MIMETIC METHODS FOR LAGRANGIAN RELAXATION OF MAGNETIC FIELDS*

S. CANDELARESI[†], D. PONTIN[†], AND G. HORNIG[†]

Abstract. We present a new code that performs a relaxation of a magnetic field toward a force-free state (Beltrami field) using a Lagrangian numerical scheme. Beltrami fields are of interest for the dynamics of many technical and astrophysical plasmas as they are the lowest energy states that the magnetic field can reach. The numerical method strictly preserves the magnetic flux and the topology of magnetic field lines. In contrast to other implementations we use mimetic operators for the spatial derivatives in order to improve accuracy for high distortions of the grid. Compared with schemes using direct derivatives we find that the final state of the simulation approximates a force-free state with a significantly higher accuracy. We implement the scheme in a code which runs on graphical processing units, which leads to an enhanced computing speed compared to previous relaxation codes.

Key words. magnetic relaxation, mimetic derivatives, Beltrami fields, code generation

AMS subject classifications. 65D25, 47F05, 65M06, 68W40, 76W05, 85-08, 85A30

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1. Introduction. For astrophysical plasmas magnetic diffusivity can be so low that one can assume the plasma to evolve on dynamic timescales according to the ideal induction equation

(1.1)
$$\frac{\partial \boldsymbol{B}}{\partial t} - \nabla \times (\boldsymbol{u} \times \boldsymbol{B}) = \mathbf{0},$$

where \boldsymbol{B} is the magnetic field and \boldsymbol{u} the plasma velocity. Such an evolution equation is most conveniently studied using the Lagrangian description of the fluid, with position vectors of fluid elements represented by $\boldsymbol{x}(\boldsymbol{X},t)$, where $\boldsymbol{x}(\boldsymbol{X},0)=\boldsymbol{X}$. For our purposes we assume that $\boldsymbol{x}(\boldsymbol{X},t)$ is differentiable for all \boldsymbol{X} and t. Equation (1.1) implies that magnetic field lines behave like material lines of the plasma [2, 3, 22], a property which can be expressed with the help of the flow of the velocity field,

(1.2)
$$\frac{\partial x(X,t)}{\partial t} = u(x(X,t),t),$$

which together with (1.1) implies that the magnetic fields at time t = 0 and at t > 0 are related by the pull-back under x(X,t) (see Appendix A):

(1.3)
$$(x^*(t)B)(X,t) = B(X,0).$$

This is a modern formulation of Alfvén's theorem [2]. Here and in the rest of this paper we express the magnetic field B(X,t) as a function of the initial grid positions X and time t. We can also express it as function of the coordinates with the different

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[†]Division of Mathematics, University of Dundee, Dundee, UK (simon.candelaresi@gmail.com, dpontin@maths.dundee.ac.uk, gunnor@maths.dundee.ac.uk).

functional form $\tilde{B}(x(X,t),t) = B(X,t)$. When describing our numerical scheme we will sometimes for simplicity suppress the explicit time dependence and simply write B(X,t) = B(X).

The velocity in (1.1) is coupled to the magnetic field via the magnetohydrodynamic (MHD) momentum balance equation in a highly nonlinear way. However, in order to determine the end state of such an evolution (in the absence of external forces) one only has to know that the presence of a nonzero viscosity will continuously extract energy from the system until a minimum energy state is reached [17]. In the absence of significant gas pressure (low plasma- β) this minimum energy state can be obtained from a simple variation of the magnetic energy density under the assumption of an ideal evolution (equation (1.1)). This results in a condition for a so-called force-free field or Beltrami field

(1.4)
$$(\nabla \times \mathbf{B}) \times \mathbf{B} = 0 \iff \nabla \times \mathbf{B} = \alpha \mathbf{B}.$$

Here α is in general a function of the spatial variables, but due to the solenoidal condition on \mathbf{B} , α has to be constant along field lines, i.e., $\mathbf{B} \cdot \nabla \alpha = 0$.

As long as one is interested only in the minimum energy state of the evolution, one can also prescribe an artificial dynamics instead of using the MHD momentum balance equation. Specifically, if one takes

(1.5)
$$u = \gamma \mathbf{J} \times \mathbf{B}; \quad \mathbf{J} = \nabla \times \mathbf{B}; \quad \gamma > 0,$$

it is easy to prove that the magnetic energy monotonically decreases until a force-free state is reached [6, 27]. This approach is called the magneto-frictional method [5]. J is the electric current density, where we normalize by setting the permeability $\mu_0 = 1$.

The question as to whether, for an arbitrary given initial field B(X,0), a corresponding Beltrami field with the same topology (i.e., satisfying (1.3) for some mapping x) exists, and if so whether it is smooth, is unsolved. Examples where the corresponding Beltrami fields have singularities (typically current sheets) exist [24]. These weak solutions occur in particular for cases where points, lines, or surfaces of vanishing magnetic field strength exist in the initial field. A debate is still ongoing under which conditions nonsmooth solutions can develop from smooth initial fields in regions of nonvanishing magnetic field [7, 13, 14, 19, 23, 25, 26]. This question was first raised by Parker [19], as a possible scenario for the onset of magnetic reconnection in the solar atmosphere, and is also known as the Parker problem.

Studying magneto-frictional relaxation numerically with an Eulerian description requires high spatial resolution in order to reduce numerical diffusion. However, the numerical diffusion can never be completely eliminated with such a standard approach. Consequently, an ideal evolution preserving the topology of \boldsymbol{B} can only be approximated. In order to circumvent this problem, Craig and Sneyd [6] used a Lagrangian approach which directly calculates the mapping $\boldsymbol{x}(\boldsymbol{X},t)$ and hence simulates a perfectly ideal evolution. The method therefore preserves the topology of field lines as well as the magnetic flux through each surface element. Additionally, $\nabla \cdot \boldsymbol{B} = 0$ is automatically preserved, thus eliminating the need for divergence cleaning.

Later, Pontin et al. [20] analyzed the quality of the force-free approximation obtained using the method of Craig and Sneyd [6]. They found that while the numerically calculated value of $(\nabla \times B) \times B$ could be minimized to an arbitrarily small value, the true value of $(\nabla \times B) \times B$ —obtained from independent measures described below—was sometimes much higher. The reason for this discrepancy was identified to be numerical errors in the derivatives that increase as the grid becomes highly

distorted. Accumulation of these errors occurs due to several multiplications of first and second derivatives that are required to obtain an expression for $(\nabla \times \mathbf{B}) \times \mathbf{B}$ in the scheme (see (2.11) and (2.12) of [6]). Consequently, it turns out that $\nabla \cdot \mathbf{J} \neq 0$ can become large for high grid distortions. It was suggested by Pontin et al. [20] that these errors could be reduced by calculating the electric current density using a mimetic method [9, 10]. Derivative operators are then represented as integrals by making use of, e.g., the Stokes or Gauss theorem. One of the great advantages of this approach is that numerically computed curls are discretely divergence free.

In the present work we apply these methods with the two-fold aim of a qualitative improvement of the force-free approximation obtained and a faster convergence. In order to assess the quality of the force-free approximation and computational efficiency of our new scheme, we also implement the classical method, as described by Craig and Sneyd [6]. The two methods are compared throughout the remainder of the paper.

2. Numerical approach: The GLEMuR code.

2.1. Magnetic field relaxation. For the evolution of the velocity field u we use the aforementioned magneto-frictional force (1.5), as it causes the magnetic energy to decay monotonically and the field to evolve toward a force-free state. For the sake of simplicity the parameter $\gamma=1$ is chosen to be constant in time and space. In principal it can depend on space and time and this can be used, for instance, to address concerns about the magneto-frictional method raised by Low [15] for cases of fixed boundaries or null points in the domain. All examples discussed below, however, do not require this.

From the pull-back formula, (1.3), an equation for the magnetic field can be derived [16] (see Appendix A),

(2.1)
$$B_i(\boldsymbol{X},t) = \frac{1}{\Delta} \sum_{j=1}^3 \frac{\partial x_i}{\partial X_j} B_j(\boldsymbol{X},0),$$

where Δ is the determinant of the Jacobian matrix $\partial x_i/\partial X_j$ and measures the local compression or expansion of the medium. This is analogous to Nanson's formula known in continuum mechanics. Equation (2.1) is used to determine \boldsymbol{B} in the Lorentz force, which is required for the numerical integration of (1.2). The other quantity required for the Lorentz force is the electric current which we determine from \boldsymbol{B} via a mimetic operator.

2.2. Mimetic operators. A property of the mimetic differential operators described by Hyman and Shashkov [9] is that they map fields defined on a discrete space, like grid points, onto a different discrete space, e.g., centers of grid faces. The curl operator maps the magnetic field, defined on grid nodes (primal mesh), onto points in the centers of the faces of grid cells (dual mesh), with the result that \boldsymbol{B} and \boldsymbol{J} are known at different locations. This is a general characteristic of mimetic operators, which map their result onto edges, faces, or cells, rather than onto the same grid points.

Terms like $(\nabla \times B) \times B$ require B and $\nabla \times B$ to be known at the same locations. Therefore, using the standard mimetic operators some sort of interpolation needs to be applied. It is not obvious which method or order of interpolation leads to numerical accuracy or stability. Although we do not have a mathematical proof on numerical stability, we will characterize situations for which interpolations may fail.

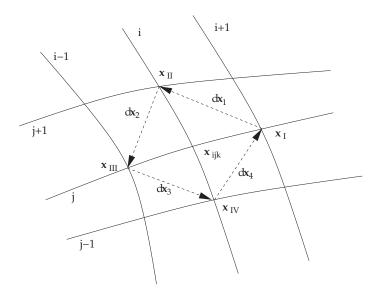


Fig. 1. Schematic representation of the vectors used for the calculations in (2.3)–(2.6). Here only the contribution from the ij-index plane is shown. For the remaining components one simply needs to cyclically permute the indices i, j, and k.

Here we take an alternative approach, as described by Pontin et al. [20], that mitigates this requirement for an explicit interpolation step. The current through a surface U bounded by the closed loop C can in general be computed using Stokes' theorem:

(2.2)
$$\int_{U} \mathbf{J} \cdot \hat{\mathbf{n}} \, dS = \oint_{C} \mathbf{B} \cdot d\mathbf{r}.$$

For the current at the grid point $x_{ijk} = x(X_{ijk}, t)$ we calculate three loop integrals in the three grid surfaces which intersect at this point. For the ij-grid surface this loop is shown in Figure 1. The right-hand side of (2.2) is evaluated as

$$(2.3) \qquad \sum_{r=1}^{4} \boldsymbol{B}_r \cdot \mathrm{d}\boldsymbol{x}_r,$$

with the difference vectors $d\mathbf{x}_i$ defined as $d\mathbf{x}_1 = \mathbf{x}_{\text{II}} - \mathbf{x}_{\text{I}}$, $d\mathbf{x}_2 = \mathbf{x}_{\text{III}} - \mathbf{x}_{\text{II}}$, etc., the magnetic field $\mathbf{B}_1 = (\mathbf{B}(\mathbf{X}_{\text{I}}) + \mathbf{B}(\mathbf{X}_{\text{II}}))/2$, etc., and the position vectors $\mathbf{x}_{\text{I}} = \mathbf{x}(\mathbf{X}_{\text{I}})$, etc., where we use the shorthand notation $\mathbf{x}(\mathbf{X}) = \mathbf{x}(\mathbf{X}, t)$. The left-hand side of (2.2) we approximate by assuming that the current is constant on the quadrilateral,

(2.4)
$$J(X_{ijk}) \cdot \sum_{r=1}^{4} n_r A_r,$$

with the four triangle elements

$$n_1 A_1 = (\boldsymbol{x}_{\text{I}} - \boldsymbol{x}_{ijk}) \times (\boldsymbol{x}_{\text{II}} - \boldsymbol{x}_{ijk})/2,$$

$$n_2 A_2 = (\boldsymbol{x}_{\text{II}} - \boldsymbol{x}_{ijk}) \times (\boldsymbol{x}_{\text{III}} - \boldsymbol{x}_{ijk})/2,$$

$$n_3 A_3 = (\boldsymbol{x}_{\text{III}} - \boldsymbol{x}_{ijk}) \times (\boldsymbol{x}_{\text{IV}} - \boldsymbol{x}_{ijk})/2,$$

$$n_4 A_4 = (\boldsymbol{x}_{\text{IV}} - \boldsymbol{x}_{ijk}) \times (\boldsymbol{x}_{\text{I}} - \boldsymbol{x}_{ijk})/2.$$
(2.5)

The sum of the four surface elements is given as

$$(2.6) \quad \boldsymbol{n}A = \sum_{r=1}^{4} \boldsymbol{n}_r A_r = \left(d\boldsymbol{x}_1 \times d\boldsymbol{x}_2 + d\boldsymbol{x}_2 \times d\boldsymbol{x}_3 + d\boldsymbol{x}_3 \times d\boldsymbol{x}_4 + d\boldsymbol{x}_4 \times d\boldsymbol{x}_1 \right) / 4.$$

Hence, the discretized version of (2.2) at the grid location x_{ijk} reads

(2.7)
$$J(\mathbf{X}_{ijk}) \cdot \mathbf{n} = \frac{1}{A} \sum_{r=1}^{4} \mathbf{B}_r \cdot d\mathbf{x}_r.$$

This equation determines the current in direction n at x_{ijk} . Together with corresponding loops in the jk- and ki-grid surfaces, complete information of all three components of the vector J is obtained. The current components in the X, Y, and Z directions are related to the projections of J along n by a system of linear equations that can be solved by inverting the matrix composed of the three normal vectors n. Note that the three normal vectors have to be linearly independent, which will always be the case so long as the grid does not collapse to being locally two-dimensional. We note that the scheme provided above makes use of a modified version of the approximated normal compared to that used by Pontin et al. [20].

While the above approach removes the requirement of an explicit interpolation step, we note that it is based on the assumption that J can be approximated as constant over the quadrilateral shown in Figure 1. For distortions on the grid scale, e.g., foldings, this approximation will no longer be appropriate.

2.3. Next-nearest-neighbor mimetic approach. For finite difference methods, a higher-order scheme, including further next nearest neighbors, may increase the stability and accuracy of the numerical simulation.¹ To test whether accuracy and stability increase with a modified loop integral for J we perform a similar calculation as in (2.7) using values like $x_{i+1,j+1,k}$ and the magnetic field on those grid points.

Equation (2.3) is here augmented by further neighbors of the point x_{ijk} , forming an octilateral (Figure 2). The loop integral over this octilateral includes eight contributions:

$$(2.8) \sum_{r=1}^{8} \boldsymbol{B}_r \cdot \mathrm{d}\boldsymbol{x}_r,$$

where the difference vectors $d\mathbf{x}_r$ and the magnetic field vectors \mathbf{B}_r are chosen in analogy to equation (2.3) (see Figure 2). The surface elements are

$$n_{1}A_{1} = dx_{2} \times dx_{3}/2, \quad n_{2}A_{2} = dx_{4} \times dx_{5}/2,$$

$$n_{3}A_{3} = dx_{6} \times dx_{7}/2, \quad n_{4}A_{4} = dx_{8} \times dx_{1}/2,$$

$$n_{5}A_{5} = dx_{A} \times dx_{B}/4, \quad n_{6}A_{6} = dx_{B} \times dx_{C}/4,$$

$$(2.9) \quad n_{7}A_{7} = dx_{C} \times dx_{D}/4, \quad n_{8}A_{8} = dx_{D} \times dx_{A}/4.$$

The sum of the surface elements results in a similar equation as (2.6):

$$\boldsymbol{n}A = \sum_{r=1}^{8} \boldsymbol{n}_r A_r = \left(\mathrm{d}\boldsymbol{x}_2 \times \mathrm{d}\boldsymbol{x}_3 + \mathrm{d}\boldsymbol{x}_4 \times \mathrm{d}\boldsymbol{x}_5 + \mathrm{d}\boldsymbol{x}_6 \times \mathrm{d}\boldsymbol{x}_7 + \mathrm{d}\boldsymbol{x}_8 \times \mathrm{d}\boldsymbol{x}_1 \right)/2$$

$$(2.10) + (\mathrm{d}\boldsymbol{x}_A \times \mathrm{d}\boldsymbol{x}_B + \mathrm{d}\boldsymbol{x}_B \times \mathrm{d}\boldsymbol{x}_C + \mathrm{d}\boldsymbol{x}_C \times \mathrm{d}\boldsymbol{x}_D + \mathrm{d}\boldsymbol{x}_D \times \mathrm{d}\boldsymbol{x}_A)/4.$$

Those elements are used in (2.7), where the matrix is inverted to calculate $J(X_{ijk})$.

¹Higher-order derivatives do not necessarily lead to higher accuracies. For sufficiently smooth solutions, however, they lead to increased stability and accuracy for most practical problems.

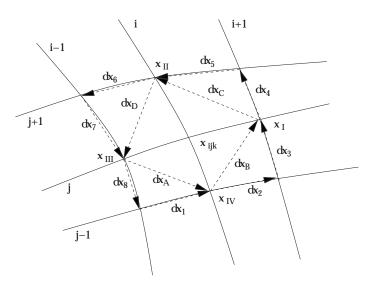


Fig. 2. Schematic representation of the vectors used for the calculations in (2.9) and (2.10). Here only the contribution from the ij-index plane is shown. For the remaining components one simply needs to cyclically permute the indices i, j, and k.

2.4. Time stepping. For the numerical integration of (1.5), we are interested in fast convergence and stability. Adaptive time steps are needed to keep the error within limits. Therefore, we use the method of lines to express the partial differential equations as a set of ordinary differential equations and apply the fifth-order adaptive time step Runge–Kutta formula [4, 21] for the time stepping.

The time step is adjusted according to the error of the calculation. If the error in x exceeds a prescribed limit the step length is reduced via

(2.11)
$$dt' = dt \left| \frac{\Lambda_0}{\Lambda} \right|^{0.2},$$

where dt and dt' are the old and adjusted time steps, Λ the maximum error in \boldsymbol{x} , as calculated in [21], and Λ_0 the desired maximum error (tolerance). If $\Lambda > \Lambda_0$, the result is rejected and recomputed with dt = dt'. Should Λ fall below $\Lambda_0/2$, dt' is increased according to the same equation, thus accelerating computation.

As we are dealing with a highly parallelizable problem, we make use of parallel computing facilities. For that, we developed a numerical code named GLEMuR (Gpu-based Lagrangian mimEtic Magnetic Relaxation), which makes use of the computing power of graphical processing units (GPUs). As the application programming interface we use CUDA [18], which has been tested and has seen various applications in computational analysis.

2.5. Boundary conditions. In the code we implement both periodic boundary conditions and so-called line-tied boundary conditions. A line-tied boundary is a boundary at which the plasma velocity is zero and the magnetic flux through any surface element is fixed (i.e., $\mathbf{B} \cdot \mathbf{n}$ fixed). Periodic boundaries for a moving grid need careful treatment, since periodic grid positions would not be physically consistent. In order to be consistent with (2.1), for a periodic boundary in, say, the z-direction we choose

$$\begin{aligned}
 x_{i,j,f-1} &= x_{ijl}, \\
 y_{i,j,f-1} &= y_{ijl}, \\
 z_{i,j,f-1} &= z_{ijl} - L_z - dZ
 \end{aligned}$$
(2.12)

for the lower boundary, where f and l are the indices for the first and last inner points of the domain in the z-direction and dZ is the initial grid spacing in the z-direction. By analogy, the upper boundary is set to

(2.13)
$$x_{i,j,l+1} = x_{ijf},$$

$$y_{i,j,l+1} = y_{ijf},$$

$$z_{i,j,l+1} = z_{ijf} + L_z + dZ.$$

With these conditions the magnetic field is automatically periodic, i.e.,

(2.14)
$$B(X_{i,j,f-1},t) = B(X_{ijl},t), B(X_{i,j,l+1},t) = B(X_{ijf},t).$$

In the results described in the following sections all boundaries are line-tied, though periodic boundaries do not qualitatively affect these results.

3. Field relaxation.

3.1. Initial configuration. Using the GLEMuR code as described above we compute the ideal evolution of initially twisted magnetic fields starting with a rectangular computational grid. For comparison purposes two initial magnetic field configurations are considered. Our primary focus is on an initial field for which we have an exact closed-form expression for the corresponding force-free field, i.e., we know exactly the expected values of $x(X,t\to\infty)$ and $B(X,t\to\infty)$. This allows us to compare in a straightforward and precise way the quality of the relaxation. The form of the initial magnetic field is given by

(3.1)
$$B(X,0) = \frac{2B_0Z}{a_z^2} \exp\left(-\frac{X^2 + Y^2}{a_r^2} - \frac{Z^2}{a_z^2}\right) \phi\left(Y\hat{e}_x - X\hat{e}_y\right) + B_0\hat{e}_z$$

with the initial magnetic field amplitude B_0 , length of the twist region a_z , width of the twist region a_r , twist angle ϕ , and Cartesian unit vectors \hat{e}_i . Unless explicitly stated, we choose $B_0 = 1$, $a_r = \sqrt{2}$, and $a_z = 2$. The twist angle is chosen as $\phi = \pi/4$, $\phi = \pi/2$, or $\phi = \pi$. The domain is a cuboid with size $L_x = L_y = 8$ and $L_z = 20$ with its center coinciding with the origin of the coordinate system (Figure 3, left panel). Since field lines turn first by some angle around the z-axis and then back by the same angle, determined by the ϕ parameter, we will call this configuration IsoHelix. Although the twist decreases like a Gaussian with distance to the center, there is a small and negligible normal component at the side boundaries of the order of $4.1 \cdot 10^{-5}$.

The expected magnetic field in the relaxed state is of the form

$$\mathbf{B}_{\text{relax}} = B_0 \hat{\mathbf{e}}_z.$$

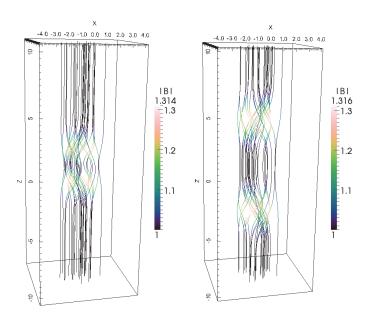


Fig. 3. Initial magnetic field for the IsoHelix field (left panel) and the Pontin09 field (right panel) with $\phi = \pi$ and $\phi_{1/2} = \pm \pi$, respectively. The colors denote the field strength. For readability, only magnetic field lines passing the origin at a radius of 2 are plotted.

For the same configuration we can compute the grid's deformation for $t \to \infty$, which takes the form

$$x_{\text{relax}} = \cos\left(\exp\left(-\frac{X^2 + Y^2}{a_r^2} - \frac{Z^2}{a_z^2}\right)\phi\right)(X\hat{e}_x + Y\hat{e}_y)$$

$$+ \sin\left(\exp\left(-\frac{X^2 + Y^2}{a_r^2} - \frac{Z^2}{a_z^2}\right)\phi\right)(Y\hat{e}_x - X\hat{e}_y)$$

$$+ Z\hat{e}_z.$$
(3.3)

In the following we also mention results obtained using an initial condition identical to that used by Pontin et al. [20]. They applied an initial magnetic field of the form

$$B(X,0) = B_0 \hat{e}_z$$

$$+ \sum_{i=1}^{2} \frac{2B_0 \phi_i}{\pi a_r} \exp\left(-\frac{X^2 + Y^2}{a_r^2} - \frac{(Z - L_i)^2}{a_z^2}\right) \times (-Y \hat{e}_x + X \hat{e}_y),$$

where the symbols are the same as in (3.1), L_i are the distances of the twist regions from the midplane, and ϕ_i are the two twist angles. We choose the domain extent and parameters to be the same values as for the IsoHelix configuration (Figure 3, right panel). Depending on the case we choose either $\phi_{1,2} = \pm \pi/2$ or $\phi_{1,2} = \pm \pi$. The expected relaxed magnetic field is also of the form (3.2). For convenience we will denote this type of initial field as Pontin09.

3.2. Diagnostics.

3.2.1. Force-freeness. The final state of our relaxation simulations should be a numerical approximation to a force-free field. That is, the final magnetic field (relaxed state) should approximately satisfy $\nabla \times \mathbf{B} = \alpha \mathbf{B}$, where α is constant along magnetic field lines. In order to quantify the quality of this approximation we make use of the variable α^* , defined as

(3.5)
$$\alpha^* = \frac{J \cdot B}{B^2}$$

[20], where in an exact force-free state α^* is constant along field lines. The magnitude of the variation of α^* along a magnetic field line provides information on the proximity to force-free equilibrium.

In principle one can choose any field line and test how much α^* varies, but that would require the tracing of field lines, which is computationally expensive and would need high precision. To circumvent this difficulty we choose the central line interval

$$(3.6) s_{\alpha} = \{(0,0,Z) : Z \in [-L_z/2, L_z/2]\}.$$

For the two configurations we know that there is one magnetic field line lying on s_{α} that is invariant in time (by symmetry). Therefore, we monitor the maximum difference of α^* between any two points on s_{α} defined as

(3.7)
$$\epsilon^* = \max_{i,j} \left(a_r \frac{\alpha^*(\boldsymbol{X}_i) - \alpha^*(\boldsymbol{X}_j)}{|\boldsymbol{X}_i - \boldsymbol{X}_j|} \right); \quad \boldsymbol{X}_i, \boldsymbol{X}_j \in s_{\alpha}.$$

3.2.2. Deviation from the exact known equilibrium. For the IsoHelix field one can simply use the deviation between the exact and numerical results as a measure of the quality of the final state. The standard deviation of the magnetic field is simply

(3.8)
$$\sigma_{\boldsymbol{B}} = \sqrt{\frac{1}{N} \sum_{ijk} (\boldsymbol{B}(\boldsymbol{X}_{ijk}) - \boldsymbol{B}_{\text{relax}}(\boldsymbol{X}_{ijk}))^2}$$

with the analytically computed magnetic field $B_{\text{relax}}(X_{ijk})$ for $t \to \infty$ and the total number of grid points N.

In analogy, the deviation from the exact grid deformation is

(3.9)
$$\sigma_{\boldsymbol{x}} = \sqrt{\frac{1}{N} \sum_{ijk} (\boldsymbol{x}(\boldsymbol{X}_{ijk}) - \boldsymbol{x}_{\text{relax}}(\boldsymbol{X}_{ijk}))^2}$$

with the analytically computed grid $x_{\text{relax}}(X_{ijk})$ for $t \to \infty$ given by (3.3).

3.2.3. Convexity. Certain mimetic methods have been shown to be stable for convex cells [11, 12]. For concave cells there is no such proof. It is therefore important to monitor the convexity of the cells. To somewhat simplify the analysis and still retain significance, we define a convexity parameter associated with grid points, although convexity is a property of polygons. At each node X_{ijk} one can define eight trihedra composed by its three nearest neighbors in index space ijk. The three vectors for the trihedra are given as

$$d\mathbf{x}^{\lambda} = \mathbf{x}_{i+\delta_{i},j,k} - \mathbf{x}_{ijk},$$

$$d\mathbf{x}^{\mu} = \mathbf{x}_{i,j+\delta_{j},k} - \mathbf{x}_{ijk},$$

$$d\mathbf{x}^{\nu} = \mathbf{x}_{i,j,k+\delta_{k}} - \mathbf{x}_{ijk}; \quad \delta_{i}, \delta_{j}, \delta_{k} \in \{-1,1\},$$

and the convexity is defined as

(3.11)
$$\kappa(\boldsymbol{X}_{ijk}) = \begin{cases} 1 & \operatorname{sgn}(\det(\mathrm{d}\boldsymbol{x}^{\lambda}\mathrm{d}\boldsymbol{x}^{\mu}\mathrm{d}\boldsymbol{x}^{\nu})) = \delta_{i}\delta_{j}\delta_{k}, \\ -1 & \text{otherwise.} \end{cases}$$

3.2.4. Magnetic energy. As discussed above, a force-free magnetic field corresponds to a minimum of the magnetic energy. It can be demonstrated that the magneto-frictional evolution equation (1.5) implies a monotonic decay of the magnetic energy [6, 27]. The reliability of the methods applied here and the quality of the relaxation is consequently also measured by the evolution of the magnetic energy in the volume V,

(3.12)
$$E_{\rm M} = \int_{V} \mathbf{B}^2 / 2 \, dV.$$

Its numerical computation on a moving grid is not trivial, since the volume $\mathrm{d}V$ surrounding each grid node changes in time. This volume is given by the determinant Δ of the Jacobian matrix multiplied by the corresponding undistorted volume $\mathrm{d}X\mathrm{d}Y\mathrm{d}Z$. Boundary points need to be weighted by a factor ζ , as part of their volume lies outside the domain. For grid points lying on domain faces $\zeta = 1/2$, on edges $\zeta = 1/4$, and on corners $\zeta = 1/8$. Thus, the magnetic energy is

(3.13)
$$E_{\mathrm{M}} = \frac{1}{2} \sum_{ijk} \zeta(\boldsymbol{X}_{ijk}) \, \boldsymbol{B}^{2}(\boldsymbol{X}_{ijk}) \, \Delta(\boldsymbol{X}_{ijk}) \, \mathrm{d}\boldsymbol{X} \mathrm{d}\boldsymbol{Y} \mathrm{d}\boldsymbol{Z}.$$

- 4. Quality of the force-free approximation. Here we describe results obtained using the GLEMuR code with mimetic differential operators based on only nearest neighbors, as described in section 2.2. These are compared with results using the classical approach with second-order spatial finite differences.
- **4.1. Evolution of diagnostic parameters.** As the magnetic field evolves, it approaches the relaxed state, which is captured by the decay of the diagnostic variables ϵ^* for the Pontin09 field and, additionally, σ_B and σ_x for the IsoHelix field (Figure 4, Tables 1 and 2).

The evolution of ϵ^* provides one window into the quality of the force-free field obtained. Comparing the results for the mimetic and classical approaches, we find that for all the configurations investigated here (Figure 4, Tables 1 and 2) the mimetic approach gives a greatly improved relaxation as measured by ϵ^* . The classical method converges to values of the order of one, almost independently of the resolution, while the mimetic approach improves this by more than four orders of magnitude with convergence toward higher resolutions (Tables 1 and 2).

In addition to the above, one can also monitor directly the normalized maximum of the Lorentz force in the domain

(4.1)
$$\xi = \max \frac{|J \times B|}{B^2}.$$

For both methods this can be seen to decay to extremely small values (Figure 4, Tables 1 and 2) that are essentially limited only by numerical roundoff errors. However, as was shown by Pontin et al. [20], these numbers can be highly misleading. In particular, it was shown that for the classical method the Lorentz force is minimized at the expense of the accuracy of, in particular, $\nabla \times B$. Comparing plots for both

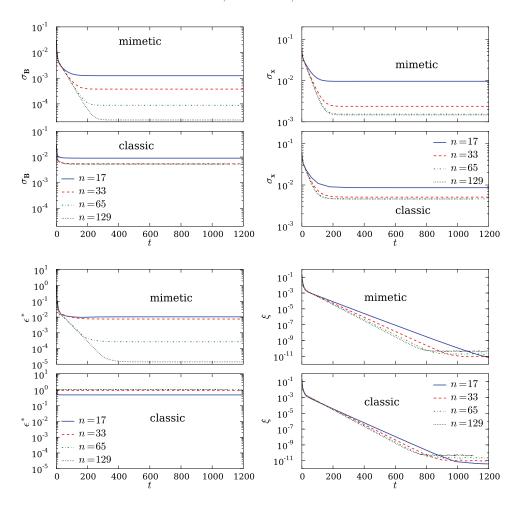


FIG. 4. Comparison of the quality of the field relaxation for the mimetic approach and the classical method as evolution in time for the IsoHelix field ($\phi = \pi/4$). The mimetic approach results in improved quality of the relaxed state as measured in particular by σ_B and σ_x , as well as ϵ^* .

the classical and mimetic methods, we see that ξ continues to decrease even after all independent measures of the force-freeness stabilize to a constant level. As a result, we do not consider the directly calculated value of ξ to be a reliable measure of the true accuracy of the force-free approximation.

Further, the value of ξ strongly depends on the resolution and the tolerance Λ_0 . The former can even have a negative effect if Λ_0 is chosen to be the same irrespective of the resolution. We explain this by the error of the grid deformation during the time stepping (equation (2.11)), where Λ_0 is set to similar values for different resolutions. If the grid error is the same for high and low resolutions, the error in the derivatives is higher for smaller grid separations, which is why we see higher values for ξ .

4.2. Deviations from the analytical solution. For the IsoHelix configuration, we can directly assess the accuracy of the method by comparing the magnetic field and grid distortion with the known exact values as measured by σ_B and σ_x . Like ξ and ϵ^* , σ_B and σ_x decrease over time, indicating the relaxation of the field toward

Table 1

Asymptotic values of the diagnostic parameters as a function of the resolution n, twist angle ϕ , and method for the numerical derivatives for the IsoHelix configuration. Runs marked by † denote the use of double precision arithmetic (64 bit) in contrast to single precision (32 bit). Hyphens mark simulation runs which do not converge.

	M - 41 1		<i>-</i>	ϵ^*		
ϕ	Method	n	ξ	ϵ	σ_{B}	$\sigma_{\boldsymbol{X}}$
$\pi/4$	$Mimetic^{\dagger}$	17	$3.5e{-12}$	$1.0e{-2}$	$1.3e{-3}$	$9.6e{-3}$
$\pi/4$	$Mimetic^{\dagger}$	33	$9.4e{-12}$	7.5e - 3	$3.8e{-4}$	$2.4e{-3}$
$\pi/4$	$Mimetic^{\dagger}$	65	$2.1e{-11}$	$2.7e{-4}$	$8.8e{-5}$	$1.5e{-3}$
$\pi/4$	$Mimetic^{\dagger}$	129	$4.4e{-11}$	$1.4e{-5}$	$2.4e{-5}$	$1.5e{-3}$
$\pi/2$	Mimetic	17	$5.0e{-5}$	$8.3e{-2}$	$4.4e{-3}$	$2.4e{-2}$
$\pi/2$	Mimetic	33	$3.0e{-5}$	$6.1e{-2}$	$1.5e{-3}$	$7.1e{-3}$
$\pi/2$	Mimetic	65	$1.5e{-4}$	$2.3e{-3}$	$3.4e{-4}$	$5.9e{-3}$
$\pi/2$	Mimetic	129	$6.1e{-4}$	$2.0e{-4}$	$1.3e{-4}$	$5.8e{-3}$
π	Mimetic	17	_	_	_	_
π	Mimetic	33	$3.7e{-4}$	$5.4\mathrm{e}{-1}$	$6.4e{-3}$	2.7e-2
π	Mimetic	65	$1.1e{-3}$	$2.7e{-2}$	$1.4e{-3}$	$2.1e{-2}$
π	Mimetic	129	4.7e - 3	$2.4e{-3}$	$1.0e{-3}$	$2.3e{-2}$
$\pi/4$	$Classic^{\dagger}$	17	$3.7e{-12}$	$4.9e{-1}$	$9.0e{-3}$	$8.6e{-3}$
$\pi/4$	$Classic^{\dagger}$	33	$1.0e{-11}$	$9.6e{-1}$	$5.4e{-3}$	$5.1e{-3}$
$\pi/4$	$Classic^{\dagger}$	65	$2.3e{-11}$	1.1	$5.2e{-3}$	$4.6e{-3}$
$\pi/4$	$Classic^{\dagger}$	129	$5.0e{-11}$	1.1	$5.3e{-3}$	$4.6e{-3}$
$\pi/2$	Classic	17	$4.4e{-5}$	$9.9e{-1}$	$2.0e{-2}$	$3.5e{-2}$
$\pi/2$	Classic	33	$7.3e{-5}$	1.9	$1.9e{-2}$	$2.4e{-2}$
$\pi/2$	Classic	65	$1.5e{-4}$	2.2	$2.0e{-2}$	$2.0e{-2}$
$\pi/2$	Classic	129	$7.4e{-4}$	2.2	$2.1e{-2}$	$1.8e{-2}$
π	Classic	17	$5.1e{-5}$	2.0	$5.5e{-2}$	$1.3e{-1}$
π	Classic	33	$8.1e{-5}$	3.8	$6.9e{-2}$	$1.2e{-1}$
π	Classic	65	$3.6e{-4}$	4.3	$7.5e{-2}$	$9.8e{-2}$
π	Classic	129	$8.3e{-3}$	4.4	$7.6e{-2}$	7.4e-2

Table 2

Asymptotic values of the diagnostic parameters as a function of the resolution n and method for the numerical derivatives for the Pontin09 configuration ($\phi = \pi/2$).

Method	n	ξ	ϵ^*
Mimetic	17	$5.4e{-5}$	$1.1e{-1}$
Mimetic	33	$3.3e{-4}$	$1.8e{-2}$
Mimetic	65	$8.9e{-4}$	$7.4e{-4}$
Mimetic	129	$4.0e{-3}$	$6.4e{-4}$
Classic	17	$1.5e{-4}$	$5.0e{-1}$
Classic	33	$1.8e{-4}$	$7.9e{-1}$
Classic	65	$9.8e{-4}$	$8.5e{-1}$
Classic	129	$2.9e{-3}$	$8.6e{-1}$

a force-free state (Figure 4). For the mimetic approach there is a reduction in these quantities, in some cases by more than two orders of magnitude. We also confirm strong improvements with increasing resolution. Their monotonic decay serves us as reassurance that the mimetic approach is very well suited for studying relaxation processes.

4.3. Magnetic energy. Motivated by previous predictions on the magnetic energy evolution [6, 27] we monitor the free magnetic energy $E_{\rm M}^{\rm free}=E_{\rm M}-E_{\rm M}^{0}$, where $E_{\rm M}^{0}$ is the magnetic energy stored in the homogeneous background field $\mathbf{B}_{\rm bkg}=B_{0}\hat{\mathbf{e}}_{z}$. From that analysis we confirm that $E_{\rm M}^{\rm free}$ and $E_{\rm M}$ decrease monotonically in time (Figure 5), which is well established even for very low grid resolutions. The classical approach allows the energy to decay only down to a certain threshold, while the

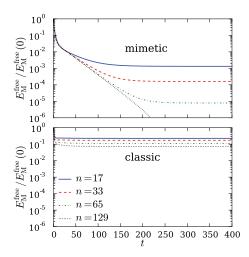


FIG. 5. Normalized relative free magnetic energy in time for the IsoHelix configuration with $\phi = \pi/4$ using the mimetic approach (upper panel) and classical approach (lower panel). As expected, the energy decreases monotonically.

mimetic approach leads to the expected decay of the free energy. This behavior also serves as additional verification that all applied methods are able to reproduce correct results within their limits.

4.4. Grid convexity and stability. Relaxation of the magnetic fields used here results in an untwisted magnetic field, which is achieved by twisting the grid in the opposite sense to the initial magnetic field twist. Increasing the field's initial twist (ϕ) also increases the expected grid distortion of the relaxed state, as the field unwinds itself. Such high distortions lead to concave grid cells, particularly for low resolutions, for which the mimetic operators might not yield a good approximation [12].

The grid distortion is clearly seen in Figure 6 where we plot the grid at the midplane Z=0 at an intermediate time for the Pontin09 configuration ($\phi=\pi$). We also plot the convexity, as defined in (3.11), where red represents convexity and blue concavity. Applying the classical method we find that the grid becomes locally concave (Figure 6, left panel) but the simulation remains stable. The mimetic approach also leads to concave cells (Figure 6, central panel), which subsequently causes jagged grid distortions and the method breaks down (Figure 6, right panel). At this time, we see a blowup of the diagnostic parameters (Figure 7) together with a drop of the time step by several orders of magnitude, at which point the simulation is stopped. Increased grid resolution can delay this blowup. Moreover, it should be stressed that while the classical approach is stable in this case, it does not result in an improved relaxed state, as measured by ϵ^* . Indeed, the mimetic approach before the blowup provides by orders of magnitude a better force-free approximation; see Figure 7.

4.5. Next-nearest-neighbors mimetic approach. Here we apply our next-nearest-neighbor curl operator to compute $J = \nabla \times B$, described in section 2.3. Subject to this study is the field for which we know its analytical solution (equation (3.1)) with $\phi = \pi/4$. For the evolution of ξ , σ_B , and σ_x we observe almost identical behavior as for the nearest-neighbor approach. In that respect there is no advantage of this method over the nearest-neighbor method. By contrast, for ϵ^* we observe an

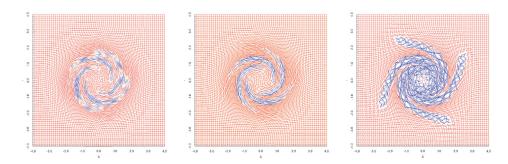


FIG. 6. Grid distortion as seen in the ij-index plane at Z=0 for the final stage of the simulation for the Pontin09 configuration ($\phi=\pi$). Red grid nodes denote convexity, while blue denote concavity as defined in (3.11). The left panel shows the grid close to relaxation using the classical method. The second and third panels show the same configuration using the mimetic method at time t=200 and t=238.2, respectively. The mimetic method breaks down shortly after the grid becomes locally concave.

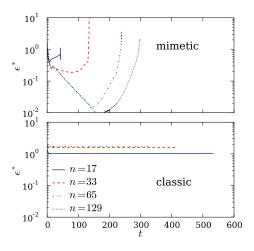


Fig. 7. Time evolution for the Pontin09 configuration $(\phi_{1/2} = \pm \pi)$ of ϵ^* for J computed by using the mimetic approach (upper panel) and classical derivatives (lower panel). For ϵ^* , the mimetic method is far superior in creating a force-free field but lacks in stability for this particular field.

improvement of up to five orders of magnitude (Figure 8). However, this method proves to be unstable for all other configurations discussed herein. Indeed, the numerical instability sets in even before the grid becomes concave, which severely limits its applicability. This suggests that including additional grid points in the mimetic approach is in general not likely to be fruitful.

5. Performance. We compare the computation time for $J = \nabla \times B$ for the classical direct approach, as used by Craig and Sneyd [6], with the mimetic approach. Since the simulation is performed on an NVIDIA graphics card model GTX 765M, we use the NVIDIA Visual Profiler tool to compare the computation time of the computation kernels for a resolution of 65^3 grid points. For computing J the classical approach requires a typical time of about 37.56 ms, while the mimetic approach only needs 19.82 ms.

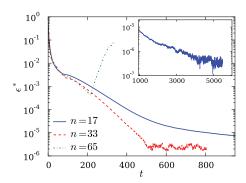


FIG. 8. Time evolution of the force-free measure ϵ^* for the next-nearest-neighbor mimetic approach using the IsoHelix configuration with $\phi = \pi/4$. The inset shows the time evolution for n=17 for longer times. Compared to the nearest-neighbor method there is an improvement of five orders of magnitude as measured by ϵ^* .

Summing up all computationally intensive floating point operations, like multiplications, divisions, and roots, we know that there are 462 multiplications and 7 divisions for the classical method. For the mimetic approach there are only 156 multiplications and 1 division but 12 roots. In both cases, multiplications and divisions by a factor of 2^n with $n \in \mathbb{N}$ are excluded from the operation count, since they only require a bitwise shift. The difference in computational working load approximately reflects the measured timings.

Currently our code runs on single GPUs only. This means that running simulations on multiple graphics cards, like on a cluster, would not increase the computational speed. Since efficient multi-GPU computation for finite difference schemes is rather labor intensive to design we left this open for future work. However, the code is currently designed such that it can in principle run on hardware with any number of multiprocessors, and has run on high-end cards like the NVIDIA Tesla K40. As the development of graphics cards is rapid we will soon be able to use our code on future hardware without computational penalties.

6. Conclusions. The question as to whether for an arbitrary given magnetic field a corresponding force-free field (Beltrami field) with the same topology exists, and if so whether it is smooth, is an important unsolved problem in plasma physics. We have presented here a new code that performs a relaxation of a magnetic field toward a force-free state using a Lagrangian numerical scheme. The method strictly preserves the magnetic flux and the topology of magnetic field lines. In contrast to other implementations we use mimetic operators for the spatial derivatives in order to improve accuracy for high distortions of the grid. We implement the scheme in a code which runs on GPUs, which leads to an enhanced computing speed compared to previous relaxation codes. Compared with schemes using direct derivatives we find that the final state of the simulation approximates a force-free magnetic field with a significantly higher accuracy. Furthermore, as expected, this accuracy improves as the resolution increases. It is found, however, that the method is only numerically stable so long as the cells of the numerical grid remain convex. This places a restriction on the proximity of the initially prescribed field to the corresponding force-free field. Increasing the number of points used in the scheme to consider next nearest neighbors is found to strongly compromise the stability, indicating that this is not a fruitful approach for such schemes.

Appendix A. Derivation of (1.3) and (2.1). To extend the initial discussion about the ideal evolution we express (1.1) in terms of a Lie-derivative of a differential 2-form β associated with the vector \mathbf{B} . The relation between the 2-form and the vector \mathbf{B} is given by the interior product $\beta = i_{\mathbf{B}}\mu$, where μ , is the standard volume form in the domain. In Cartesian coordinates (X^1, X^2, X^3) this reads

$$\beta = \beta_{23} \mathrm{d}X^2 \wedge \mathrm{d}X^3 + \beta_{13} \mathrm{d}X^1 \wedge \mathrm{d}X^3 + \beta_{12} \mathrm{d}X^1 \wedge \mathrm{d}X^2,$$

where

$$\beta_{12} = B_3, \ \beta_{23} = B_1, \ \beta_{13} = -B_2.$$

Hence, (1.1) is equivalent to

$$\frac{\partial}{\partial t}\beta(\mathbf{X},t) + \mathcal{L}_{\mathbf{u}}\beta(\mathbf{X},t) = 0,$$

where \mathcal{L}_{u} is the Lie-derivative with respect to u. This is in turn the differential formulation of

$$(\boldsymbol{x}^*(t)\beta)(\boldsymbol{X},t) = \beta(\boldsymbol{X},0),$$

where the star indicates the pull-back operation (see [1, p. 370] and [8, pp. 140–143]). Writing this out we get

$$(\boldsymbol{x}^*(t)\beta)(\boldsymbol{X},t) = \beta_{ij}(\boldsymbol{x}(\boldsymbol{X},t),t)\frac{\partial x^i}{\partial X^k}\frac{\partial x^j}{\partial X^l}dX^k \wedge dX^l, \quad i,j,k,l \in \{1,2,3\} \text{ and } i < j.$$

One can solve this equation for $\beta(X,t)$, using the formula for the adjoint of the Jacobian matrix. Translating this back into components of the vector field B leads to (2.1).

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