MagTense: a micromagnetic framework using the analytical demagnetization tensor

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Abstract
We present the open source micromagnetic framework, MagTense, which utilizes a novel discretization approach of rectangular cuboid or tetrahedron geometry “tiles” to analytically calculate the demagnetization field. Each tile is assumed to be uniformly magnetized, and from this assumption only, the demagnetization field can be analytically calculated. Using this novel approach we calculate the solution to the µmag standard micromagnetic problems 2, 3 and 4 and find that the MagTense framework accurately predicts the solution to each of these. Finally, we show that simulation time can be significantly improved by performing the dense demagnetization tensor matrix multiplications using NVIDIA CUDA.

1. Introduction
At the microscopic scale, magnetic systems can be described by the formalism of micromagnetics. The purpose of this formalism is to determine the magnetization field within magnetic materials, and to calculate its time-evolution. Micromagnetic models are used for a plethora of magnetic calculations, including e.g. micromagnetics of rare-earth efficient permanent magnets Fischbacher et al. [2018]; Gong et al. [2019] and calculation of spin waves in nanometre-scale patterned magnetic elements Kim [2010]. The physics of micromagnetics can only in rare cases be solved analytically, and therefore numerical models are needed to simulate the physical behavior of the studied systems. Numerous micromagnetic models exist and some are also published as open source software Donahue and Porter [1999]; Vansteenkiste et al. [2014]; Leli-aert et al. [2018].

In a micromagnetic model, the governing equations must be discretized and represented by a finite number of degrees of freedom Kumar and Adeyeye [2017], and a number of distinct interactions takes place between the discretized moments in the systems. These interactions can be described by different magnetic fields, such as an exchange field and an anisotropy field. Of these it is the calculation of the demagnetization field, also known as the stray field, that is the most time consuming part Abert et al. [2013]. There are in general three methods that have been applied to calculate the demagnetization field: finite difference based fast Fourier transform methods, tensor-grid methods and finite-element methods, although other methods such as fast multipole methods also exist Van de Wiele et al. [2008]; Palmesi et al. [2017]; Kumar and Adeyeye [2017]. All methods have inherent advantages and shortcomings. The finite difference Fourier-based approaches are extremely fast but unfortunately require that the magnetic moments lie on a regular grid Vansteenkiste and Van de Wiele [2011]; Vansteenkiste et al. [2014]; Ferrero and Manzin [2020], although methods for non-uniform grids Livshitz et al. [2009]; Exl and Schrefl [2014] and multi-layer materials Lepadatu [2019] have also been proposed. The tensor-grid method Exl et al. [2012] is still under investigation for stability and correctness Abert et al. [2013]. The finite element method allows easy meshing of complex geometries but also requires that an environment outside the object of interest is modelled to ensure that the correct field is calculated.

So far the direct analytical calculation of the demagnetization field has only been explored in combination with a Fourier transform method Vansteenkiste and Van de Wiele [2011]. Here we present a novel micromagnetic model that uses an analytically exact approach to calculate the full demagnetization tensor for computations in real space. This ensures that the calculated magnetic field is correct, and at the same time allows for easy meshing of complex geometries. The disadvantage is an increase in the computational cost of the model.

2. Physics of the model
2.1 Governing equations
In the following we will denote the magnetization field by $M(x)$, where $x$ denotes an arbitrary point. One of the fundamental assumptions in micromagnetism is that the norm of $M$ is equal at all points to the saturation magnetization $M_s$. Therefore, it is customary to express the magnetization field as $M(x) = M_s m(x)$, where $m$ is the reduced magnetization that is normalized at any point. For heterogeneous materials the saturation magnetization can also be space-dependent, but it is always assumed to be a local known property of the magnetic
material.

The equation of motion governing the dynamics of micromagnetic systems is the Landau-Lifshitz equation J. García-Cervera [2014]:

$$\frac{dm}{dt} = -\gamma m \times H_{\text{eff}} - \alpha m \times (m \times H_{\text{eff}})$$  \hspace{1cm} (1)

where $H_{\text{eff}}$ denotes the effective field, a generalization of the concept of magnetic field that incorporates the effect of different interaction mechanisms. Indeed, the Landau-Lifshitz equation is a time-dependent partial differential equation within continuum physics that describes macroscopic classical effects as well as a semiclassical treatment of collective quantum effects.

The first and second terms on the right-hand side of Eq. 1 are called the precessional and damping term, respectively. In fact, at any point the reduced magnetization vector precesses around the effective field with a frequency determined by the gyromagnetic ratio $\gamma$ and rotates towards the effective field with a speed determined by the damping parameter $\alpha$. Due to the damping term, the cross product $m \times H_{\text{eff}}$ decreases over time, thus reducing the precession amplitude. At equilibrium, the magnetization vector is aligned with the effective field at any point. Therefore, the equilibrium state is described by the first Brown’s equation:

$$m \times H_{\text{eff}} = 0$$  \hspace{1cm} (2)

As can be seen from Eq. 1, when this condition is verified the magnetization is stationary: $\frac{dm}{dt} = 0$. Similarly, the second Brown’s equation states that, at the border of the magnet

$$m \times \left( A_{\text{exch}} \frac{\partial m}{\partial n} \right) = 0,$$  \hspace{1cm} (3)

where $A_{\text{exch}}$ is the exchange interaction constant and $n$ is the normal vector of the boundary. Since $m$ and $\frac{dm}{dn}$ are orthogonal, Eq. 3 can only be fulfilled by $\frac{dm}{dn} = 0$, which corresponds to Neumann boundary conditions. It should also be stressed that the Landau-Lifshitz equation preserves the norm of $m$ at any point.

### 2.2 Free energy and effective field

The effective field is associated with the micromagnetic free energy of the system, denoted by $\mathcal{G}$. The free energy is expressed as the volume integral of a corresponding volumetric energy density $g$:

$$\mathcal{G}[m] = \int_{\Omega} dV \ g(x, m, \nabla m_x, \nabla m_y, \nabla m_z)$$  \hspace{1cm} (4)

where $\Omega$ denotes the region occupied by the magnetic material. As can be seen from Eq. 4, the energy density may depend on the magnetization $m$ and the spatial gradients of its $x$, $y$ and $z$ components. The effective field is proportional to the first variation of the free energy with respect to $m$:

$$H_{\text{eff}} = -\frac{1}{\mu_0 M_s} \frac{\delta \mathcal{G}}{\delta m}$$  \hspace{1cm} (5)

where $\mu_0$ denotes the vacuum permeability. Therefore, the damping term in the Landau-Lifshitz equation has the effect of decreasing the free energy, and the minima of $\mathcal{G}$ with respect to $m$ are equilibrium configurations, i.e. satisfying Brown’s first equation.

### 2.3 Interaction mechanisms

As mentioned in Sec. 2.1, the effective field is composed of several terms associated with different interaction mechanisms, each obtained from a corresponding term of the free energy by calculating the first variation.

- **External field** - This term represents the classical macroscopic magnetostatic interaction of the magnetization vector with an external applied $H_a$. The energy density is given by:

$$g_{\text{ext}} = -\mu_0 M_s (m \cdot H_a)$$  \hspace{1cm} (6)

The effective field associated with this interaction is simply coincident with the applied field itself:

$$(H_{\text{eff}})_{\text{ext}} = H_a$$  \hspace{1cm} (7)

- **Demagnetization field** - Similar to the interaction with the external field, the demagnetization term expresses the classical magnetostatic interaction of the magnetization with the field $H_d$ generated in $\Omega$ by its own magnetization $M(x)$:

$$g_{\text{demag}} = -\left( \frac{\mu_0 M_s}{2} \right) (m \cdot H_d)$$  \hspace{1cm} (8)

The demagnetization field depends linearly on $m$ according to the following expression:

$$H_d(x) = \int_{\Omega} dV' N(x, x') m(x')$$  \hspace{1cm} (9)

where $N$ denotes the demagnetization tensor. As for the previous term, the effective field associated with this interaction is coincident with the demagnetization field:

$$(H_{\text{eff}})_{\text{demag}} = H_d$$  \hspace{1cm} (10)

The effect of this interaction is to reduce the field generated outside $\Omega$ by driving the magnetization vector at any point towards the local magnetic field, thus forming closed loops.

- **Exchange interaction** - This term is a semiclassical treatment of a collective quantum effect. Derived from the Heisenberg Hamiltonian, it expresses the tendency of the magnetization vector in a point to align to the magnetization vector of the surrounding points. In the semiclassical formalism this is achieved by minimizing the spatial derivatives of $m$:

$$g_{\text{exch}} = A_{\text{exch}} \left( ||\nabla m_x||^2 + ||\nabla m_y||^2 + ||\nabla m_z||^2 \right)$$  \hspace{1cm} (11)

where $A_{\text{exch}}$ is the exchange interaction constant and $m$ is the magnetization vector. The energy density $g_{\text{exch}}$ is given by:

$$g_{\text{exch}} = A_{\text{exch}} \left( ||\nabla m_x||^2 + ||\nabla m_y||^2 + ||\nabla m_z||^2 \right)$$  \hspace{1cm} (11)

The exchange interaction term is coincident with the demagnetization field when the reduced magnetization vector is aligned with the effective field at any point.

The effective field associated with this interaction is simply coincident with the applied field itself:

$$(H_{\text{eff}})_{\text{exch}} = H_d$$  \hspace{1cm} (12)

The effect of this interaction is to reduce the field generated outside $\Omega$ by driving the magnetization vector at any point towards the local magnetic field, thus forming closed loops.

- **Brown’s equation** - This term represents the classical macroscopic magnetostatic interaction of the magnetization vector with an external applied $H_a$. The energy density is given by:

$$g_{\text{ext}} = -\mu_0 M_s (m \cdot H_a)$$  \hspace{1cm} (6)

The effective field associated with this interaction is simply coincident with the applied field itself:

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where $A_{\text{exch}}$ denotes the exchange interaction constant. The effective field associated with the exchange interaction can be computed from the corresponding energy density using variational methods:

$$\left(H_{\text{eff}}\right)_{\text{exch}} = \left(\frac{2A_{\text{exch}}}{\mu_0 M_s}\right) \left(\frac{\partial^2 m}{\partial x^2} + \frac{\partial^2 m}{\partial y^2} + \frac{\partial^2 m}{\partial z^2}\right). \quad (12)$$

The effect of this interaction is thus analogous to that of a vector diffusion equation, i.e. driven by magnetization gradients across $\Omega$. As for $M_s$, $A_{\text{exch}}$ may be space-dependent for heterogeneous materials, in which case the effective field would be

$$\left(H_{\text{eff}}\right)_{\text{exch}} = \left(\frac{2}{\mu_0 M_s}\right) \nabla \cdot (A_{\text{exch}} \nabla m) \quad (13)$$

and in case of an abrupt material change,

$$A_{\text{exch}} \frac{\partial m}{\partial n_1} = A_{\text{exch}} \frac{\partial m}{\partial n_2}, \quad (14)$$

where $n$ is the normal vector to the material interface.

### Crystal anisotropy
This term describes the anisotropic effects due to the local orientation of the crystal lattice. For the case of uni-axial anisotropy, the energy density is expressed as:

$$g_{\text{anis}} = -K (m \cdot \hat{e}_K)^2 \quad (15)$$

where $K$ denotes the anisotropy constant and $\hat{e}_K$ is a unit vector oriented along the easy axis of the crystal. These two parameters can also be space-dependent. Typically, $\hat{e}_K$ has a different direction in each crystal grain. The effective field associated to the uni-axial anisotropy term is given by:

$$\left(H_{\text{eff}}\right)_{\text{anis}} = \frac{2K}{\mu_0 M_s} (m \cdot \hat{e}_K) \hat{e}_K. \quad (16)$$

Therefore, the effect of this term is to rotate the magnetization vector towards the local direction of the easy axis, i.e. $\pm \hat{e}_K$. When the anisotropy constant is negative, this term rotates $m$ away from $\pm \hat{e}_K$, towards the plane that is normal to $\hat{e}_K$. In this case, $\hat{e}_K$ and its normal plane are referred to as “hard axis” and “easy plane”, respectively.

These four interactions are the most well studied mechanisms within the field of micromagnetics. However, it is possible to model other physical phenomena with a similar formalism, e.g. magnetostriction 

### 3. Implementation: the MagTense framework

Except for a few cases, most notably that of a uniformly magnetized particle described by the Stoner-Wohlfarth model, micromagnetics problems can only be approached by means of numerical computations. It is thus necessary to discretize the governing equations so that $m(x)$ is represented by a finite number of degrees of freedom. As mentioned previously the most common discretization methods are finite elements methods and finite difference methods typically with a Fourier-based approach. Generally, the computation of the demagnetization field introduced in Eq. 9 is the most computationally intensive task in the calculation and evolution of $m(x)$.

In this work, we present a novel open-source micromagnetsm and magnetostatic framework MagTense, which uses neither the finite difference nor the finite element approach to calculating the demagnetization field, but instead relies on an analytically correct expression of the demagnetization tensor. The discretization approach used in MagTense subdivides the region $\Omega$ into $N$ “tiles”, adjacent to each other. The magnetization field $m(x)$ is thus represented by a vector of $3N$ components, three for each tile.

The demagnetization field generated by each of the tiles at the centers of all of the other tiles is computed analytically by assuming that the tile is uniformly magnetized. The analytical expression of the demagnetization tensor $N$ is known for many different shapes, such as the rectangular cuboid (rectangular prism) Smith et al. [2010] and the tetrahedron Nielsen et al. [2019], thus allowing great flexibility to build meshes that are suitable for the geometry being investigated. The total demagnetization field at any point is obtained by summing the individual contributions from all the tiles in the magnet domain $\Omega$. This concept is illustrated in Fig. 1 for a magnet domain meshed as rectangular prisms.

Besides the calculation of the demagnetization field, the other components of the effective field must also be calculated, i.e. the exchange field, the external field and anisotropy field. For the case of a Cartesian mesh of rectangular cuboids the exchange interaction field can easily be computed by applying standard finite difference schemes. However, several methods are also available for the case of unstructured polyhedral meshes Sozer et al. [2014]. These will be discussed in a subsequent work, as determining the ideal differential operator for an unstructured mesh is not a trivial task. The implementation of the external field and anisotropy terms is straightforward, as these only depend on the local properties of the tile.

### 3.1 The numerical implementation

In the following, we will denote with the symbol $m$ the $3N$-element vector representing the magnetization field over the discretized mesh. Since the exchange, anisotropy and demagnetization terms of the free energy are all quadratic with respect to $m$, the corresponding terms of the effective field are linear. Therefore, they can all be computed by multiplying the $3N$-element vector by corresponding $3N \times 3N$ matrices.
The matrix associated with the quadratic terms, and by Abert et al. [2013]. The external field energy is linear with respect to the vector associated with \( k \) for \( k = 1, \ldots, 3 \). Denoting by \( A \) the matrix associated with the quadratic terms, and by \( b \) the vector associated with the external field term, the \( k \)th component of the total effective field is computed as:

\[
(H_{\text{eff}})_k = \sum_{j=1}^{3N} A_{kj} m_j + b_k \quad \text{for} \quad k = 1, \ldots, 3N
\] (17)

All the matrices need to be computed only once, at the beginning of the simulation. As the magnetization field evolves during the simulation according to Eq. 1, the effective field is updated and the time-derivative of each component of \( m \) can be computed. The integration of the equation of motion is here performed with an explicit Runge-Kutta (4,5) formula.

The framework MagTense is available in both a version that performs the matrix calculations on the CPU, but also a version that uses CUDA for performing the dense matrix-multiplication needed to calculate the demagnetization field, for increasing the calculation speed Lopez-Diaz et al. [2012]; Leliaert and Mulkers [2019]. All examples presented here as well as all model software is available from www.magtense.org.

3.2 Computational resources

The challenge in using the full demagnetization tensor for micromagnetic simulations is that the demagnetization tensor scales as \( N^2 \) where \( N \) is the number of tiles. This is in contract to e.g. the FFT method, where the storage requirement is only of size \( N \) and the computational complexity is \( N\log(N) \). For a micromagnetic problem with periodic boundary conditions the use of the direct demagnetization tensor has been shown to being computationally significantly slower than the evaluation of the tensor using FFT, as expected from this scaling Krüger et al. [2013]. The analytical formulation of the demagnetization tensor can also lead to numerical errors in the evaluation of the tensor Krüger et al. [2013], with the error at large distances being of the same magnitude as the elements of the tensor itself Lebecki et al. [2008], but this is mostly a problem for sample with periodic boundaries, i.e. samples of infinite size, as comparisons for finite-sized samples have shown that the direct calculation of the demagnetization field ensured the lowest error of the various demagnetization field calculations methods Abert et al. [2013].

To reduce the problem of the \( N^2 \) scaling of the demagnetization tensor, MagTense utilizes the fact that the tensor is symmetric Moskowitz and Della Torre [1966], reducing the scaling to \( (N(N+1)/2)^2 \) which is almost \( N^2/2 \). Furthermore, the demagnetization tensor is stored as a float (4 byte) and not a double (8 byte) also reducing the required memory.

However, it is clear that for large-scale micromagnetic simulation the number of elements in the demagnetization tensor will likely have to be reduced, either though a smart meshing of the structure to be simulated or through removing selected entries in the demagnetization tensor. The latter approach will be discussed subsequently in the next section.

4. Verification of the MagTense Framework

To verify the MagTense micromagnetic framework and the use of the full demagnetization tensor to compute the effective field, we consider three of the standard problems in micromagnetics as described in the \( \mu \)mag standard problems NIS. All the models described below are directly available as part of the MagTense framework Bjørk and Nielsen [2019]. For each standard problem the accuracy of the model is assessed, and in a single case the computation time of the model is also evaluated.

4.1 \( \mu \)mag standard problem 2

In standard problem 2, hysteresis curves must be calculated for a thin film of length \( L \), width \( d = L/5 \) and thickness \( t = \)
An external field is applied in the [1,1,1] direction, and the length and the height of rectangular cuboid tiles equal to 100 μm constant and the width of the thin film width d of the exchange interaction. The results are specified in terms of the thin film dimensions change relative to the length scale depend on the scaled geometry of the problem, as the scale hysteresis curve Streibl et al. [1999]. The hysteresis loop will be closed when the magnitude of the field is slowly decreased to calculate the remanence, which is the value of the net magnetization at which the projection of the magnetization along the y-axis is zero, and the remanence, which is the value of the ratio between the magnetization and the magnetic moment.

In MagTense we utilize a Cartesian mesh with the number of rectangular cuboid tiles equal to 100 × 20 × 3 in the x-, y- and z-direction respectively. We calculate a hysteresis loop with an applied field starting at 0.08 H/Mₘ and decreasing the field in 4000 steps to a value of -0.08 H/Mₘ. The micromagnetic damping parameter, α, and the precession parameter, γ, are not specified in the problem statement. Here we take the value of α = 4000, which is close to the value for standard problem 4, while we take γ = 0, as we are only interested in the final steady state configuration. At each value of the field, we evolve the magnetization 40/ns to find the new equilibrium configuration of the magnetization, before the external field is changed again. These 40 ns are sufficient to reach the equilibrium configuration for almost all cases considered. The aim is to calculate the coercivity, which is the value of the field H_{c}/Mₘ at which the projection of the magnetization along the field is zero, and the remanence, which is the value of M/Mₘ at Hₘ = 0.

It should be note that taking γ = 0 is a valid approximation,
even if the Landau-Lifshitz-Gilbert (LLG) equation, i.e. not the Landau-Lifshitz equation presented in Eq. 1, is considered. This is because in the limit of large damping coefficients $\alpha$ the gyromagnetic factor in the LLG scales as $\gamma/\alpha^2$, whereas the damping factor scales as $\gamma/\alpha$. As the damping coefficient tends to infinity, the gyromagnetic factor then tends to 0 much faster than the damping does. Thus the approach of using $\gamma = 0$ is valid for this standard problem.

Shown in Fig. 2 is the thin film coercivity and remanence as a function of $d/l_{\text{ex}}$ both for the MagTense model as well as results published in the $\mu$mag problem solutions NIS. As can be seen from the figure, the MagTense model is in agreement with the majority of the published $\mu$mag results for both the coercivity and the remanence values. In MagTense both of these values converge the slowest at low values of $d/l_{\text{ex}}$. This can be seen in Fig. 2d, where at the start of the coercivity curve, at 0.08 $H/M_s$, the simulation for the lowest $d/l_{\text{ex}}$ requires a number of steps in the external field before the initial magnetization configuration has reached the correct equilibrium state. This is in line with previous results, which have demonstrated that in the small particle limit convergence can be difficult to obtain Donahue et al. [2000].

Here, the difficulty of obtaining converging at the small particle limit is most likely caused by the chosen ODE45 solver, which is not well suited to stiff problems. The demagnetization tensor approach used is correct also for length scales smaller than the exchange length, so it is not an imprecise calculation of the demagnetization field that is causing the slow convergence. An additional factor in the slow convergence at small length scales could be the chosen discretization of the finite difference exchange operator. This is chosen to be a 3-element stencil, which is correct to second order, but when the exchange coupling is strong a higher order operator may be needed. It has also previously been shown that the spatial discretization with an analytical demagnetization calculations can result in the Landau-Lifshitz equation becoming stiff Shepherd et al. [2014].

### 4.2 $\mu$mag standard problem 3

In standard problem 3, the goal is to calculate the minimum energy state in the single domain limit of a cubic magnetic particle of side length $L$. The standard problem consists on calculating the energy of two different states, known as the flower state and the vortex state respectively, to determine the value of $L$ at which there is a change in the minimum energy state of the cube from one of the states to the other.

The edge length of the considered cube, $L$, is similarly to standard problem 2 expressed in terms of the exchange length scale, $l_{\text{ex}} = \sqrt{A_{\text{exch}}/K_{\text{m}}}$, where $K_{\text{m}}$ is the magnetostatic energy density given by $K_{\text{m}} = M_s^2/2\mu_0$. A uniaxial anisotropy with magnitude $K_{z} = 0.1K_{\text{m}}$ along a principal axis of the cube is assumed.

Shown in Fig. 3 is the side length in units of $l_{\text{ex}}$ at which the minimum energy state switches between the flower and the vortex state as a function of resolution, i.e. the number of cubic magnetic tiles in the model. As can be seen from the figure, the crossover points approach two of the three tabulated $\mu$mag solutions as the resolution is increased. Extrapolating the energy crossover to an infinitely fine resolution by fitting a power law of $L_{\text{cross}} = a n^b + c$ to the data gives $L_{\text{cross}} = 8.477 \pm 0.007$, which is in line with two of the published $\mu$mag results.

### 4.3 $\mu$mag standard problem 4

In standard problem 4 the dynamic evolution of a film with a thickness of 3 nm, a length of 500 nm and a width of 125 nm is considered. We assume that the problem can be discretized in a two-dimensional grid, i.e. that the thickness is only resolved by a single tile.

First an initial equilibrium $s$-state must be calculated, which is obtained by first applying a saturating magnetic

**Figure 3.** The side length at which the energy crossover between the flower and vortex state happens, as a function of resolution, i.e. the number of tiles, $n$, in each direction. The tabulated $\mu$mag solutions, which are extrapolated to an infinitely fine grid, are shown as well. Two of these, the Rave et al. and Martins et al. overlap.

**Figure 4.** State of the system at an intermediate step of the relaxation process for applied field 2 in standard problem 4. The magnetization direction in each point is indicated by the black arrows and by the color. In the figure, the thickness of the film in the $z$ direction has been exaggerated for clarity.
field along the [1,1,1] direction and then slowly removing this field and letting the magnetization equilibrate. Following this one of two different magnetic fields is applied instantaneously at \( t = 0 \) to the initial \( s \)-state and the time evolution of the magnetization as the system moves towards equilibrium in the new fields is simulated. The two fields considered are field 1: \( \mu_0 H = [-24.6, 4.3, 0.0] \) mT and field 2: \( \mu_0 H = [-35.5, -6.3, 0.0] \) mT.

The parameters relevant for the problem are an exchange constant of \( A_{\text{exch}} = 1.3 \cdot 10^{-11} \) J/m, a saturation magnetization of \( M_s = 8.0 \cdot 10^5 \) A/m, a damping parameter of \( \alpha = 4.42 \cdot 10^3 \) and a precession parameter of \( \gamma = 2.21 \cdot 10^5 \).

Using MagTense we compute the initial \( s \)-state and the following dynamical evolution for various resolutions of the uniform grid. The resolution is varied as \([20n, 5n, 1]\) with values of \( n = 1 \) to 9. Fig. 4 shows an example of magnetization distribution over the cross-section of the film. This configuration corresponds to applied field 2 after \( t = 0.5 \) ns from the beginning of the relaxation process.

Computing the mean of previously published models as tabulated in the \( \mu \text{mag} \) results, the error for the MagTense model can be defined as

\[
err = \sum_i \int_0^{100\text{ns}} |m_i - \langle m_i, \mu \text{mag} \rangle| \, dt,
\]

summing the different magnetization components and integrating over the first 100 ns of the time evolution. In Fig. 5a the error is shown as a function of resolution. As can be seen from the figure, the error decreases as the mesh resolution is increased, as expected. At a resolution of \([160,40,1]\), i.e. \( n = 8 \), the model has clearly converged, resulting in a minimum error between the model and the \( \mu \text{mag} \) tabulated data. In Fig. 5b the simulation time for the dynamic part of standard problem 4, i.e. without the calculation of the initial \( s \)-state, is shown as a function of resolution for different hardware configuration. The cases considered are a simulation where all computations are done on an Intel W-3235 CPU, as well as calculations where the demagnetization field matrix multiplications are done using CUDA on a P2200, 2080 Ti and a RTX 3090 Nvidia GPU respectively. As can be seen using CUDA for the calculations results in a significant improvement in computation time, with a similar magnitude as previously seen using CUDA micromagnetic frameworks Lopez-Diaz et al. [2012]; Leliaert and Mulkers [2019].
The dynamic evolution of the system is shown Fig. 6 for a resolution of [160,40,1] along with the standard deviation of the published \( \mu_{\text{mag}} \) solutions. As can be seen MagTense predicts an evolution that clearly is in line with the existing published \( \mu_{\text{mag}} \) data.

### 4.3.1 Influence of thresholding the demagnetization tensor

It is of interest to explore whether or not the demagnetization tensor can be reduced in complexity from a fully dense matrix to a sparse one in order to improve the speed and memory use of the simulation. We have therefore implemented a feature in the MagTense framework such that the demagnetization tensor can be reduced in size by removing all elements of the tensor smaller than a given threshold value, i.e. all values below the threshold are truncated. This also allows the demagnetization tensor, which normally is a dense matrix, to be allocated as a sparse matrix, thus saving memory, which is especially important when doing CUDA calculations. The effect of adjusting the threshold value for the computed solution to standard problem 4 is investigated in the following.

We consider a resolution of [160,40,1]. A threshold is chosen and the values which in absolute value is smaller than the threshold is set to zero. For example for a threshold of 20% the absolute smallest fifth of the elements larger than zero in the tensor is truncated to zero. In Fig. 7a the reduction in simulation time versus the error is shown for various values of the threshold, with the calculations performed on a Intel W-3235 CPU. As can be seen a reduction in computation time is possible without sacrificing accuracy. The computed dynamic solution is shown in Fig. 7b for a threshold value of 40% and 70%, respectively, to illustrate that the solution, especially for the 40% case, is still accurate compared to the tabulated \( \mu_{\text{mag}} \) solutions. Note that for small values of the threshold, the calculation time increases because when a threshold is applied the demagnetization tensor is cast as a sparse matrix. When the number of elements in the tensor is still close to the fully dense matrix, there is a significant overhead in the sparse matrix multiplication of the demagnetization tensor with the magnetization, leading to the increase in simulation time.

It should be noted that thresholding the demagnetization tensor is the simplest and crudest method of reducing the number of entries in the tensor and is a technique that should be used with caution. As the magnitude of the field of a single tile decreases as \( \sim 1/r^3 \) away from the tile, this truncation strategy will in essence result in that tiles only "see" other tiles in a more or less spherical region around themselves. This can be problematic as the number of tiles within such a sphere with radius \( r \) grows as \( r^3 \), thus in effect cancelling out the \( 1/r^3 \) decreasing influence of the individual tiles, making the demagnetization tensor important across the whole sample. A better but more complicated strategy for removing elements from the demagnetization tensor is to employ a multi-grid approach that scales the grid with the distance from a respective tile or use an adaptive mesh refinement technique Ramstock et al. [1996]; Hirano and Hayashi [1999]; Garcia-Cervera and Roma [2006]; Sun and Monk [2006]. Another approach is to uses different approximations for the demagnetization tensor depending on the distance to the respective tile Lebecki et al. [2008]; Krüger et al. [2013].

Another strategy could be to calculate the full demagnetization tensor and subsequently perform an FFT transformation of it and use this matrix in calculations, transforming the magnetization to Fourier-space and back in the matrix multi-
application with the demagnetization tensor. For the currently used regular grids this will make the method identical to the existing FFT frameworks, but for irregular grids this may be a computationally efficient strategy.

5. Conclusion

We have presented the MagTense micromagnetic framework, which utilizes a discretization approach of tiles of either rectangular cuboid or tetrahedron geometry to analytically calculate the demagnetization field. The only assumption in this approach is that a tile is uniformly magnetized.

Using this novel approach, which differs from existing finite difference and finite element schemes, we calculated the solution to a range of the \( \mu \text{mag} \) standard problems in micromagnetism and showed that the MagTense framework accurately calculates the solution to each of these.

Finally, we explored the simulation time in MagTense and showed that this can be significantly improved by performing the dense demagnetization tensor matrix multiplications using CUDA, and also that the demagnetization tensor can be reduced in complexity to increase the calculation speed without significantly sacrificing the precision of the solution, albeit with possible caveats depending on the shape of the sample.

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