σ and having increasing magnetisation until the set $\mathcal{S}_N[m_+(N)]$ is reached.

Notice that, for fixed σ and N , the number k of steps of increasing magnetisation to reach $\mathcal{S}_N[m_+(N)]$ is fixed, namely $k = \frac{N}{2}(m_+(N) - m(\sigma))$.

We want to partition $A_N(\sigma)$ according to the values of the first $k+1$ elements of its paths. Given a sequence $\pi \in \mathcal{S}_N^{k+1}$, let us denote by $\{\pi\}$ the set of all paths in $A_N(\sigma)$ in which the first $k+1$ elements are exactly given by π , namely

$$\{\pi\} = \left\{ (\sigma(0), \sigma(1), \dots, \sigma(k), \sigma(k+1), \dots) \in A_N(\sigma) : (\sigma(0), \dots, \sigma(k)) = \pi \right\}. \quad (4.39)$$

Notice that, by definition of $A_N(\sigma)$, $\{\pi\}$ is empty for many $\pi \in \mathcal{S}_N^{k+1}$. We denote by $B_N(\sigma)$ the set of all the sequences $\pi \in \mathcal{S}_N^{k+1}$ such that $\{\pi\}$ is not empty. Thus, we obtain the following partition of $A_N(\sigma)$

$$A_N(\sigma) = \bigsqcup_{\pi \in B_N(\sigma)} \{\pi\}. \quad (4.40)$$

Fix $\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]$, then one simply notices that

$$\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_+(N)]} < \tau_{\mathcal{S}_N[m_{\varepsilon_N}]} \right) \geq \mathbb{P}_\sigma(A_N(\sigma)) = \sum_{\pi \in B_N(\sigma)} \mathbb{P}_\sigma(\{\pi\}). \quad (4.41)$$

Thus, we first find a lower bound on $\mathbb{P}_\sigma(\{\pi\})$ independent of π in $B_N(\sigma)$ and later we compute the cardinality of $B_N(\sigma)$. Fix $\pi = (\sigma(0), \sigma(1), \sigma(2), \dots, \sigma(k)) \in B_N(\sigma)$, then we have

$$\mathbb{P}_\sigma(\{\pi\}) = \prod_{i=1}^k p_N(\sigma(i-1), \sigma(i)) = \frac{1}{N^k} \prod_{i=1}^k \exp \left(-\beta [H(\sigma(i)) - H(\sigma(i-1))]_+ \right) \\ \geq \frac{C^k}{N^k} \prod_{i=1}^k \exp \left(-\beta N [E(m_i) - E(m_{i-1})]_+ \right) = \frac{C^k}{N^k} \prod_{i=1}^k \exp \left(-\beta \left[-2m_{i-1} - \frac{2}{N} - 2h \right]_+ \right), \quad (4.42)$$

where $m_i = m(\sigma(i))$, $C = \exp(-\beta|2 - 2h|)$ and we used the following fact

$$\frac{\exp \left(-\beta [H(\sigma(i)) - H(\sigma(i-1))]_+ \right)}{\exp \left(-\beta N [E(m_i) - E(m_{i-1})]_+ \right)} = \frac{\exp \left(-\beta [H(\sigma(i)) - H(\sigma(i-1))]_+ \right)}{\exp \left(-\beta \left[-2m_{i-1} - \frac{2}{N} - 2h \right]_+ \right)} \\ \geq \exp \left(-\beta [H(\sigma(i)) - H(\sigma(i-1))]_+ \right) = \exp \left(-\beta \left[-\frac{2}{N} \sum_{j:j \neq r} J_{jr} \sigma(i-1)_j - 2h \right]_+ \right) \\ \geq \exp \left(-\beta \left[2 - 2h - \frac{2}{N} \right]_+ \right) \geq \exp \left(-\beta |2 - 2h| \right), \quad (4.43)$$

where r is the index of the spin to be flipped to go from $\sigma(i-1)$ to $\sigma(i)$. Therefore, recalling that $m_i \in [m_{\varepsilon_N}, m_+(N)]$, we obtain the following lower bound independent of π

$$\mathbb{P}_\sigma(\{\pi\}) \geq \frac{C^k}{N^k} \prod_{i=1}^k \exp \left(-\beta \left[-2m_{\varepsilon_N} - \frac{2}{N} - 2h \right]_+ \right) = \frac{C^k}{N^k}. \quad (4.44)$$

Indeed, for ε_N sufficiently small, m_{ε_N} is close to $m_+(N) > 0$, allowing us to assume $m_{\varepsilon_N} > 0$. Therefore, $-2m_{\varepsilon_N} - \frac{2}{N} - 2h < 0$, which implies the last equality in (4.44).

We are left to compute the cardinality of $B_N(\sigma)$, with $\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]$, namely we have to count all paths from σ to $\mathcal{S}_N[m_+(N)]$ with increasing magnetisation and length $k + 1$. Any of these paths is characterised by a final spin $\bar{\sigma} \in \mathcal{S}_N[m_+(N)]$ and a sequence of negative spins which are flipped. Notice that $\bar{\sigma}$ is reachable by σ through a path with increasing magnetisation if and only if the two following properties are satisfied: $\bar{\sigma}$ has k positive spins more than σ and, for all $i \in [N]$, $\sigma_i = +1$ implies $\bar{\sigma}_i = +1$. Thus, a configuration $\bar{\sigma} \in \mathcal{S}_N[m_+(N)]$ reachable by σ through a path with increasing magnetisation is characterised by the k spins which are negative in σ and positive in $\bar{\sigma}$. Therefore, the number of reachable configurations $\bar{\sigma}$ is

$$\binom{\frac{1}{2}N(1 - m_{\varepsilon_N})}{k} = \binom{\frac{1}{2}N[1 - m_+(N) + \theta_N]}{\frac{1}{2}N\theta_N}, \tag{4.45}$$

being $\frac{1}{2}N(1 - m_{\varepsilon_N})$ the number of negative spins of $\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]$ and $k = \frac{1}{2}N\theta_N$, where θ_N has been defined in Section 4.1.

The number of paths with increasing magnetisation from $\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]$ to a reachable $\bar{\sigma} \in \mathcal{S}_N[m_+(N)]$, both fixed, is $k!$, namely the number of permutations of the k negative spins which are flipped along a path. Thus, being $k = \frac{1}{2}N\theta_N$, the cardinality of $B_N(\sigma)$ is

$$\left(\frac{1}{2}N\theta_N\right)! \binom{\frac{1}{2}N[1 - m_+(N) + \theta_N]}{\frac{1}{2}N\theta_N}. \tag{4.46}$$

Going back to (4.41), we obtain

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_+(N)]} < \tau_{\mathcal{S}_N[m_{\varepsilon_N}]} \right) &\geq \sum_{\pi \in B_N(\sigma)} \mathbb{P}_\sigma(\{\pi\}) \\ &\geq \left(\frac{C}{N}\right)^{\frac{1}{2}N\theta_N} \left(\frac{1}{2}N\theta_N\right)! \binom{\frac{1}{2}N[1 - m_+(N) + \theta_N]}{\frac{1}{2}N\theta_N} \\ &= e^{-\frac{1}{2}N\theta_N \log \frac{N}{C}} \frac{N(1 - m_+(N) + \theta_N)!}{2} \left[\frac{N(1 - m_+(N))}{2} \right]^{-1}. \end{aligned} \tag{4.47}$$

Using Stirling's approximation $n! = \sqrt{2\pi n} e^{n(\log n - 1)}(1 + o(1))$ and the notation

$$k_{\theta_N} = \frac{1 - m_+(N) + \theta_N}{1 - m_+(N)}, \tag{4.48}$$

we obtain

$$\begin{aligned} &\frac{N(1 - m_+(N) + \theta_N)!}{2} \left[\frac{N(1 - m_+(N))}{2} \right]^{-1} \\ &= \sqrt{k_{\theta_N}} \exp \left[\frac{N(1 - m_+(N))}{2} \log(k_{\theta_N}) + \frac{1}{2}N\theta_N \log \left(\frac{N(1 - m_+(N) + \theta_N)}{2} \right) - \frac{1}{2}N\theta_N \right] (1 + o(1)). \end{aligned} \tag{4.49}$$

Thus, since $k_{\theta_N} \geq 1$ and $C = \exp(-\beta|2 - 2h|)$, we conclude by

$$\begin{aligned} &\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_+(N)]} < \tau_{\mathcal{S}_N[m_{\varepsilon_N}]} \right) \\ &\geq \sqrt{k_{\theta_N}} e^{-\frac{N}{2}(\theta_N \log(\frac{N}{C}) + \theta_N - (1 - m_+(N)) \log(k_{\theta_N}) - \theta_N \log(\frac{N}{2}(1 - m_+(N) + \theta_N)))} (1 + o(1)) \\ &\geq e^{-\frac{N}{2}(\theta_N \log(\frac{N}{C}) + \theta_N - (1 - m_+(N)) \log(k_{\theta_N}) - \theta_N \log(\frac{N}{2}) - \theta_N \log(1 - m_+(N) + \theta_N))} (1 + o(1)) \\ &= e^{-\frac{N}{2}(\theta_N \log(2) + \theta_N \beta|2 - 2h| + \theta_N - (1 - m_+(N) + \theta_N) \log(1 - m_+(N) + \theta_N))} \\ &\quad \times e^{-\frac{N}{2}((1 - m_+(N)) \log(1 - m_+(N)))} (1 + o(1)) \\ &= e^{-N\ell_N(\theta_N)} (1 + o(1)). \quad \square \end{aligned} \tag{4.50}$$

To prove Lemma 4.2 we used the following fact.

Lemma 4.5. For $\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]$, for N sufficiently large and any $\gamma \in (0, 1)$,

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) \\ \leq \exp \left(-\beta N(1-\gamma) \left[f_\beta(m_-) + \delta_N - f_\beta \left(m_{\varepsilon_N} - \frac{2}{N} \right) - \varepsilon_N \right] \right). \end{aligned} \quad (4.51)$$

Proof. Let us denote by $W_N(m)$ the event of making the first flip in $\mathcal{S}_N[m]$.

For $\sigma \in \mathcal{S}_N[m_{\varepsilon_N}]$, conditioning on the first step, we obtain

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) \\ = \mathbb{P}_\sigma \left(W_N \left(m_{\varepsilon_N} + \frac{2}{N} \right) \right) \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \mid W_N \left(m_{\varepsilon_N} + \frac{2}{N} \right) \right) \\ + \mathbb{P}_\sigma \left(W_N \left(m_{\varepsilon_N} - \frac{2}{N} \right) \right) \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \mid W_N \left(m_{\varepsilon_N} - \frac{2}{N} \right) \right) \\ = \mathbb{P}_\sigma \left(W_N \left(m_{\varepsilon_N} + \frac{2}{N} \right) \right) \sum_{\sigma' \in \mathcal{S}_N \left[m_{\varepsilon_N} + \frac{2}{N} \right], \sigma \sim \sigma'} \mathbb{P}_{\sigma'} \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) \\ + \mathbb{P}_\sigma \left(W_N \left(m_{\varepsilon_N} - \frac{2}{N} \right) \right) \sum_{\sigma' \in \mathcal{S}_N \left[m_{\varepsilon_N} - \frac{2}{N} \right], \sigma \sim \sigma'} \mathbb{P}_{\sigma'} \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right). \end{aligned} \quad (4.52)$$

The first term vanishes because all the probabilities in the sum are zero. Thus, we get the upper bound

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) \\ \leq \sum_{\sigma' \in \mathcal{S}_N \left[m_{\varepsilon_N} - \frac{2}{N} \right], \sigma \sim \sigma'} \mathbb{P}_{\sigma'} \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right). \end{aligned} \quad (4.53)$$

Using first Lemma 4.6 which gives bounds uniform in σ' , we obtain

$$\begin{aligned} \mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) \\ \leq \sum_{\sigma' \in \mathcal{S}_N \left[m_{\varepsilon_N} - \frac{2}{N} \right], \sigma \sim \sigma'} \exp \left(-\beta N(1-\gamma) \left[f_\beta(m_-) + \delta_N - f_\beta(m(\sigma')) \right] \right) \\ = N \frac{1 + m_{\varepsilon_N}}{2} \exp \left(-\beta N(1-\gamma) \left[f_\beta(m_-) + \delta_N - f_\beta \left(m_{\varepsilon_N} - \frac{2}{N} \right) \right] \right) \\ = \exp \left(-\beta N(1-\gamma) \left[f_\beta(m_-) + \delta_N - f_\beta \left(m_{\varepsilon_N} - \frac{2}{N} \right) - \frac{\log N + O(1)}{\beta N(1-\gamma)} \right] \right) \\ \leq \exp \left(-\beta N(1-\gamma) \left[f_\beta(m_-) + \delta_N - f_\beta \left(m_{\varepsilon_N} - \frac{2}{N} \right) - \varepsilon_N \right] \right), \end{aligned} \quad (4.54)$$

where in the last inequality we used that, for N large enough, $\frac{\log N + O(1)}{\beta N(1-\gamma)} \leq \varepsilon_N$ (which converges to $\varepsilon > 0$, see Step 4 in Section 4.1). \square

In the proofs of Lemmas 4.2 and 4.5, we use the following fact.

Lemma 4.6. For $\sigma \in \mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$, for N sufficiently large and any $\gamma \in (0, 1)$,

$$\mathbb{P}_\sigma \left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]} \right) \leq \exp \left(-\beta N(1-\gamma) \left[f_\beta(m_-) + \delta_N - f_\beta(m(\sigma)) \right] \right). \quad (4.55)$$

Proof. For $\sigma \in \mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$,

$$\mathbb{P}_\sigma\left(\tau_{\mathcal{S}_N[m_-(N)]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]}\right) \leq \mathbb{P}_\sigma\left(\tau_{\mathcal{S}_N[m_{\delta_N}]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]}\right), \quad (4.56)$$

being $m_-(N) < m^* < m_{\delta_N} < m_{\varepsilon_N} < m_+(N)$, for N sufficiently large. Therefore, we focus on finding an upper bound on the right hand side of (4.56).

Assume that there exists a function ψ super-harmonic in $\mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$. As a consequence, $0 > L\psi(\sigma) = \frac{\partial}{\partial t} \mathbb{E}_\sigma[\psi(\sigma(t))]$. This implies $\mathbb{E}_\sigma[\psi(\sigma(t))] \leq \mathbb{E}_\sigma[\psi(\sigma(s))]$, for all $s < t$. Take $s = 0$, and $\sigma(0) = \sigma$, therefore $\mathbb{E}_\sigma[\psi(\sigma(t))] \leq \psi(\sigma)$, for all $t > 0$. Thus, $\psi(\sigma(t))$ is a super-martingale. For the integrable stopping time $T = \tau_{\mathcal{S}_N[m_{\delta_N}]} \wedge \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]}$, we use Doob's optional stopping theorem for super-martingales to show that, for all σ in the domain $\mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$ of ψ , $\mathbb{E}_\sigma[\psi(\sigma(T))] \leq \psi(\sigma)$. Therefore,

$$\psi(\sigma) \geq \mathbb{E}_\sigma[\psi(\sigma(T))] \geq \min_{\sigma' \in \mathcal{S}_N[m_{\delta_N}]} \psi(\sigma') \mathbb{P}_\sigma\left(\tau_{\mathcal{S}_N[m_{\delta_N}]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]}\right), \quad (4.57)$$

which implies that

$$\mathbb{P}_\sigma\left(\tau_{\mathcal{S}_N[m_{\delta_N}]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]}\right) \leq \frac{\psi(\sigma)}{\min_{\sigma' \in \mathcal{S}_N[m_{\delta_N}]} \psi(\sigma')}. \quad (4.58)$$

For a suitably chosen ψ the latter inequality will yield the desired upper bound. Now we are left with the choice of a suitable $\psi : \mathcal{S}_N \rightarrow \mathbb{R}$ such that $L\psi(x) < 0$, for all $x \in \mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$. We define a function ψ which depends on a parameter $\gamma \in (0, 1)$ and is constant on fixed magnetisation sets, i.e, for all $\sigma \in \mathcal{S}_N$,

$$\psi(\sigma) = \phi(m(\sigma)), \quad (4.59)$$

where $\phi : [-1, 1] \rightarrow \mathbb{R}$ is defined by

$$\phi(m) = \exp(\beta N(1 - \gamma)f_\beta(m)). \quad (4.60)$$

Our choice of ψ is similar to the one used by Bianchi, Bovier and Ioffe in [1, Proposition 6.4]. The choice of γ is relevant in (4.22).

We claim and prove later in Lemma 4.7 that ψ is super-harmonic in $\mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$. Therefore, we conclude the proof by inserting ψ in (4.58) and obtaining

$$\begin{aligned} \mathbb{P}_\sigma\left(\tau_{\mathcal{S}_N[m_{\delta_N}]} < \tau_{\mathcal{S}_N[\{m_+(N), m_{\varepsilon_N}\}]}\right) &\leq \frac{\exp(\beta N(1 - \gamma)f_\beta(m(\sigma)))}{\min_{\sigma' \in \mathcal{S}_N[m_{\delta_N}]} \exp(\beta N(1 - \gamma)f_\beta(m(\sigma')))} \\ &= \exp(\beta N(1 - \gamma)[f_\beta(m(\sigma)) - f_\beta(m_{\delta_N})]) \\ &= \exp(-\beta N(1 - \gamma)[f_\beta(m_-) + \delta_N - f_\beta(m(\sigma))]), \end{aligned} \quad (4.61)$$

where we used the definition of m_{δ_N} (see Section 4.1). □

We are now left with the proof of the super-harmonicity of ψ , which is used in the proof of Lemma 4.6.

Lemma 4.7. ψ defined in (4.59) is super-harmonic in $\mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$.

Proof. We have to prove that $L\psi(x) < 0$, for all $x \in \mathcal{S}_N[[m_{\delta_N}, m_{\varepsilon_N}]]$. Set $\tilde{m} = m(x)$. As usual, we try to rewrite the terms appearing in the expression for $L\psi(x)$ in terms of their

mean-field version.

$$\begin{aligned}
 L\psi(x) &= \sum_{y \in \mathcal{S}_N} p(x, y) [\psi(y) - \psi(x)] \\
 &= \frac{1}{N} \sum_{y \in \mathcal{S}_N} \mathbb{1}_{y \sim x} \exp(-\beta[H(y) - H(x)]_+) \\
 &\quad \times \left[\exp(\beta(1 - \gamma)Nf_\beta(m(y))) - \exp(\beta(1 - \gamma)Nf_\beta(m(x))) \right] \\
 &= \frac{1}{N} \sum_{m \in \Gamma_N} \exp(-\beta N[E(m) - E(\bar{m})]_+) \\
 &\quad \times \left[\exp(\beta N(1 - \gamma)f_\beta(m)) - \exp(\beta N(1 - \gamma)f_\beta(\bar{m})) \right] \\
 &\quad \times \sum_{y: m(y)=m} \mathbb{1}_{x \sim y} \frac{\exp(-\beta[H(y) - H(x)]_+)}{\exp(-\beta N[E(m) - E(\bar{m})]_+)} \\
 &\leq \sum_{m \in \Gamma_N} \exp(-\beta N[E(m) - E(\bar{m})]_+) \phi(\bar{m}) \left[\exp(\beta N(1 - \gamma)[f_\beta(m) - f_\beta(\bar{m})]) - 1 \right] \\
 &\quad \times e^{2\beta} \left[\frac{1 + \bar{m}}{2} \mathbb{1}_{\bar{m} - \frac{2}{N}}(m) + \frac{1 - \bar{m}}{2} \mathbb{1}_{\bar{m} + \frac{2}{N}}(m) \right],
 \end{aligned} \tag{4.62}$$

where ϕ is defined in (4.60) and we used the upper bound $\exp(2\beta)$ on $G(\sigma, m')$ as in the proof of the upper bound on capacity (see (3.8), (3.9)).

Now, recalling definition (1.20), we use the following notation

$$r_+ = \tilde{r}_N(\bar{m}, \bar{m} + \frac{2}{N}) = \exp\left(-2\beta \left[-\frac{1}{N} - (\bar{m} + h)\right]_+\right) \frac{1 - \bar{m}}{2}, \tag{4.63}$$

$$r_- = \tilde{r}_N(\bar{m}, \bar{m} - \frac{2}{N}) = \exp\left(-2\beta \left[-\frac{1}{N} + \bar{m} + h\right]_+\right) \frac{1 + \bar{m}}{2}, \tag{4.64}$$

and, for all $m \in \Gamma_N \setminus \{1\}$,

$$g(m) = \frac{N}{2} [f_\beta(m + \frac{2}{N}) - f_\beta(m)]. \tag{4.65}$$

Therefore, we can rewrite (4.62) as

$$\begin{aligned}
 L\psi(x) &\leq e^{2\beta} \phi(\bar{m}) r_+ \left[\exp(2\beta(1 - \gamma)g(\bar{m})) - 1 \right] \\
 &\quad + e^{2\beta} \phi(\bar{m}) r_- \left[\exp(-2\beta(1 - \gamma)g(\bar{m} - \frac{2}{N})) - 1 \right] \\
 &= e^{2\beta} \phi(\bar{m}) r_+ G_+,
 \end{aligned} \tag{4.66}$$

where

$$G_+ = \left(e^{2\beta(1-\gamma)g(\bar{m})} - 1 \right) + \frac{r_-}{r_+} \left(e^{-2\beta(1-\gamma)g(\bar{m} - \frac{2}{N})} - 1 \right). \tag{4.67}$$

Being $e^{2\beta}$, $\phi(\bar{m})$ and r_+ positive, we have only to show that $G_+ < 0$. First we notice that

$$g(m) = -m - h + \frac{1}{\beta} I'(m) + O\left(\frac{1}{N}\right) \tag{4.68}$$

and similarly

$$g\left(m - \frac{2}{N}\right) = -m - h + \frac{1}{\beta} I'(m) + O\left(\frac{1}{N}\right). \tag{4.69}$$

Therefore,

$$g(m) - g\left(m - \frac{2}{N}\right) = O\left(\frac{1}{N}\right). \tag{4.70}$$

Then, since $I'(m) = \frac{1}{2} \log\left(\frac{1+m}{1-m}\right)$ (see (1.16)), and using (4.69) we have

$$\begin{aligned} \frac{r_-}{r_+} &= \frac{1 + \bar{m} \exp\left(2\beta\left[-\frac{1}{N} - (\bar{m} + h)\right]_+\right)}{1 - \bar{m} \exp\left(2\beta\left[-\frac{1}{N} + \bar{m} + h\right]_+\right)} \\ &= \frac{1 + \bar{m}}{1 - \bar{m}} \exp\left(-2\beta(\bar{m} + h)\right) \left(1 + O\left(\frac{1}{N}\right)\right) \\ &= \exp\left(2I'(\bar{m}) - 2\beta(\bar{m} + h)\right) \left(1 + O\left(\frac{1}{N}\right)\right) \\ &= \exp\left(2\beta\left[g\left(\bar{m} - \frac{2}{N}\right) + \bar{m} + h + O\left(\frac{1}{N}\right)\right] - 2\beta(\bar{m} + h)\right) \left(1 + O\left(\frac{1}{N}\right)\right) \\ &= \exp\left(2\beta g\left(\bar{m} - \frac{2}{N}\right)\right) \left(1 + O\left(\frac{1}{N}\right)\right). \end{aligned} \tag{4.71}$$

Therefore, rearranging (4.67) and then using (4.70) and (4.71), we obtain

$$\begin{aligned} G_+ &= \left[\exp\left(2\beta(1 - \gamma)g(\bar{m})\right) - 1 \right] \left[1 - \frac{r_-}{r_+} \exp\left(-2\beta(1 - \gamma)g\left(\bar{m} - \frac{2}{N}\right)\right) \right] \\ &\quad + \frac{r_-}{r_+} \left[\exp\left(2\beta(1 - \gamma)\left[g(\bar{m}) - g\left(\bar{m} - \frac{2}{N}\right)\right]\right) - 1 \right] \\ &= \left[\exp\left(2\beta(1 - \gamma)g(\bar{m})\right) - 1 \right] \left[1 - \exp\left(2\beta\gamma g\left(\bar{m} - \frac{2}{N}\right)\right) \left(1 + O\left(\frac{1}{N}\right)\right) \right] \\ &\quad + \frac{r_-}{r_+} \left[\exp\left(2\beta(1 - \gamma)O\left(\frac{1}{N}\right)\right) - 1 \right]. \end{aligned} \tag{4.72}$$

Notice that, for every $m \in [m_{\delta_N}, m_{\varepsilon_N}] \subset [m^*, m_+)$, $g(m)$ is negative, being f_β strictly decreasing in $[m^*, m_+)$. As a consequence, $e^{2\beta(1-\gamma)g(\bar{m})} - 1 < 0$. Furthermore, for N sufficiently large, $1 - e^{2\beta\gamma g(\bar{m} - \frac{2}{N})} \left(1 + O\left(\frac{1}{N}\right)\right) > 0$, implying that the first term in (4.72) is negative.

Moreover, $\frac{r_-}{r_+} \geq 0$ is uniformly bounded from above, for N sufficiently large. Therefore, since β is finite, $\gamma \in (0, 1)$ and the term $\left[\exp\left(2\beta(1 - \gamma)O\left(\frac{1}{N}\right)\right) - 1\right]$ is positive but converging to zero as N grows to infinity, the second term in (4.72) is negligible.

Therefore, for N sufficiently large, G_+ is negative, concluding the proof. \square

4.4 Lower bound on the harmonic sum

In this section we provide the main ideas to prove the second part of Theorem 1.7, namely the lower bound on the harmonic sum in (4.1).

Proof of Theorem 1.7. Lower bound. The proof is very similar to the proof of the upper bound we gave in Section 4.2, therefore we omit the details. The main contribution is given once again by the sum on $\mathcal{S}_N[U_{\delta, N}(m_-)]$.

We have,

$$\begin{aligned} \sum_{\sigma \in \mathcal{S}_N} \mu_{\beta, N}(\sigma) h_{m_-, m_+}^N(\sigma) &\geq \sum_{\sigma \in \mathcal{S}_N[U_{\delta, N}(m_-)]} \mu_{\beta, N}(\sigma) h_{m_-, m_+}^N(\sigma) \\ &= \sum_{\sigma \in \mathcal{S}_N[U_{\delta, N}(m_-)]} \mu_{\beta, N}(\sigma) - \sum_{\sigma \in \mathcal{S}_N[U_{\delta, N}(m_-)]} \mu_{\beta, N}(\sigma) (1 - h_{m_-, m_+}^N(\sigma)) \\ &\geq \sum_{m \in U_{\delta, N}(m_-) \setminus [-1, m_-(N)]} \mathcal{Q}_{\beta, N}(m) \\ &\quad - \sum_{\sigma \in \mathcal{S}_N[U_{\delta, N}(m_-) \setminus [-1, m_-(N)]]} \mu_{\beta, N}(\sigma) \mathbb{P}_\sigma\left(\tau_{\mathcal{S}_N[m_+(N)]} < \tau_{\mathcal{S}_N[m_-(N)]}\right). \end{aligned} \tag{4.73}$$

The first term, i.e. the sum on the mesoscopic measure $\mathcal{Q}_{\beta,N}$, gives the main contribution. This sum can be estimated from below using the lower bound in Corollary 2.5, obtaining a lower bound similar to the second upper bound in Corollary 2.7 and applying the saddle point method as in (4.11). More precisely, using notation (1.38), we have the following lower bound for $s > 0$:

$$\sum_{m \in U_{\delta,N}(m_-) \setminus [-1, m_-(N)]} \mathcal{Q}_{\beta,N}(m) \stackrel{P(s)}{\geq} \frac{e^{\kappa-s} \exp(-\beta N f_{\beta}(m_-))}{Z_{\beta,N} \sqrt{(1-m_-^2) \beta f_{\beta}''(m_-)}} (1 + o(1)). \quad (4.74)$$

The second term in (4.73), appearing with a negative sign in front, is estimated via an upper bound, obtaining

$$\begin{aligned} \sum_{\sigma \in \mathcal{S}_N[U_{\delta,N}(m_-) \setminus [-1, m_-(N)]]} \mu_{\beta,N}(\sigma) \mathbb{P}_{\sigma} \left(\tau_{\mathcal{S}_N[m_+(N)]} < \tau_{\mathcal{S}_N[m_-(N)]} \right) \\ \leq \frac{e^{s+\alpha} \exp(-\beta N f_{\beta}(m_-))}{Z_{\beta,N}} \sqrt{\frac{2}{\pi(1-m_+^2)}} e^{-\beta N c} (1 + o(1)), \end{aligned} \quad (4.75)$$

which is negligible compared to the right hand side of (4.74), concluding the proof.

We omit the proof of (4.75) being it again technical and very similar to the proof of the upper bound (4.26) in Part 4 of Section 4.2. An analogue construction to the one given in Section 4.1 and similar proofs to those in Section 4.3 are needed. The main difference consists in restricting the analysis on a right neighbourhood of $m_-(N)$ instead of a left neighbourhood of $m_+(N)$. □

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