

Nonlinear simulations to optimize magnetic nanoparticle hyperthermia

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Magnetic nanoparticle hyperthermia is an attractive emerging cancer treatment, but the acting microscopic energy deposition mechanisms are not well understood and optimization suffers. We describe several approximate forms for the characteristic time of Néel rotations with varying properties and external influences. We then present stochastic simulations that show agreement between the approximate expressions and the micromagnetic model. The simulations show nonlinear imaginary responses and associated relaxational hysteresis due to the field and frequency dependencies of the magnetization. This suggests that efficient heating is possible by matching fields to particles instead of resorting to maximizing the power of the applied magnetic fields. © 2014 AIP Publishing LLC. [http://dx.doi.org/10.1063/1.4867987]

Magnetic nanoparticle (MNP) hyperthermia is considered a potentially useful addition to current cancer treatment modalities¹ yet consensus has not been reached as to the precise mechanism of nanoparticle heating.² Simple models including linear response³ and Stoner-Wohlfarth hysteresis⁴ have been used to predict hyperthermia performance from various MNPs. Both models are approximations requiring small applied fields and equilibrium, respectively, but occasionally have been applied beyond their valid range to predict optimal heating parameters. This is confusing to readers not familiar with the theory, and data are often in conflict with the theoretical predictions.^{3,5}

In response to this, we demonstrate a more general approach to modeling MNP heating using nonlinear stochastic differential equations. We examine the phenomena of field dependent characteristic timescales, relaxational hysteresis curves, and nonlinear imaginary magnetization responses with the hope of informing decisions to optimize hyperthermia.

A Néel rotation model for MNP hyperthermia is justified because single-domain ferromagnetic particles experimentally display the best heating properties to date.⁵ Biological targeting schemes may also direct decisions for particle sizes, shapes, or surface construction,⁶ but these choices are beyond the scope of this paper. An upper bound on the field's power could be the limit where hyperthermia's benefit of specific cytotoxicity is overwhelmed by eddy-current damage of healthy tissue. In experiment,⁷ patients could tolerate fields with a product of field amplitude and frequency below 10⁴ T/s, so we do not exceed this value.

The time dynamics of MNPs can be calculated using the micromagnetic stochastic differential equation of Landau, Lifshitz, and Gilbert (LLG). The "LLG" equation derives from the Larmor precession of a spin in a magnetic field appended by a phenomenological velocity-dependent damping term.⁸ It is written in terms of the normalized magnetic moment direction \mathbf{m} of each nanoparticle, and the effective field \mathbf{H} that can be defined as a partial derivative of the free

energy with respect to the magnetic moment. The LLG equation is then for each particle (denoted with subscript i)

$$\frac{\mathrm{d}\mathbf{m}_i}{\mathrm{d}t} = \frac{\gamma}{1+\alpha^2} [\mathbf{H} \times \mathbf{m}_i + \alpha \mathbf{m}_i \times (\mathbf{H} \times \mathbf{m}_i)], \qquad (1)$$

with the electron gyromagnetic ratio $\gamma = 1.76 \cdot 10^{11} \text{Hz/T}$ and a dimensionless magnetic damping parameter α . We include additional physics including anisotropy, dipoledipole interactions, and thermal fluctuations by modifying the free energy and, thus, the effective field. Now

$$\mathbf{H} = \underbrace{H_o \hat{z} \cos \omega t}_{\text{applied}} + \underbrace{\frac{2\hat{n}E_k}{\mu} \mathbf{m}_i \cdot \hat{n}_i}_{\text{anisotropy}} + \underbrace{\frac{\mu \mu_0}{4\pi d^3} \langle \mathbf{m} \rangle}_{\text{dipole}} + \underbrace{\mathbf{h}(t)}_{\text{stochastic}}, \quad (2)$$

where the externally applied field has amplitude H_o and frequency $\omega = 2\pi f$. We assume a single anisotropy axis \hat{n}_i in a random direction for each particle arising from shape and crystallinity effects; this creates two energy minima, and requires a rotation of the particle to overcome the anisotropy energy barrier $E_K = K_a V_c$ with anisotropy constant K_a and core volume V_c assumed the same for all particles (a size distribution will change the dynamics, but presently, we avoid this complication). The magnitude of the moment is $\mu = M_s V_c$ with saturation magnetization M_s . A mean dipole field is included that depends on the average magnetization of all the other particles; its strength is determined from the magnetic moment, the permeability of free space μ_0 , and an average particle spacing $d = c_N^{-3}$ that is computed from the particle concentration c_N .

Thermal fluctuations of the field are included with a stochastic field with zero mean and unit standard deviation

$$\langle \mathbf{h}(t) \rangle = 0, \quad \langle \mathbf{h}_j(t) \mathbf{h}_k(t') \rangle = \frac{2kT\alpha}{\mu\gamma} \delta_{jk} \delta(t-t'), \quad (3)$$

where the Dirac delta function implies the noise field is white and is spatially correlated by the Kronecker delta where the indices imply the direction, e.g., $j, k \in x, y, z$.

Before solving numerically, it is possible to glean some insight with analytical approximations. The timescale of a

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thermal rotation over the anisotropy barrier is referred to as the Néel relaxation time⁹

$$\tau_{\rm N} = \frac{\tau_0}{2} \sqrt{\frac{\pi}{\sigma^3}} e^{\sigma} \quad \text{with} \quad \tau_0 = \frac{\mu}{2\gamma kT} \frac{(1+\alpha^2)}{\alpha}, \qquad (4)$$

controlled by the ratio of anisotropic to thermal energy $\sigma = E_K/kT$ with Boltzmann's constant k and temperature T. At room temperature and typical saturation magnetizations, the magnitude of τ_0 is on the order of the usual quoted value of 10^{-10} s.

A sample of MNPs will attempt to align to a magnetic field. If a stronger field is imposed, the particles will align faster.¹⁰ Only equilibrium fluctuations are considered in the Néel time quoted above, so it does not describe this phenomenon. Brown wrote a field-inclusive characteristic time from a high barrier approximation ($\sigma \gg 1$) to the Fokker–Planck equation describing the thermally assisted movements between anisotropic and magnetic field energy minima¹¹

$$\tau_{\rm hi} = \left(\frac{\tau_{\rm N}}{1 - \epsilon^2}\right) \left(\frac{e^{\sigma \epsilon^2}}{\cosh \xi + \epsilon \sinh \xi}\right). \tag{5}$$

The applied magnetic field is accounted for with the unitless $\xi = \mu H/kT$. We define the ratio of anisotropic to magnetic energy as $\epsilon = \xi/2\sigma$, and thus, when $\xi = 0$, the expression reduces to the equilibrium Néel time.

From the LLG equation, it is possible to approximate the *mean* magnetization if the correlation functions between variables are approximately zero. This is physically equivalent to requiring high amplitude fields or low temperatures, so that the stochastic term is negligible. We are only interested in the average magnetization in the direction parallel to a constant applied field M_z , so, we set $\mathbf{H} = \hat{z}H_z$, and the LLG equation simplifies to

$$\frac{\mathrm{d}M_z}{\mathrm{d}t} = -\frac{\gamma \alpha H_z}{\left(1 + \alpha^2\right)} \left(1 - M_z^2\right),\tag{6}$$

where we have used the fact that the magnetization magnitude is conserved ($\mathbf{m}^2 = 1$). Integrating both sides assuming the initial magnetization and time are both zero results in

$$M_{\rm eq} = \tanh\left[-\frac{\gamma\alpha H_z}{(1+\alpha^2)}t\right] = \tanh\left[-\frac{t}{\tau_c}\right],\tag{7}$$

where we have interpreted the constant which has dimensions of frequency as an inverse of the characteristic time

$$\tau_c = \frac{(1+\alpha^2)}{\gamma \alpha H_z} = \frac{2\tau_0}{\xi}.$$
(8)

This expression corroborates our intuition that the characteristic time is shorter for higher field strengths. Setting $\alpha = 1$ gives the minimum $\partial_{\alpha} \tau_c = 0$ corresponding to the magnetization switching in a single precession time. For $\alpha \gg 1$ the rotations are overdamped, resulting in phase lagging. For $\alpha \ll 1$, the magnetizations will precess significantly. From this microscopic interpretation, α is the key to delivering heat. Unfortunately, α is not well understood in terms of fundamental nanoparticle variables. The analogous derivation with an oscillating magnetic field (with $\mathbf{H} = H_o \hat{z} \cos \omega t$) leads to

$$M_z = -\tanh\left[\frac{\xi}{2\omega\tau_0}\sin\omega t\right] = -\tanh\frac{\sin\omega t}{\omega\tau_c},\qquad(9)$$

where, now, instead of a decaying magnetization, the oscillations are parameterized by a constant that is proportional to the field strength divided by the frequency. Accordingly, saturation decreases with increasing frequency, reducing net relaxation losses.

We have described several ways to approximate the characteristic time of a nanoparticle sample. Now, to test these times and their respective ranges of validity, we resort to numerical simulations of the LLG equation using a second order Heun integration scheme in the sense of Itō.¹² The white noise field is implemented as a Wiener process with a Gaussian distribution of magnitudes. Thus statistical moments of the sample magnetization can be developed. Unless specified, each simulation uses $i = 10^5$ repeated integrations (understood by using *i* to be the number of particles) and 2^{10} time-steps. Nanoparticles are spherical with 5 nm radii, $\alpha = 1$, $M_s = 31$ emu/g, and T = 300 K.

To compare relaxation times, particles are initialized in the \hat{x} direction. A constant field is instantaneously introduced in the \hat{z} direction. Though the average magnetization M_x decays to zero without a field due to thermal randomizing, it is forced to zero as the particles align with the field in a time determined by the field amplitude. We plot approximate average magnetizations of the form $M_x = \exp(-t/\tau)$, where the τ are our various time constants. The data for various field amplitudes and anisotropy constants are shown in Fig. 1.

In Fig. 1(a), there is no static field, τ_{hi} is identical to τ_N , and both agree with the stochastic simulations. Here, τ_c is ill-defined. As the field increases but does not overcome the anisotropy energy, as in (b), the equilibrium expression τ_N is no longer accurate, and τ_{hi} is the best approximation. In this regime, classic Stoner–Wohlfarth⁴ type hysteresis curves are found. When the static field amplitude is increased beyond that of the anisotropy as in (c), the high-barrier τ_{hi} approximation breaks down because there are no longer two energy minima. At this point, τ_c is the most valid approximation. Interestingly, the simulated magnetization dips below zero because the high amplitude field also causes increased



FIG. 1. Plots of simulated magnetizations ($i = 10^4$ particles) and calculated exponentially decaying magnetizations with multiple time constants based on different approximations. The combinations of ξ and σ specify the regime and the approximations match as they should.

precession, a physical phenomena that cannot be modeled with the simple exponential-decay model.

Another point is that a typical MNP concentration for hyperthermia ($c_N \approx 10^{13}$ particles/ml) leads to an approximate distance apart of 100 nm, so that the mean dipole fields are orders of magnitude smaller than the typical hyperthermia fields, and affect the dynamics minimally. Other studies show that dipole effects are actually detrimental to heating unless obvious particle chains are formed.¹³

Now simulating an oscillating applied field, we examine how the field amplitude and frequency affect the magnetization. Shown in Fig. 2, with no anisotropy or dipole fields, we see that a hysteresis emerges in plots of the oscillating field against the resulting magnetizations and that there is a peak in area as field amplitude is increased. This can be interpreted by using the characteristic time as in Eq. (8). When the oscillatory applied field is at its maximum value in its cycle, the characteristic time is the shortest. Then, as the field approaches zero, the alignment is slower. Hence, the magnetization takes longer to return to zero than to saturate, and a phase-lag occurs as a result of relaxation. This is distinct from the adiabatic hysteresis that derives from the Stoner–Wohlfarth model.⁴

We use *A* to denote the percentage of the total possible area covered by the normalized hysteresis loop, per cycle of the applied field. *A* is plotted as a function of field amplitude and frequency in Fig. 3. A maximum *A* appears for certain field and frequency combinations and has the correct qualitative scaling behavior expected theoretically by Eq. (9). In particular, if the frequency is increased, the field must also be increased to maintain the same magnetization dynamics. We purposefully stay within the physically tolerable regime⁷ of maximal magnetic field power to suggest hyperthermia could benefit from these peaks or from the flexibility to adjust field *or* frequency to maintain the peak values.

The hysteresis loop area is an enticing metric to visualize peaks when simulating hyperthermia, but, to be practically important, this factor must be put in common experimental units of specific power loss per mass "SLP" of particles (W/g). Thus



FIG. 2. Relaxational hysteresis curves neglecting anisotropy ($\sigma = 0$) for various field amplitudes at 1 kHz. $\alpha = 10$ is used so that the peak in loop area is visible by eye.



FIG. 3. Visualization of the field and frequency dependence of the percent loop area *A*, as predicted by the magnetization dependence.

$$SLP = \frac{\mu H_o f}{\rho V} A,$$
 (10)

where ρ is the particle mass-density in g/m³.

This definition of SLP means that increasing fieldamplitudes or frequencies will increase heating. For the peaks in *A* to affect SLP, they must overpower this linear increase. We only observe this more than linear peaking when $\alpha > 1$. Because α acts as the strength of the rotational magnetic viscosity (the constant in front of the velocity dependent drag), it should be calculated from nanoparticle properties, yet for now remains an experimentally determined parameter and must be treated with care.

Another metric for the dissipative losses is the imaginary component of a Fourier transform of the magnetization. From the same numerical simulations, the imaginary first and third harmonics at several frequencies are plotted in Fig. 4 for a range of field amplitudes. The data are normalized to the maximum value of the first harmonic. In Fig. 4(a), the same peaks are visible at the same fields and frequencies as compared to Fig. 3. In (b), the third harmonic response indicates significant nonlinear components in the lagging magnetization and peaked structure as well. Clearly, here, a physical description only including linear response theory would be incomplete to model heating.

We have demonstrated stochastic methods that allow generalized hyperthermia modeling throughout regimes including equilibrium, large anisotropy, and large applied field amplitudes. We specifically examine the regimes where each approximation is valid to elucidate regime-appropriate timescales. We show in particular that the equilibrium relaxation times will not in general describe the timescales for particle rotations in magnetic fields, and that linear response theory must be used with care as higher harmonics change



FIG. 4. The first and third imaginary harmonic components of a Fourier transform of the magnetization. Peaks at specific field amplitudes and frequencies are visible, and significant higher harmonics are highlighted.

the physics substantially, and the SLP is not reducible to a simple power law.³

We find that a lagging mechanism that causes heat deposition (due to relaxation effects) contains field amplitude and frequency peaks that can be visualized in the imaginary components of the Fourier transform, as well as, via the hysteresis loop area. Varying the field or the frequency accordingly keeps the particles in the optimal regime, a result that can be qualitatively seen from the analytic solution to the high-field magnetization (Eq. (9)). Adding anisotropic contributions will increase heating when the applied field is small enough so that there are still two minima (see Fig. 1). In this case, the relaxation time can be computed from the high barrier approximation Eq. (5). The present results suggest the possibility for tuning the field and frequency separately to optimize heating while maintaining realistic power ranges that are physiologically relevant and practically engineered. Though we only consider Néel rotations in this work, we expect a similar relaxational hysteresis for particles that physically rotate given the similar form of the Brownian differential magnetization equation. In either case, any advantages deriving from the peaks in relaxational hysteresis could be used concurrently with the significant advances garnered from engineering the particles themselves.

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