How to guess the inter magnetic bubble potential by using a simple perceptron?

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**Abstract.** – It is shown that magnetic bubble films behaviour can be described by using a 2 D super-Ising hamiltonian. Calculated hysteresis curves and magnetic domain patterns are successfully compared with experimental results taken in literature. The reciprocal problem of finding parameters of the super-Ising model to reproduce computed or experimental magnetic domain pictures is solved by using a perceptron neural network.

Thin films with perpendicular magnetization show widespread technological applications, particularly as high density storage devices. Thus, our understanding, at a fundamental level, of the way the magnetic domains are organized in such films is extremely important. This letter is focussed on films having cylindrical shape bubble and labyrinthine magnetic domains. These films have been extensively studied in the seventies as they were considered as a possible support for high density magnetic recording, essentially through the possibility of manipulating bubbles [1]. There is nowadays a renewed interest for bubble and stripe domains [2, 3], because of the appearance of new techniques, like Magnetic Force Microscopy (MFM) or Secondary Electron Microscopy with Polarization Analysis (SEMPA), showing these domain geometries at the nanometer scale. The theoretical models used up to now to describe these domains were developed 30 years ago [4–6], and consist of a comparison of the energies of domains having different ideal geometries. Such an approach is the simplest one can imagine but provides a limited understanding of the mechanism of domain formation. More elaborated descriptions based on linearized micromagnetic equations [7] or Ginzburg-Landau formalism [8] have been proposed. These approaches describe qualitatively the transition from bubble to stripe structure, but they do not allow quantitative comparison with experimental results.

It will be explained in a first part that bubble films behave as an Ising system with long-range interaction (super-Ising) between bubbles considered as giant spins. The influence of the interaction potential parameters (range and shape) on the domain geometry and on the magnetization curve will be discussed. The comparison between simulations and experimental results will help to determine the relevant parameters. In a second part, we solve the inverse problem i.e the determination of the interaction potential from a simulated or experimental magnetic map. The treatment proposed is a perceptron type neural network (NNT) searching for correct values of the potential using an error-correction learning rule [13]. To our knowledge, it is the first time that neural network techniques are proposed to analyse magnetic
domain pattern geometries.

The magnetic behaviour of bubble films can be understood starting from a magnetic mono-domain situation. This state becomes unstable if an initially large perpendicular magnetic field is decreased to reach a critical value \( B_N \) (nucleation field). In fact, for such a field, a situation with lower energy is obtained if a bubble with a diameter \( d_0 \) is formed (critical bubble). The competition between the wall and magnetostatic energies determines \( d_0 \). The magnetostatic energy favours the formation of domains and varies, in first approximation, with the volume \((V_0 \propto d_0^3)\) of the bubble. The wall energy, that does not favor the presence of domains, is proportional to the bubble's surface \( (S_0 \propto d_0)\). The wall thickness is assumed to be negligible \([1–6]\). The characteristic length \( l = \sigma_w / 4\pi M_s^2 \) reflects the balance between the wall energy and the magnetostatic energy, \( \sigma_w \) is the wall energy density and \( M_s \) is the saturation magnetization of the material. Typically, for cobalt or iron-palladium alloy, \( l \) is around 10 nm. This length allows us to treat a non-dimensional problem by choosing \( l, 4\pi M_s \) and \( 4\pi M_s^2 h d_0^2 \) as respectively length, field and energy units. The function \( d_0(h) \) has been tabulated \([1, 5]\) and can be approached by \( d_0 = 0.15h + 3.86 \) for \( h \leq 50 \); the tilded quantities are non-dimensional variables.

We have considered up to now an isolated bubble. The particular case of two \( \text{down} \) magnetized critical bubbles in an \( \text{up} \) magnetized film (see fig. 1) is the following. Either the two bubbles are not in contact and they only interact via the dipolar field, the wall energy is then \( 2 \times \sigma_w S_0 \), or the bubbles are in contact and the wall energy becomes \( \sigma_w (2 \times S_0 - S_1), S_1 \) being the contact surface between the two bubbles. These two cases can be summarized in a single model on a square lattice in which each cell \( i \), of size \( d_0 \times d_0 \) is given the value of \( \eta_i \), corresponding to \( \text{up} \) and \( \text{down} \) magnetization. Then the interaction energy between two cells reads:

\[
\tilde{E}_{i,j}^{\text{int}} = \frac{1 - \eta_i \eta_j}{2d_0} \delta_{j,i\pm 1} + |\tilde{E}_{\text{dip}}(i - j)| \eta_i \eta_j
\]

where the first term is the wall energy and the second term the dipole energy. The Hamiltonian fully describing the interaction between cells is given by:

\[
\mathcal{H} = \frac{1}{2} \sum_{i,j} \tilde{V}_{i,j} \eta_i \eta_j - \sum_i \eta_i B
\]

with \( \tilde{V}_{i,j} = \delta_{j,i\pm 1}/2d_0 - |\tilde{E}_{\text{dip}}(i - j)| \) if \( i \neq j \) and \( \tilde{V}_{i,i} = 0 \). The second term accounts for the Zeeman coupling between each bubble's momentum and the applied magnetic field \( B \). The Hamiltonian \( \mathcal{H} \) is formally identical to a super-Ising one. Similar analysis, using the competition between wall and dipolar energies, were previously presented in \([9, 10]\) but it is the first time that bubble magnetic films are shown as \( 2 \text{D} \) super-Ising systems. It can be noticed that a similar Hamiltonian could be written to describe perpendicularly magnetized nanoscale dot arrays \([11]\).

If the dipolar field of a square cell is approximated by the one of a cylinder of radius \( R_0 = d_0/2 \), \( \tilde{E}_{\text{dip}} \) reads:

\[
\tilde{E}_{\text{dip}}(|i - j|) = \frac{\pi R_0}{2h} \int_0^{+\infty} d\epsilon \left( \frac{J_1(\epsilon)}{\epsilon} \right)^2 J_0(\epsilon |i - j|) \left( 1 - e^{-\epsilon h/R_0} \right)
\]

where \( J_0 \) and \( J_1 \) are Bessel functions of order respectively 0 and 1. The dipole term is short ranged even if its range increases with the film thickness (see fig. 1). For simplification of the
forthcoming discussion, the potential $\tilde{V}$ is approximated by:

$$\tilde{V}(|i-j|) = \begin{cases} 
\tilde{V}_+ & \text{if } 0 < |i-j| \leq \tilde{R}_+ \\
\tilde{V}_- & \text{if } \tilde{R}_+ < |i-j| \leq \tilde{R}_- 
\end{cases}$$

(4)

where $V_+$ is the sum of the dipolar and the wall energy between first neighbours, $V_-$ is the dipolar energy, $\tilde{R}_+$ the distance to the first neighbours and $\tilde{R}_-$ is the range of the potential. For the cobalt films studied in [2], with thickness $\tilde{h} = 50$ and $\tilde{d}_0 \approx 10$, $\tilde{V}_+$ and $\tilde{V}_-$ can be respectively fixed to 0.02 and -0.02, $\tilde{R}_- = 4$ and $\tilde{R}_+ = \sqrt{2}$. This choice of $\tilde{R}_+$ is equivalent to consider also the second neighbours of a cell as nearest neighbours. This allows the model to be more isotropic and attenuates the lattice effects.

With $\Delta_i = \sum_k \tilde{V}(k) \eta_{i+k} + \tilde{B}$, the average momentum of a cell $i$, for a given environment $\text{env}_i$, is in the canonical ensemble:

$$\langle \eta_i \rangle_{\text{env}_i} = \tanh \left( \frac{4\pi M_s^2 d_0^2 h_k B}{k_B T} \Delta_i \right)$$

(5)

Since the ratio $V_+/k_B \approx 10^5 K$, in a reasonable $T$ range, one can let $T$ go to 0+ in (5) and let the lattice follow the rule:

$$\eta_i = \begin{cases} 
\text{sign} (\Delta_i) & \text{if } \Delta_i \neq 0 \\
\text{random choice} & \text{if } \Delta_i = 0 
\end{cases}$$

(6)

The case $\Delta_i = 0$ corresponds to $\eta_i$ equals $\pm 1$ being equiprobable. This is crucial since decreasing $B$ from the monodomain state (all $\eta_i = +1$), the magnetization starts to flip for $B_N = -\sum_k \tilde{V}(k)$, i.e. $\Delta_i = 0$. Without dipolar field, the magnetization sharply flips for $B_N < 0$ with no domain nucleation (in this case $B_N$ is similar to a coercitive field). With dipolar field, $B_N$ is possibly positive and, in this case, the magnetization begins to decrease

Fig. 1 – Interaction dipolar energy between two bubbles of same sign as a function of their separation distance $R$ for different film thicknesses. The inset shows two down cyllindar shape bubbles in an up magnetized film.
Fig. 2 – Simulated magnetization curve. The computed patterns (100 × 100 cells) show the domain evolution at different points of the hysteresis curve when the field is decreased, starting from the saturated up state. Black and white parts correspond to up and down domains.

Fig. 3 – Computed patterns in the demagnetized state for different potential values, $R_+ = \sqrt{2}$ and $R_- = 4$. Black and white parts correspond respectively to up and down domains.

by creating domains: $B_N$ is the nucleation field.

The hysteresis curve shown in fig. 2 is calculated for the parameters defined above. The similarity with the experimental results from [2] is striking. Our computed value of $B_N \simeq 10$ kOe ($M_s = 1480 \text{ erg.cm}^{-3}$ for cobalt) is in good agreement with regard to the experimental value ($\simeq 12$ kOe). The behaviour described by [4–6] is found in our simulations. When the field is decreased starting from the saturated up state (see fig. 2): down bubbles first appear, decreasing the field further, bubbles are elliptically deformed (stripe out), then coalesce to form a maze structure, the domain walls move progressively making up domain size decreases, and finally up bubbles appear. We may emphasize that the whole evolution is completely described using only the hamiltonian (2) with no further hypothesis.

Domains in a demagnetized state for different potential values are shown in fig. 3. The $V_+$ part of the potential is arbitrarily fixed to unity since the calculation is performed under zero field. The demagnetized state is obtained starting from a full random initialization. This corresponds physically to demagnetize the film through annealing it at an infinite temperature, and to cool it down abruptly. We have tested that it is not necessary to use a complex procedure of simulated annealing to reach the ground state of the system. Our domain patterns in fig. 3 can be compared to the MFM images obtained by V. Gehanno et al. [3], for different film thicknesses. The domain geometry evolves from a maze type structure like our fig. 3a to a fig. 3c type domain structure by decreasing the film thickness from 50 nm to 5 nm, i.e. by decreasing the $V_-$ dipolar part of the potential. The pattern in fig. 3d fits well to SEMPA images obtained by R. Allenspach et al. [12] for ultra-thin cobalt films ($\sim 0.8$ nm) on the (111) gold surface. Since $d_0 \approx 50$ nm, the image size in fig. 3a–d is around $10 \mu\text{m} \times 10 \mu\text{m}$,
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\[ \delta_i^{(1)} \delta_i^{(2)} \delta_i^{(3)} \]

\[ V = W \]

\[ \bar{\zeta}^{(i)} \]

\[ \bar{W} \]

Fig. 4 – Schematic representation of the first, second and third layers of the \( i \) cell neighbours. The potential \( \bar{V} \) is constant on each layer.

Fig. 5 – \( C_+ \) and \( C_- \) points are marked by respectively + and -. If a hyperplane \( (P) \) separating \( C_+ \) from \( C_- \) points is found, a vector \( W \) perpendicular to \( (P) \) verifies that the \( \zeta^{(i)} \) projection on \( W \) is positive or negative depending \( \zeta^{(i)} \) belongs respectively to \( C_+ \) or \( C_- \) classes.

which gives a domain size: 2-5 \( \mu m \) for fig. 3d to 50-100 nm for fig. 3a-c, in good agreement with experimental data [3, 12].

We now solve the inverse problem: can we from a given calculated or experimental domain pattern, infer the unknown potential \( V \)? One needs to find the relation between the state of the \( i \) cell and \( env_i \). The \( i \) cell’s environment can be described with the vector \( \bar{\zeta}^{(i)} \) whose \( q \) components are \( \zeta_q^{(i)} = \sum_{p \in \delta_q(i)} \eta_p \), where the \( \delta_q(i) \) set states for the \( i \) cell’s \( q \) layer of neighbours (see fig. 4). We also define a vector \( \bar{W} \) with components \( W_q \) equal to \( \tilde{V}(|i-j|) \), the \( j \) cell belonging to \( \delta_q(i) \). We separate the \( \bar{\zeta}^{(i)} \) vectors into two classes \( C_+ \) and \( C_- \) depending on \( \eta_i \) respectively equals +1 or -1. If one excepts the marginal case where \( \Delta_i = 0 \) (which is only true for \( B = B_N \)), the rule (6) reads:

\[
\begin{align*}
\bar{\zeta}^{(i)} \in C_+ & \iff \langle \bar{\zeta}^{(i)}, \bar{W} \rangle + \bar{B} > 0 \\
\bar{\zeta}^{(i)} \in C_- & \iff \langle \bar{\zeta}^{(i)}, \bar{W} \rangle + \bar{B} < 0
\end{align*}
\]  

where \( \langle \bar{\zeta}^{(i)}, \bar{W} \rangle = \sum_q \bar{\zeta}_q^{(i)} W_q = \sum_k \bar{V}_k \eta_{i+k} \) is the euclidian scalar product. In other words, one has to find a hyperplane \( (P) \) separating \( C_+ \) points from \( C_- \) points, thus a \( \bar{W} \) vector perpendicular to \( (P) \) will be solution of eq. (7) (the case \( \bar{B} = 0 \) is illustrated on fig. 5).

We start with a randomly chosen vector \( \bar{W}^{(0)} \). For a given \( i \) cell, the \( \bar{\zeta}^{(i)} \) vector is calculated. If \( \langle \bar{W}^{(0)}, \bar{\zeta}^{(i)} \rangle + \bar{B} \) and \( \eta_i \) have the same sign, another cell is tested. In the other case, \( \bar{W}^{(0)} \) is modified according to the rule

\[
\bar{W}^{(1)} = \bar{W}^{(0)} + \gamma \left[ \eta_i - \text{sign} \left( \langle \bar{W}^{(0)}, \bar{\zeta}^{(i)} \rangle + \bar{B} \right) \right] \bar{\zeta}^{(i)}
\]

and so on until a vector \( \bar{W}^{(n)} \) is found, verifying \( \eta_i = \text{sign} \left( \langle \bar{W}^{(n)}, \bar{\zeta}^{(i)} \rangle + \bar{B} \right) \) for every \( i \).
Finaly, the $\vec{W}^{(n)}$ components give the $\vec{V}$ potential we are looking for.

In fact, this algorithm follows closely the Rosenblatt’s perceptron NNT [13] in its simplest expression since only one neuron is needed. The $N$ components $\zeta_{q}^{(i)}$ are injected in the $N$ cells input layer of the perceptron. The neuron answer $\sigma$ is $\text{sign} \left( \left\langle \vec{W}^{(m)}, \vec{\zeta}^{(i)} \right\rangle + \vec{B} \right)$, the $W_{q}^{(m)}$ being the synaptic coupling strengths between the $q$ cell of the input layer and the neuron. If the perceptron makes a mistake ($\eta_{i} \neq \sigma$), the synaptic couplings are corrected following eq. 8, this equation defining the error-correction learning rule of the perceptron. Papert and Minsky [14] have shown that the perceptron NNT converges in a finite number of steps if $\gamma > 0$ and if the classification problem of the $\vec{\zeta}^{(i)}$ vectors between the two classes $C_{+}$ and $C_{-}$ is a linearly separable one [15]. In our case, the separability condition is verified by construction and for $\gamma = 0.5$, a correct potential is found after only one scan of a $250 \times 250$ cells domain pattern.

The guessed potentials $V/V(1)$ from demagnetized domain structures of fig. 3 are shown in fig. 6. The shape of the potential (positive between first neighbours and negative for other neighbours) as well as $R_{-}$ and $R_{+}$ are estimated correctly. The determination of $V_{-}/V_{+}$ is less accurate because of statistical fluctuations. These latter have been reduced here by taking the mean value of the curve between neighbour points (dashed line). We have only used demagnetized states because it is often the only information that can be experimentally obtained, but one can use domain patterns corresponding to different values of $B$ as an input and statistical fluctuations will be strongly reduced. Nevertheless, the above results show that even using a single image, and such a simple perceptron algorithm, it is possible to obtain probing results.

In summary, magnetic bubble films can be described using a 2D super-Ising hamiltonian. The potential allowing to fit an experimental or calculated domain pattern can be guessed by a perceptron neural network. The universality of the super-Ising model and of the bubble and
maze domains [16] suggests this approach could be exploited in numerous other problems as
demixion processes [17], Turing mechanisms [18] or surface gases [19] to analyse experimental
images.

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