

Models of Earth’s main magnetic field incorporating flux and radial vorticity constraints

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Abstract

We describe a new technique for implementing the known constraints on magnetic fields arising from two hypotheses about the fluid core of the Earth, namely the frozen-flux hypothesis and the hypothesis that the core is in magnetostrophic force balance with negligible leakage of current into the mantle. These hypotheses lead to time-independence of the integrated flux through certain “null-flux patches” on the core surface, and to time-independence of their radial vorticity. Although the frozen-flux hypothesis has received attention before, constraining the radial vorticity has not previously been attempted. We describe a parametrisation and an algorithm for preserving topology of radial magnetic fields at the core surface while allowing morphological changes. The parametrisation is a spherical triangle tessellation of the core surface. Topology with respect to a reference model (based on data from the Oersted satellite) is preserved as models at different epochs are perturbed to optimize the fit to the data; the topology preservation is achieved by the imposition of inequality constraints on the model, and the optimization at each iteration is cast as a bounded value least-squares problem. For epochs 2000, 1980, 1945, 1915 and 1882 we are able to produce models of the core field which are consistent with flux and radial vorticity conservation, thus providing no observational evidence for the failure of the underlying assumptions. These models are a step towards the production of models which are optimal for the retrieval of frozen-flux velocity fields at the core surface.

1 Introduction

Direct observations of the Earth’s magnetic field play an important role, alongside palaeomagnetic measurements, dynamo simulations and laboratory experiments, in contributing to our ultimate understanding of the processes which generate and sustain the field over time. Earth’s main magnetic field has been modelled repeatedly since the first least squares models were derived by Gauss, but the advent of vector satellite observations with launching of the Magsat satellite in late 1979 ushered in an era in which high quality geomagnetic data are routinely used to address the problem of obtaining accurate and consistent models of the Earth’s main magnetic field at the core mantle boundary. Although direct observations do provide the highest resolution images of the field, they remain limited in their resolution, not least because of the presence of the magnetised crust which obscures the shortest wavelength features of the field. Even the most accurate satellite data, currently being provided by the Oersted and Champ satellites, are unable to unequivocally

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determine the number and locations of patches of opposite polarity flux at the core surface, as demonstrated neatly by O'Brien (1996). Therefore there remains an unavoidable ambiguity in the images of the field that they can provide.

Beginning with the work of Roberts & Scott (1964), many techniques have been developed over the last 40 years to produce large scale maps of the geomagnetic field at the core mantle boundary (CMB) from a given data set. Most adopt a least squares rationale for fitting the model to the imperfect data, along with some other methodology for making the problem well-posed; both models at a particular epoch (for example, Shure *et al.*, 1982; Gubbins and Bloxham, 1985; Bloxham *et al.*, 1989) and time dependent models (Bloxham and Jackson, 1992; Jackson *et al.*, 2000) have successfully been produced using this approach.

The thrust of our paper is to test whether it is possible to construct models of the core magnetic field which satisfy the known constraints which arise when one makes two simplifying assumptions: that the core is a perfect conductor and that the mantle is an insulator. Although these assumptions cannot represent geophysical reality, they are likely to be very good approximations; this prejudice can be used in the geophysical inverse problem to choose from the huge space of models compatible with any particular dataset. We see the construction of these models as a prerequisite to the construction of self-consistent models of fluid flow at the CMB.

The frozen-flux hypothesis is motivated by the differing time scales associated with diffusion and advection in the induction equation, which describes the evolution of the geomagnetic field. The difference is so great that over the time period that one is able to generate field models from direct observations of the magnetic field (ca. 400 years) the diffusive term plays a secondary role and can reasonably be neglected. This in turn implies that the magnetic field lines are frozen to the fluid of the liquid outer core and, therefore, changes in the magnetic field are due solely to advection of the field by the flow. Mapping the field at the CMB also gives partial information on the fluid flow at the CMB.

Adopting the frozen-flux approximation imposes rigorous constraints on resulting field models. Under the frozen-flux approximation, null flux curves (curves on which the radial component of the field is zero) are material lines and can neither be created nor destroyed, implying that the topology of field models must be constant throughout time. This means that once a null flux patch (defined as a patch on the core surface bounded by a null flux curve) exists it cannot split apart to form two patches, nor can it merge with any other such patch thereby lessening the overall number.

Null flux patches are subject to additional constraints, namely the constraint of constant integrated flux (Backus, 1968) and (when additional approximations are invoked) the radial vorticity constraint (Jackson, 1996). The frozen-flux constraint simply states that the flux linked by a patch must be preserved at all times; this is described by equation (15) below. The vorticity constraint, described in Jackson (1996)¹, requires that the mantle is an insulator so that the radial current vanishes identically at the CMB. Then Kelvin's celebrated circulation theorem applies and the equatorially-projected area of a null-flux patch must remain invariant in

¹Note a typographical error in Jackson (1996), namely the omission of a minus sign on the right-hand-side of equations (2.9)–(2.11); this does not alter the conclusions.

time. This arises because of the magnetostrophic force balance at the core surface, owing to the very small values of the Ekman and Rossby numbers in the core. Figure 1 illustrates these ideas schematically.

We make it quite clear at this point that we understand that the preservation of topology is the weakest part of the frozen-flux hypothesis: when two null flux curves become very close, it can take a miniscule amount of diffusion for them to merge. That being said, the integrated flux of the merged patch will be the same as the sum of the fluxes of the two original patches, and the same is true of the radial vorticities. To avoid any misunderstanding, let us clarify that we employ the topological constraints as a necessary vehicle to enable us to keep relevant fluxes and vorticities correct; it is indeed the case that in some of our models flux patches have become so close that it would require only tiny diffusion for them to merge. We do not see this as a problem, since they individually contain the correct fluxes and vorticities.

The frozen-flux approximation was examined in some detail in a series of papers dealing first with estimates of secular variation (Gubbins, 1984) and subsequently with the main field (Gubbins and Bloxham, 1985, Bloxham and Gubbins, 1985, 1986). The problem of developing geomagnetic field models for successive epochs that actually incorporate frozen-flux constraints was first considered by Bloxham & Gubbins (1986). They note two methods for testing the frozen-flux hypothesis: comparison of differences in flux integrals between epochs in light of formal error estimates, and construction of models consistent with the constraints, followed by examination of the degradation in quality of fit to the observations. In applying these tests to data from 1959.5, 1969.5 and 1980 they tentatively concluded that there was evidence for violation of the hypothesis — noting that this result was sensitive to the formal error bars on flux integrals. In an analysis over a longer time interval Bloxham and Gubbins (1985) noted that their core field models evaluated at around 65 year intervals from 1715 to 1980 were inconsistent with Backus’s necessary conditions on the frozen-flux hypothesis. In a later development, Constable *et al.* (1993) enforced the frozen flux constraints by constraining the field model being developed, in this case epoch 1945, to have the same flux patch integrals as a given reference model and minimizing differences from the chosen model. The reference model chosen comprised 8 flux patches and was obtained from the Magsat dataset, epoch 1980. Constable *et al.* (1993) concluded that the frozen-flux approximation was valid, at least over this interval, since they found models satisfying the constraints (15).

O’Brien *et al.* (1997) provided a more elegant method of enforcing the frozen-flux constraints, when generating field models. They devised an ingenious method for constraining the topology of any field model by the use of inequality constraints; a simple example will illustrate the point. It is crucial that in a patch of negative flux all the field values remain negative everywhere, otherwise nesting of the patch occurs with a patch of positive flux and the topology of the patch is lost. It is this preservation of topology that is the most difficult aspect of the problem to address, at least from a computational standpoint. O’Brien *et al.* (1997) were able to preserve the topology of the patches, but the positions of the patches were prescribed in the modeling process to conform with those in a reference model — once specified they were not allowed to move, and occasional artificial surgical procedures had to be implemented during the modeling. These limitations may not conform to reasonable field behaviour when one considers the kinds of motions that

appear to take place with westward drift and other forms of secular variation.

In this paper we extend the works of Constable *et al.* (1993) and O’Brien *et al.* (1997) (hereafter known as CPS and OCP respectively) by developing a more flexible algorithm for handling the topological constraints and constructing field models conforming to both the flux and vorticity constraints, defined by (15) and (16). We generate an epoch 2000 reference model, from the high quality Oersted satellite data which has recently been made available, as our starting point for this work. From this we construct four single-epoch models, for the years 1980, 1945, 1915 and 1882, and examine to what extent the frozen-flux approximation is valid over this temporal range by attempting to fit the data satisfactorily, while conforming to both the flux and vorticity constraints. Our results indicate that it is possible to find simple, smooth models of the field which satisfy the required constraints; we therefore consider that there are no observational grounds for rejecting the frozen flux hypothesis, nor the hypothesis of magnetostrophic balance (which has never been tested before).

The organisation of the paper is as follows: Section 2 describes the parametrisation of the magnetic field on the core surface, along with the topology preservation algorithm which is the crux of the procedure for easily finding optimal models fitting the data and the constraints. Section 3 describes the reference model, generated using the recent Oersted dataset, that we use to provide the values of the patch integrals that all other models must achieve. Section 4 describes the results for epochs 1980, 1945, 1915 and 1882.

2 Mathematical Formulation

2.1 Unconstrained modelling

Much of the mathematical detail is explained in CPS and OCP (to which we refer readers), however below we briefly review their methodology, paying particular attention to modifications we have introduced.

Following CPS, whose notation is followed henceforth, the model $B_r(\hat{\mathbf{s}})$ is defined at a finite set of N nodes on the sphere of unit radius, representing the (normalised) surface of the Earth’s core. The nodes are defined by covering the sphere with a tessellation of spherical triangles, generated by bisections and trisections of the regular icosahedron on the unit sphere (Baumgardner and Fredrickson, 1985; see Figure 2). This results in an approximately even distribution of spatial nodes over the unit sphere. This representation turns out to be significantly more flexible than the more familiar representation of the field in terms of spherical harmonics (e.g. Langel, 1987) and is much more suited to our purposes. For example, under this parametrisation the value and sign of the field at a single point is dependent on at most 3 node values (those adjacent to the point in question) whereas under the spherical harmonic representation the point value depends on the weighted sum of the entire set of model parameters; this leads to considerable difficulties when one wants to constrain the sign of the field at a particular point, one of the operations that we find necessary to perform in what follows.

The components of the field \mathbf{B} at a position \mathbf{r} outside the core are given by the convolution of the model

with the appropriate Green's functions:

$$B_{r,\theta,\phi}(\mathbf{r}) = \int_{\partial C} B_r(\hat{\mathbf{s}}) G_{r,\theta,\phi}(\mathbf{r}, \hat{\mathbf{s}}) d^2 \hat{\mathbf{s}} \quad (1)$$

Here the subscripting r, θ, ϕ is shorthand for the three spherical components (r, θ, ϕ) , $\hat{\mathbf{s}}$ is a position on the core surface, ∂C indicates integration over the core surface, and the circumflex denotes a unit vector. G denotes the Green's function which, for each spherical component, is given by

$$G_r(\mathbf{r}, \mathbf{s}) = \frac{1}{4\pi} \left[\frac{\rho^2(1 - \rho^2)}{R^3} - \rho^2 \right], \quad (2)$$

$$G_\theta(\mathbf{r}, \mathbf{s}) = -\frac{1}{4\pi} \frac{\rho^3(1 + 2R - \rho^2)}{R^3 T} (\hat{\mathbf{s}} \cdot \hat{\boldsymbol{\theta}}), \quad (3)$$

$$G_\phi(\mathbf{r}, \mathbf{s}) = -\frac{1}{4\pi} \frac{\rho^3(1 + 2R - \rho^2)}{R^3 T} (\hat{\mathbf{s}} \cdot \hat{\boldsymbol{\phi}}), \quad (4)$$

Here, the core radius $s = 3485\text{km}$, $\rho = s/|\mathbf{r}|$, $\mu = \hat{\mathbf{s}} \cdot \hat{\mathbf{r}}$, $R = \sqrt{1 - 2\mu\rho + \rho^2}$ and $T = 1 + R - \mu\rho$ (see CPS for more details).

To generate a continuous representation of the field everywhere, B_r at position \mathbf{s} within and on the edges of each spherical triangle is defined by linear interpolation of the vertex values, after the triangle is projected onto the plane tangent at its circumcentre using a gnomonic projection (see CPS):

$$\mathbf{s} = \sum_{i=1}^3 \alpha_i \mathbf{x}_i \quad \text{and} \quad B_r(\mathbf{s}) = \sum_{i=1}^3 \alpha_i B_r(\mathbf{x}_i) \quad (5)$$

where B_r is defined at vertices \mathbf{x}_i of the spherical triangle, and $\sum_{i=1}^3 \alpha_i = 1$. A complementary view is that attached to each node is a two-dimensional tent-like basis function, with 5 or 6 planar faces rising to the apex at the node in question (the ambiguity in the number of faces arises from the different number of nearest neighbours resulting from the original tessellation: the 20 nodes of the original icosahedral tessellation have 5-fold coordination). Figure 3 illustrates an example basis function for a 5-fold coordinated node, showing the local support. The basis function shows the influence of the model parameter in synthesising the model: the entire model is generated by summing the basis functions weighted by their node values. In one-dimensional problems, these basis functions are known as the B-splines of order 2 (see for example, Lancaster & Salkauskas, 1986).

We note that in order for monopoles to be excluded from the model, we must impose the additional constraint

$$\int_{\partial C} B_r d^2 \hat{\mathbf{s}} = 0 \quad (6)$$

which would not otherwise be obeyed by an arbitrary model under this parametrisation (though note that a model computed with frozen flux constraints imposed [§2.2] will automatically fulfill this condition provided the reference model does). We can write the general relationship between a datum d_j at position \mathbf{r} and the model \mathbf{b} as

$$d_j = \mathcal{F}_j(\mathbf{b}) \quad (7)$$

where $\mathbf{b} = \{b_i\}$ are the N model parameters (i.e. the values of the field at the N nodes) and \mathcal{F}_j represents the forward functional relating the model \mathbf{b} to the datum d_j . For the models reported in this paper $N = 812$. In the case of linear data, such as measurements of the X , Y or Z components of the field, the relationship between a datum can be written

$$\mathbf{d} = \mathbf{A}\mathbf{b} \quad (8)$$

and \mathbf{A} contains the inner products of the appropriate Green's function with the basis functions for each of the node values b_j ; these form the rows of the so-called design matrix. Inner products for the forward problem (and for the fluxes and radial vorticities) are computed using a weighted sum of samples of the function in the plane triangle, using an optimal degree-5 cubature rule (again, see CPS for more details).

By virtue of the compact support of the basis functions which describe B_r , the location of every null flux curve is defined solely by the values of the nodes which are immediately adjacent to it. Thus finding null flux curves requires simply looking for adjacent nodes of opposite sign. This feature is central to the application of inequality constraints which is described in §2.2.

Following the standard regularised least squares methodology (e.g. Parker, 1994) we seek a model estimate $\mathbf{b} = \{b_1, b_2, \dots, b_N\}$ which fits the data $\mathbf{d} = \{d_1, d_2, \dots, d_M\}$ while minimising some measure of the model complexity, i.e. we seek that \mathbf{b} which minimises

$$\Gamma(\mathbf{b}) = \sum_{j=1}^M \left(\frac{d_j - \mathcal{F}_j(\mathbf{b})}{\sigma_j} \right)^2 + \lambda_s \mathcal{R}(\mathbf{b}) = (\mathbf{d} - \mathcal{F}(\mathbf{b}))^T \mathbf{C}_e^{-1} (\mathbf{d} - \mathcal{F}(\mathbf{b})) + \lambda_s \mathcal{R}(\mathbf{b}) \quad (9)$$

Here σ_j is the uncertainty in d_j which enters the (diagonal) data covariance matrix \mathbf{C}_e . The final term \mathcal{R} , describes the model complexity and the damping parameter λ_s represents the trade-off between model smoothness and fit to the data: we use λ_s to achieve a fit to the data of one standard deviation on average, namely that the misfit

$$\mathcal{M} = \sqrt{\frac{1}{M} \sum_{j=1}^M \left(\frac{d_j - \mathcal{F}_j(\mathbf{b})}{\sigma_j} \right)^2} \quad (10)$$

has the value unity. In what follows we adopt the F_4 norm of Shure *et al.*, (1982) as our regularisation; we minimise the squared horizontal gradient of the radial component of the field, integrated over the core surface:

$$\mathcal{R} = F_4(B_r) = \int_{\partial C} |\nabla_h B_r|^2 d^2 \hat{\mathbf{s}} = \mathbf{b}^T \mathbf{C}_m^{-1} \mathbf{b} \quad (11)$$

(Other forms of \mathcal{R} could equally well have been used, for example the Ohmic heat norm or the dissipation norm: qualitatively the same result would be obtained.) For more details on the evaluation of \mathcal{R} see CPS.

In (9) the fit to the data is measured by the squared misfit, a commonly used criterion for discriminating between models (see the discussion in Parker, 1994). This would also be a maximum likelihood estimator if the errors contaminating the data set were to originate from a Gaussian distribution. However, we caution that there is accumulating evidence that the distribution of errors may be far from Gaussian (for example

Walker and Jackson, 2000). Nevertheless, bearing this cautionary note in mind we shall proceed with the least squares approach. In general \mathcal{F} is nonlinear so an iterative approach is employed: \mathcal{F} is linearised about an estimate of the solution \mathbf{b}^k , where the superscript denotes the k th estimate, and then (9) is solved for an estimate \mathbf{b}^{k+1} :

$$\mathbf{b}^{k+1} = \left[\mathbf{A}^T \mathbf{C}_e^{-1} \mathbf{A} + \lambda_s \mathbf{C}_m^{-1} \right]^{-1} \mathbf{A}^T \mathbf{C}_e^{-1} \left(\mathbf{d} - \mathcal{F}(\mathbf{b}^k) + \mathbf{A} \mathbf{b}^k \right) \quad (12)$$

where \mathbf{A} is the matrix of derivatives of the data with respect to the model at the k^{th} iterate. As in OCP we use the Newton algorithm variant that solves for the vector \mathbf{b}^{k+1} , rather than a perturbation $\delta \mathbf{b}$ to \mathbf{b}^k , in order that inequality constraints (which apply to \mathbf{b}^{k+1} and not to $\delta \mathbf{b}$) can be imposed. The linear system is actually solved using the QR algorithm, which obviates the need to generate the normal equations $\mathbf{A}^T \mathbf{C}_e^{-1} \mathbf{A}$; the QR version (Lawson & Hanson, 1995) of the system is passed to the algorithm BVLS (see §2.2 and OCP) when constraints are imposed. Clearly for linear data the terms $\mathcal{F}(\mathbf{b}^k)$ and $\mathbf{A} \mathbf{b}^k$ cancel, but, somewhat surprisingly, this is also the case for H and F intensity data, illustrating that the nonlinearity is very slight for these data. We illustrate this for H , but a similar relation follows for F :

$$\frac{\delta H}{\delta b_i} = H^{-1} \left(X \frac{\delta X}{\delta b_i} + Y \frac{\delta Y}{\delta b_i} \right) \quad (13)$$

so

$$\mathbf{A} \mathbf{b}^k = \frac{\delta H}{\delta b_i} b_i = H^{-1} \left(X \frac{\delta X}{\delta b_i} b_i + Y \frac{\delta Y}{\delta b_i} b_i \right) = H^{-1} (X^2 + Y^2) = H = \mathcal{F}(\mathbf{b}^k) \quad (14)$$

The iterative procedure continues until a converged (see later) solution is obtained. For all nonlinear data we work with the true magnetic elements (X, Y, Z, H, F, I, D), rather than the linear Bauer (1894) representation of I and D employed by OCP and CPS. This is because the flux and radial vorticity constraints (see below) imposed on the model result in the problem being nonlinear; we found that the additional nonlinearity due to I and D caused little extra computational expense.

2.2 Constrained modelling

The frozen-flux approximation (Roberts and Scott, 1965) has some severe consequences on the magnetic field at the core surface. Since diffusion is neglected, null flux curves (curves on which the radial magnetic field $B_r = 0$) cannot connect or break, thus implying that the topology of the field must be preserved throughout time. Further to this there are constraints on each patch (defined by the region on the core surface enclosed by a null-flux curve) which must also be preserved throughout time; namely the flux constraint (Backus, 1968) and the radial vorticity constraint (Jackson, 1996).

The flux and vorticity constraints are given by

$$\mathcal{B} = \frac{d}{dt} \int_{S_j} B_r d^2 \hat{S}_j = 0 \quad (15)$$

$$\mathcal{V} = \frac{d}{dt} \int_{S_j} \cos \theta d^2 \hat{S}_j = 0 \quad (16)$$

respectively, where S_j is the region bound by a null-flux curve and θ is the colatitude (see Backus, 1968; Jackson, 1996 for more details). To enforce (15) and (16) we follow CPS (who only consider constraint (15)) and define a reference model, \mathbf{b}^* say, which designates the number of patches, P , to be preserved, together with the flux and vorticity associated each patch. This gives us a set of P target fluxes and a set of P target vorticities which any epoch model estimate must match. In enforcing constraints (15) and (16) on the model estimate, the function (9) is expanded to the problem of minimising

$$\Gamma(\mathbf{b}) = \sum_{j=1}^M \left(\frac{d_j - \mathcal{F}_j(\mathbf{b})}{\sigma_j} \right)^2 + \lambda_s \mathcal{R}(\mathbf{b}) + \lambda_f \sum_{j=1}^P (\mathcal{B}_j(\mathbf{b}) - \mathcal{B}_j(\mathbf{b}^*))^2 + \lambda_v \sum_{j=1}^P (\mathcal{V}_j(\mathbf{b}) - \mathcal{V}_j(\mathbf{b}^*))^2 \quad (17)$$

Here λ_f and λ_v are parameters defining how well constraints (15) and (16) are enforced. For example, $\lambda_f = 0$ implies that (15) is not enforced, while $\lambda_v = 0$ implies (16) is not enforced. Equations (15) and (16) are implemented via quadratic constraints in (17) and therefore $\mathcal{B}_j(\mathbf{b}^*)$ and $\mathcal{V}_j(\mathbf{b}^*)$ can be thought of as $2P$ additional “data”, supplied by the reference model, and the “forward function” for each of the $2P$ “data” is made up of a suitably weighted vector, describing the contribution of the nodes within patch S_j to its flux or vorticity integral.

In order to optimize the system (17) we use a Newton algorithm. For the linear data we use the matrix linking the data to the model parameters, and for the non-linear data we use the Frechét derivatives of the data with respect to the model parameters. For the patch integrals we need the derivatives with respect to perturbations in the model. Consider the general integral

$$I = \int_{\mathbf{x}=f(\mathbf{b})} g(\mathbf{b}, \theta, \phi) dS \quad (18)$$

Then the derivative is in general

$$\delta I = \int_{\mathbf{x}=f(\mathbf{b})} \delta g(\mathbf{b}, \theta, \phi) dS + \oint_{\mathbf{x}=f(\mathbf{b})} g(\mathbf{b}, \theta, \phi) (\delta \mathbf{x} \cdot \hat{\mathbf{n}}) dl \quad (19)$$

where $\delta \mathbf{x} \cdot \hat{\mathbf{n}}$ is the component of the perturbation of the boundary in the direction $\hat{\mathbf{n}}$ normal to the boundary due to a change in the model. We find

$$\delta \mathbf{x} \cdot \hat{\mathbf{n}} = - \frac{\delta B_r}{|\nabla_{\mathbf{h}} B_r(\mathbf{r})|} \quad (20)$$

for both our integrals where the boundaries are the null-flux curves. Although the null-flux integrals (15) are non-linear, it is a felicitous fact that a perturbation $\delta \mathcal{B}$ can be related to a perturbation in the model $\delta \mathbf{b}$ as

$$\delta \mathcal{B} = \mathbf{B} \delta \mathbf{b} \quad (21)$$

with no approximation at all by virtue of the fact that the boundary terms in the perturbation of (15) vanish (Bloxham & Gubbins, 1985). This is not the case for the Frechét derivatives of \mathcal{V}_j with respect to the b_i , for which the boundary integral is the only non-vanishing contribution.

The system is solved by iterating to a converged solution using a Bounded Variable Least Squares algorithm (BVLS) to find a field model which retains the topology of the given reference model, \mathbf{b}^* (described in the next section, §2.3). We omit a description of the methodology of BVLS here and refer readers to Lawson and Hanson (1995) and Stark and Parker (1995) who give a lucid description of the method. With the topological constraints satisfied we are able to enforce constraints (15) and (16) by choosing λ_f and λ_v , appearing in (17), such that a desired conformity is achieved. The final converged model is such that both the normalised misfit and model complexity (\mathcal{R}) have converged, the model fits the data satisfactorily, and the average percentage error (averaged over the P patches) of the flux and vorticity constraints is less than 1%.

2.3 Topology Preservation

At the heart of our optimization procedure is an algorithm to retain topology of a model from iteration to iteration while allowing changes in morphology. Topology preservation ensures that the number and nestedness of patches remains the same; morphological changes are allowed, so that patches can change their shape, size and location.

The concept of neighbours is central to the algorithm. Under the icosahedral discretization the coordination (or the number of nearest-neighbours) of any node is either sixfold or fivefold (the former being overwhelmingly the more common). The “sign” of a node is the sign of the value of b_i at that node. At any one iteration only certain nodes are allowed to change sign from their current values. The nodes which are allowed to change sign form the “unconstrained set” \mathcal{U} , and the nodes which are not allowed to change sign form the “constrained set” \mathcal{C} . All variables are assigned upper and lower bounds \mathbf{u} and \mathbf{l} so that $l_j \leq b_j \leq u_j$. The constrained set are assigned upper and lower bounds in the following way:

$$\begin{aligned} \text{If } b_j \in \mathcal{C} \text{ and } b_j < 0 \quad l_j &= -|\aleph|; \quad u_j = -|\epsilon| \\ \text{If } b_j \in \mathcal{C} \text{ and } b_j > 0 \quad l_j &= |\epsilon|; \quad u_j = |\aleph| \end{aligned} \tag{22}$$

where in our application we have set $\epsilon = 10\mu T$ and $\aleph = 10^4\mu T$, though in reality the upper bounds are generally impotent because of the regularisation of the model that is applied.

Only nodes whose radial field value is close to zero are likely to usefully change sign, so on any one null flux curve ∂S we begin by ranking adjacent nodes in terms of their proximity to ∂S . Once the ordered list has been made we move down the list, assigning to \mathcal{U} each node sequentially, provided it satisfies a set of “topology rules”. We distinguish between “inside” and “outside” the patch (the distinction, although arbitrary, can be definitely assigned). We give an example in Figure 4 of the unconstrained nodes at a particular iteration of a model. The nodes marked with black dots are unconstrained, and are free to change sign. All other nodes must retain their current sign. The figure illustrates the most important topology rule: if a node is an interior node then it is prevented from being assigned to \mathcal{U} if any exterior neighbour is already in \mathcal{U} (and vice versa). This prevents the creation of new null flux patches. There are a considerable number of more elaborate rules designed to deal with patches containing very small numbers of nodes; a

comprehensive list of the rules is detailed in the Appendix. As an example, there should be more than one node within any patch at any time. A patch containing only one node would necessarily always have that node’s sign constrained (to prevent the patch disappearing) but this means that the patch cannot migrate without first enlarging. Problems such as this one can be eliminated by increasing the resolution of the model.

The algorithm works extremely effectively: patches can migrate easily across the core from iteration to iteration; changes in size and shape are accomplished smoothly as the model vector is updated.

3 The Reference Model

The reference model, \mathbf{b}^* is generated from a subset of the high quality Oersted data set described in Olsen *et al.* (2000). The dataset comprises 3907 vector triples (B_r, B_θ, B_ϕ) together with 2148 total field intensity measurements, F , spanning an interval of approximately 4 months around epoch 2000. Using the secular variation estimate of the IGRF we reduced the data to epoch 2000 and, furthermore, we remove the Dst-dependent external magnetic field model of Olsen *et al.* (2000). Finally, following OCP, we decimate the resulting data into 5 degree cells in order to remove any possible error correlation due to crustal noise. This results in a final data set of 1228 vector triples and 998 scalar measurements with excellent global coverage.

Associated with each data measurement is an error estimate and, following Olsen *et al.* (2000), we allow for $2.25nT$ in the scalar measurements F . Holme and Bloxham (1996) describe a method in which the vector triples (B_r, B_θ, B_ϕ) are rotated to a new coordinate system (B_B, B_\perp, B_3) , where B_B is in the direction of the field \mathbf{B} at the data coordinate, and B_\perp, B_3 are in the directions $\hat{\mathbf{n}} \wedge \mathbf{B}$ and $\mathbf{B} \wedge (\hat{\mathbf{n}} \wedge \mathbf{B})$ respectively. With respect to the Oersted data $\hat{\mathbf{n}}$ represents a unit vector in the direction of the SIM bore-sight (Olsen *et al.*, 2000). In this new rotated coordinate system the errors on the field components (B_B, B_\perp, B_3) are uncorrelated. Following Holme and Bloxham (1996), (B_B, B_\perp, B_3) are assigned weightings defined by the parameters χ and ψ , together with the field strength $|\mathbf{B}|$. We set $\chi = 60''$, $\psi = 10''$, see Olsen *et al.* (2000).

In Figure 5 we illustrate \mathbf{b}^* resulting from the Oersted data, plotted at the core surface $s = 3485km$. We implement a clamping scheme to make the least-squares method insensitive to outliers: final iterations are performed rejecting data whose residuals exceed 3 standard deviations. The normalised misfit \mathcal{M} of the model is 1.01 and the number of data rejected is 26 vector triples and 171 scalar measurements. Although this may seem quite a large number of rejected F data, we account for this by the fact that topological constraints (8 patches) are enforced on the model, so as to make direct comparisons with the work of CPS. (If the topological constraints are removed a reference model can be found with misfit of 1.01; in this case only 40 F data are rejected, however the model comprises 11 patches.) The flux strengths and vorticities of each patch, together with a description of their position, can be found in Tables 1a and 1b, where we use the same patch identification as CPS. Comparing Figure 5 with the 1980 reference model of CPS (therein called STT1980) we see that the reference models are similar, although \mathbf{b}^* is slightly rougher than STT1980 (also

indicated by the flux integrals of \mathbf{b}^* (Table 1a) being of greater magnitude). We find this to be the case for our own models for 2000 and 1980, see Table 2. This is to be expected as the Oersted data are of higher quality than those used by CPS (Magsat data) when developing STT1980.

4 Results

We now proceed to investigate whether it is possible to generate field models consistent with the epoch 2000 reference model \mathbf{b}^* , described above. If we are able to obtain such models which, while fitting the data adequately, satisfy both the flux and vorticity constraints, then we are in a strong position to state that the frozen flux approximation of Roberts and Scott (1965) is valid, at least over the time scale considered in this paper.

We investigate the validity of the frozen flux approximation over five epochs; namely 2000 (which provides the reference patch integrals), 1980, 1945, 1915 and 1882. The 1980 Magsat dataset is described in detail in OCP, to which we refer readers for more details. Here we merely state that the data are linear and comprise 1262 X , 1262 Y and 1654 Z measurements. The data set for 1945 comprises observatory and survey data and is described in detail in CPS; briefly the data are made up of 555 X , 554 Y , 634 Z and 322 D measurements. The data distribution is poor in contrast to the satellite data as it is confined to land based measurements (see figures 3, 4, 5 of CPS). However this gives us a good opportunity to study the effect of differing data coverage on the constrained modelling. The 1915 dataset is derived from the 20th century historical database described in Bloxham *et al.* (1989), and contains a considerable amount of oceanic data from the ship *Carnegie*. The dataset comprises 1156 X , 1021 Y , 1455 Z , 628 H , 173 I and 1394 D data. Finally the data set for 1882 is the one used in Walker and Jackson (2000) and comprises approximately 10 000 observations with reasonably global coverage (see Figure 2 of Walker & Jackson(2000)).

In Figure 6 we illustrate the radial component of the Earth's magnetic field, plotted at the core mantle boundary, for epoch 1980, constrained by \mathbf{b}^* . Comparing Figures 5 and 6 we see that they are remarkably similar; a fact which is reassuring for two reasons. First both models are generated from high quality data with excellent global coverage and second, and more importantly, the two models are only 20 years apart. The observable differences between the two models are fairly small scale, for example the slight change in the negative flux patch over the Antarctic. The reference model \mathbf{b}^* is slightly rougher than epoch 1980; this is to be expected as the data from Oersted is of a higher quality (smaller error budget) than that of the Magsat mission. Comparing fluxes (Table 1a) and vorticities (Table 1b) of epoch 1980 with \mathbf{b}^* we see that they give excellent agreement, indicating that the frozen flux approximation is valid over this short period. Comparing Figure 6 with STT1980 of CPS we see that they are in good agreement, although Figure 6 is slightly rougher than STT1980 due to the stronger constraints placed on each patch.

Figure 7 illustrates the model estimate for epoch 1945. Comparing with \mathbf{b}^* we first note how much smoother the model is, as shown by the value of \mathcal{R} for the two models. This is because the data is confined to land based observations, thus giving a much poorer global coverage than the satellite data used to construct

the reference model. However, even with the contrast in model complexities, we are able to produce a model for epoch 1945 satisfying both the flux and vorticity constraints of \mathbf{b}^* , see Tables 1a, 1b. Readers might be concerned that visually the patch fluxes in Figures 5 and 7 appear not to be conserved (for example, compare the positive (red) southern hemisphere patches); rest assured that this is not the case. The large peaks in magnetic field strength are more localised in Figure 5, whereas in Figure 7 they have been smoothed out over a greater area. The net effect is to conserve the flux, as illustrated in Table 1a, but to the naked eye the models may look somewhat misleading. The main difference between the morphology of the two field models is in patch 3, the “Bone” in the southern hemisphere. We feel that this is a consequence of the contrast in data distributions over the southern region, rather than an effect due to the constraints imposed on the system.

Figures 8 and 9 illustrate the model estimates for epochs 1915 and 1882 respectively. Comparing all the plots we see that patches 5 and 7 (Easter Island and St. Helena) remain fairly stationary throughout time, whereas patches 6 and 8 (W. and E. Pacific) are much more active, changing both their latitude and longitude throughout time. The two large negative regions of flux within patch 1 (N. Hemisphere) are approximately independent of longitude in all epoch models, although the region over Canada does drift down into North America as we track back in time. This effect was also found in the unconstrained, time dependent, model of Jackson *et al.* (2000). Finally, we note that, once again, the epoch models for 1915 and 1882, whilst fitting the data satisfactorily, conform to both the flux and vorticity constraints.

Table 2 summarises the characteristics of the model. We find that all the models have less complexity, measured by the roughness \mathcal{R} , than the reference model. This is precisely what we would hope, since none of the data sets have the resolving power of the reference model data set. This point is reinforced in Figure 10, which shows the Lowes magnetic energy spectrum of the field, as a function of spherical harmonic degree. None of the models with constraints require more energy than the 2000 reference model, and the imposition of the constraints does not affect their smooth decay of power with degree l .

5 Conclusions

The main result of the paper is that we find no insuperable difficulties in calculating models for the last century which are consistent with the frozen-flux hypothesis. This is encouraging, since theoretically we expect the core’s behaviour to be well approximated in this way. Recent dynamo calculations support this view: both Roberts & Glatzmaier (2000) and Rau *et al.* (2000) find that the frozen flux hypothesis is a useful one, though their approaches are different. Roberts & Glatzmaier (2000) analyse the unsigned flux, which is the sum of all the fluxes through the null-flux patches considered here, but without regard to sign; it is constant for a frozen-flux core. They find that over a 150 year timespan the change in the unsigned flux is only 3%, which they understandably consider to be small. Rau *et al.* (2000) test their ability to recover velocity fields from dynamo calculations under the frozen-flux approximation, and again find a fair degree of success in doing so. In these calculations the magnetic Reynolds number (the ratio of diffusive to advective

timescales) is between 100 and 300, possibly slightly low for the core.

Many attempts have been made to recover the fluid flow at the CMB from surface observations, for comparison with dynamo models, and for calculations of core angular momentum and core-mantle coupling. With occasional exceptions (e.g. Waddington *et al.*, 1995), these calculations are performed by using models of global geomagnetic main field and secular variation as the “data” for the inversion. Although diffusive theories exist, the vast majority of methods for inverting for fluid flow at the CMB are based on the frozen-flux hypothesis. Therefore we see the models we have developed herein as a logical step towards the self-consistent derivation of fluid motions (i.e. the fluid motions are based on models which are consistent with the assumptions required by the fluid motion inversion). In the future we will report on the calculation of time-dependent magnetic field models that satisfy the same constraints as in the present paper.

Acknowledgements

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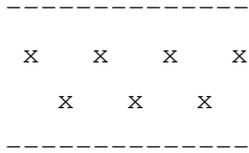
Appendix

Here we explain the rules for the division of nodes into the constrained and unconstrained sets, \mathcal{C} and \mathcal{U} respectively, introduced briefly in section 2.3. A certain null-flux patch has nodes interior $b_i^{(i)}$ (numbering $N^{(i)}$) and exterior $b_i^{(e)}$ (numbering $N^{(e)}$) to it — the most natural assignment of the nodes, which we implement, is that the set with the least number of nodes is interior. We only consider nodes adjacent to a null-flux curve (NFC), and assign them to the interior or exterior sets, also ranking them in terms of their proximity to the curve. We then work through this list, beginning with the node closest to the curve, and decide if it is allowed to be free to change sign (enter set \mathcal{U}) based on a set of rules which ensure that the topology of the curve is preserved: in particular we prevent a curve from disappearing altogether, or from breaking up into multiple patches. Rule 1, below, is the most important criterion; rules 2–6 are implemented to deal with patches containing only a few nodes, and are redundant if the density of nodes is increased in the model to a suitable level. The final check (last paragraph) to prevent merging of patches always needs to be performed.

Let the node under consideration for assignment to \mathcal{U} or \mathcal{C} be $b_i^{(i)}$. Let $n_j^{(i)}$ and $n_j^{(e)}$ be nearest neighbours which are interior and exterior respectively. The interior neighbours can be subdivided into boundary neighbours $n_j^{(b)}$ and hinterland neighbours $n_j^{(h)}$; a boundary neighbour is adjacent to a NFC, whereas a hinterland neighbour is not. Clearly the maximum number of boundary neighbours is five, because if a node has six boundary neighbours by definition it is a hinterland node (not adjacent to a NFC) and does not need considering.

The rules

1. If any $n_j^{(e)} \in \mathcal{U}$, set $b_i^{(i)} \in \mathcal{C}$; apply same logic also to external nodes $b_i^{(e)}$. This rule prevents the generation of extra patches as a result of two adjacent nodes which previously had opposite signs (NFC lies between them) swapping signs.
2. If $N^{(i)} = 1$ set $b_i^{(i)} \in \mathcal{C}$ (patch with only one node must be fixed to prevent patch disappearing)
3. If $N^{(i)} = 2$ and one node already in \mathcal{U} set $b_i^{(i)} \in \mathcal{C}$ (again, prevent patch disappearing)
4. If $N^{(i)} = 3$ and two neighbours already in \mathcal{U} set $b_i^{(i)} \in \mathcal{C}$ (all live on same triangle; three cannot be free or patch may disappear)
5. Check against boundary neighbours and hinterland neighbours: use Table 3.
6. Check interior free list to see if any interior neighbours are free. This can occur because the free nodes are not adjacent to the same portion of the NFC; schematically they are of the form (where --- is a NFC and \times represents a free node)



A final check now needs to be performed to prevent merging of different patches, which results from exterior nodes which are in \mathcal{U} being neighbours; if both changed sign the patches would merge. We implement this using multiple passes through the list of exterior nodes in \mathcal{U} . At each pass, for each node in the (exterior) free list of the two patches we check to see if there are either common nodes or adjacent nodes. If so, we discard the node which is furthest from the NFC from the free list, and then repeat the pass until no nodes in \mathcal{U} are adjacent and exterior. Note that we must treat the magnetic equator as special, because it is the only NFC which has nested NFCs; therefore we must check both its interior and exterior list of free nodes to ensure that there is a fixed node between any free node associated with it and any free node associated with another curve.

Figure Captions

Figure 1. Schematic diagram illustrating the conservation of flux and radial vorticity. In this diagram, negative flux is white and the intensity of positive flux is given by grey-shades; the total flux over the surface is zero. A null-flux curve is the line of zero flux, and a null-flux patch is enclosed by that curve. The frozen flux conditions allow a null-flux patch to shrink in size, provided that the flux linked by the patch is preserved. However, the imposition of Kelvin’s theorem means the only way that a patch can shrink is to change the latitude of at least part of the patch (bottom panel compared to top).

Figure 2. The tessellation of the sphere by spherical triangles, generated by bisections and trisections of the regular icosahedron on the unit sphere (Baumgardner and Fredrickson, 1985). The planar triangles are shown here for clarity, although the computations take into account the sphericity of the core by using spherical triangles.

Figure 3. The basis function associated with a 5-fold coordinated node, illustrated on a small section of a polyhedral rather than a spherical core. The vertical axis shows the amplitude of the basis function; it is zero everywhere apart from in the triangles adjacent to the node to which it is associated. The basis functions are linear “tent” functions, with value unity at the vertex and zero on the edges. The local support is the key to the efficient computations reported herein.

Figure 4. The assignment of nodes to the sets \mathcal{U} (unconstrained) and \mathcal{C} (constrained), the sets where signs of the nodes are either free to change sign or must remain of the current sign respectively. The assignments ensure that the topology of null-flux patches is retained. Shown here is an example with nodes which are free to change sign coloured black; all others must retain their current sign.

Figure 5. The reference model \mathbf{b}^* , epoch 2000, plotted on the core surface, radius $s = 3485km$. Red colours indicate flux into the core, blue contours are flux out of the core; the contour interval is $100\mu T$. Misfit = 1.01, $\mathcal{R} = 0.484\mu T/km$. Aitoff equal area projection.

Figure 6. The 1980 constrained field model, plotted on the core surface. Colours as in Figure 5. Misfit=0.994, $\mathcal{R} = 0.431\mu T/km$.

Figure 7. The 1945 constrained field model, plotted on the core surface. Colours as in Figure 5. Misfit=0.973, $\mathcal{R} = 0.337\mu T/km$.

Figure 8. The 1915 constrained field model, plotted on the core surface. Colours as in Figure 5. Misfit=0.976, $\mathcal{R} = 0.325\mu T/km$.

Figure 9. The 1882 constrained field model, plotted on the core surface. Colours as in Figure 5. Misfit=1.071, $\mathcal{R} = 0.370\mu T/km$.

Figure 10. Comparison of spherical harmonic energy spectra (Lowe spectra) for the unconstrained and constrained models.

Tables

Patch	Position	Epoch				
		2000	1980	1945	1915	1882
1	N. pole	53.77	53.77	53.77	53.82	53.77
2	N. hemisphere	-17566	-17565	-17573	-17350	-17561
3	Bone	-1317.5	-1317.9	-1316.3	-1310.6	-1316.4
4	S. hemisphere	18932	18933	18940	18709	18926
5	Easter Island	-21.66	-21.66	-21.84	-21.66	-21.66
6	W. Pacific	27.95	27.68	26.88	27.94	28.12
7	St. Helena	-143.51	-143.34	-143.42	-143.35	-143.55
8	E. Pacific	34.55	34.31	34.34	34.33	34.31

Table 1a. Fluxes, in μT , of the topologically constrained models. Epoch 2000 defines the reference model, \mathbf{b}^*

Patch	Position	Epoch				
		2000	1980	1945	1915	1882
1	N. pole	0.079	0.079	0.079	0.079	0.079
2	N. hemisphere	2.916	2.916	2.916	2.916	2.916
3	Bone	-0.559	-0.559	-0.559	-0.558	-0.557
4	S. hemisphere	-2.445	-2.445	-2.445	-2.445	-2.446
5	Easter Island	-0.017	-0.017	-0.017	-0.017	-0.017
6	W. Pacific	0.032	0.032	0.032	0.032	0.032
7	St. Helena	-0.032	-0.032	-0.032	-0.032	-0.032
8	E. Pacific	0.024	0.024	0.024	0.024	0.024

Table 1b. Vorticities, in steradians, of the topologically constrained models. Epoch 2000 defines the reference model, \mathbf{b}^*

Model	Misfit \mathcal{M}	Roughness \mathcal{R}
2000	1.010	0.484 $\mu T/km$
1980	0.994	0.431 $\mu T/km$
1945	0.973	0.337 $\mu T/km$
1915	0.976	0.325 $\mu T/km$
1882	1.071	0.370 $\mu T/km$

Table 2. Comparison of model misfits and roughnesses.

$\#n_j^{(b)}$	$\#n_j^{(h)}$	Assignment
1	0	\mathcal{U}
2	0	\mathcal{C}
2	1,2	\mathcal{U}
3,4,5	0	\mathcal{C}

Table 3. Rules for assigning a node to \mathcal{U} or \mathcal{C} . $\#n_j^{(b)}$ is the number of boundary neighbours, and $\#n_j^{(h)}$ the number of hinterland neighbours.

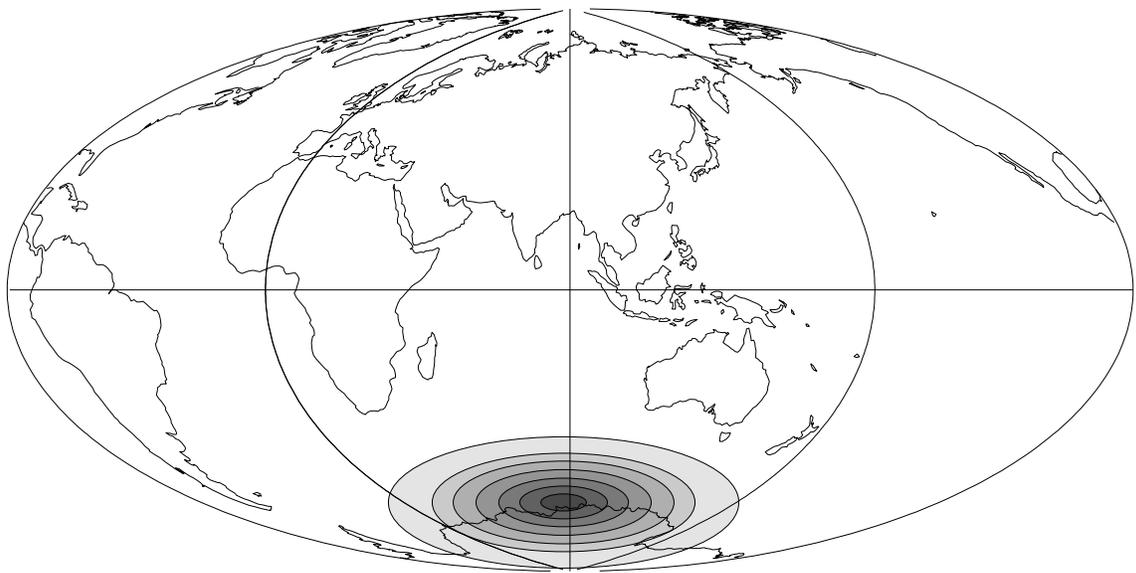
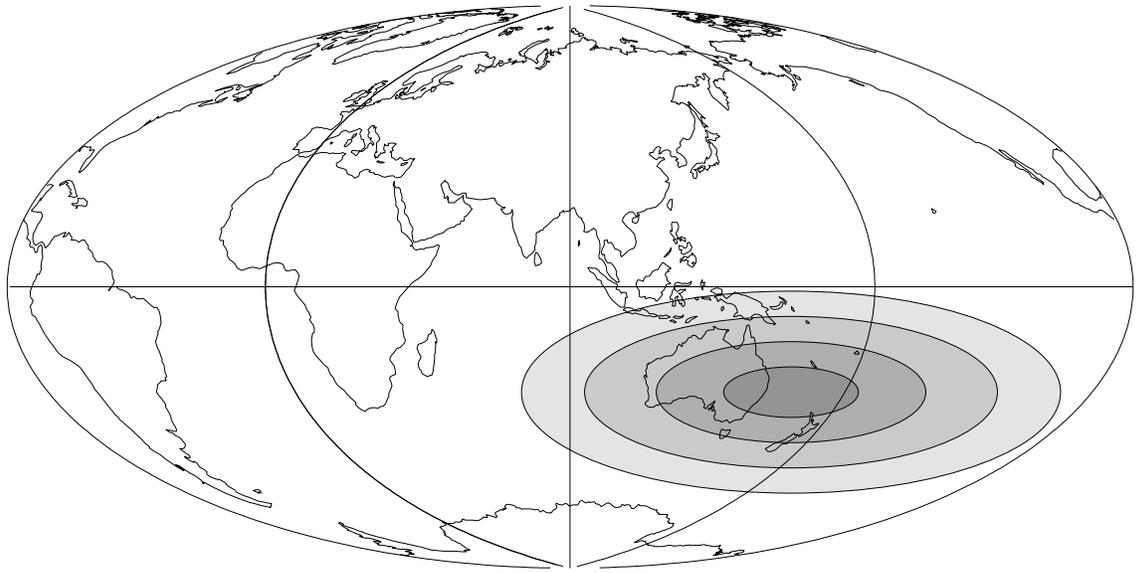


Fig. 1

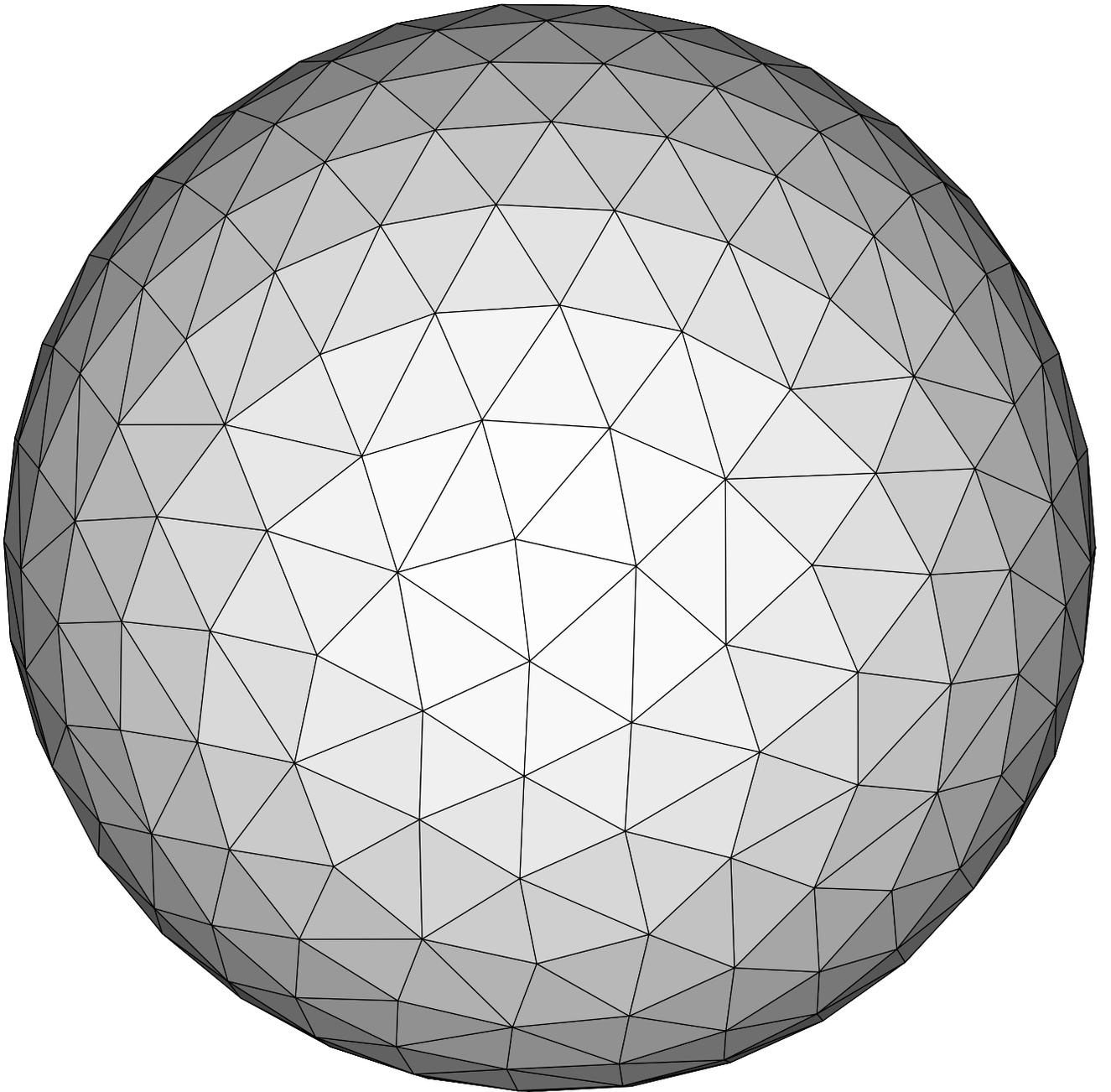


Fig. 2

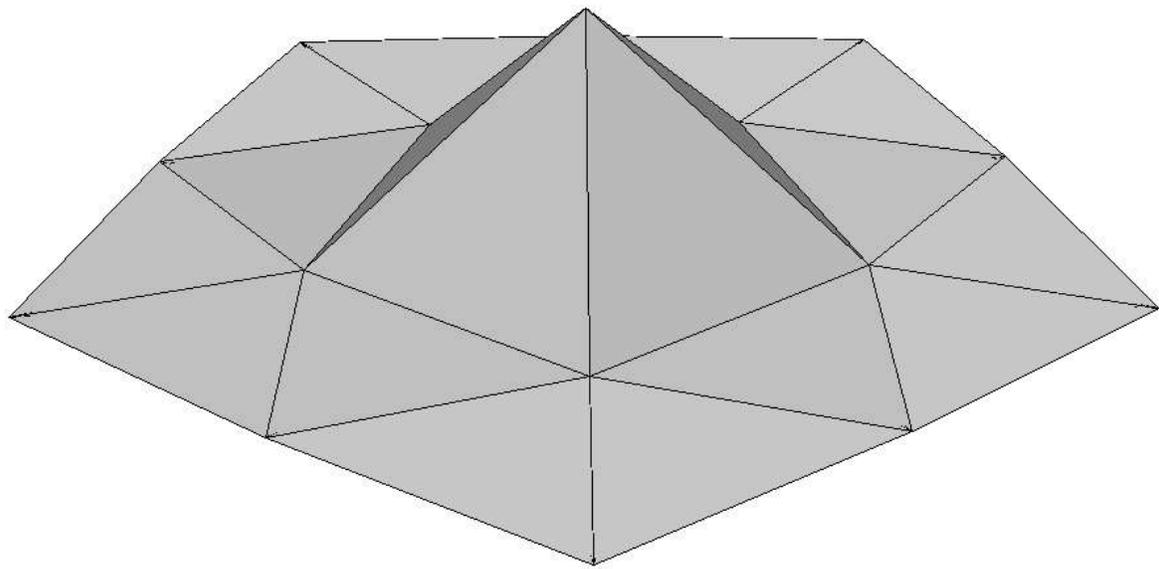


Fig. 3

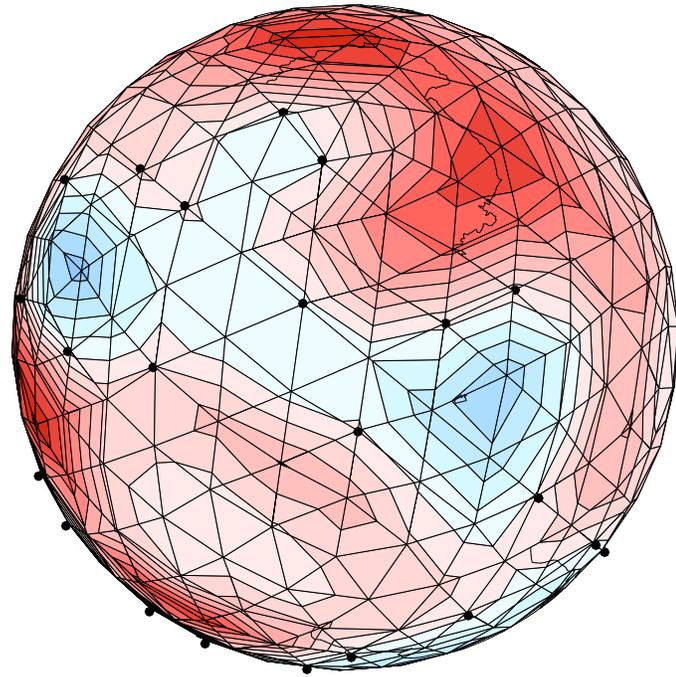
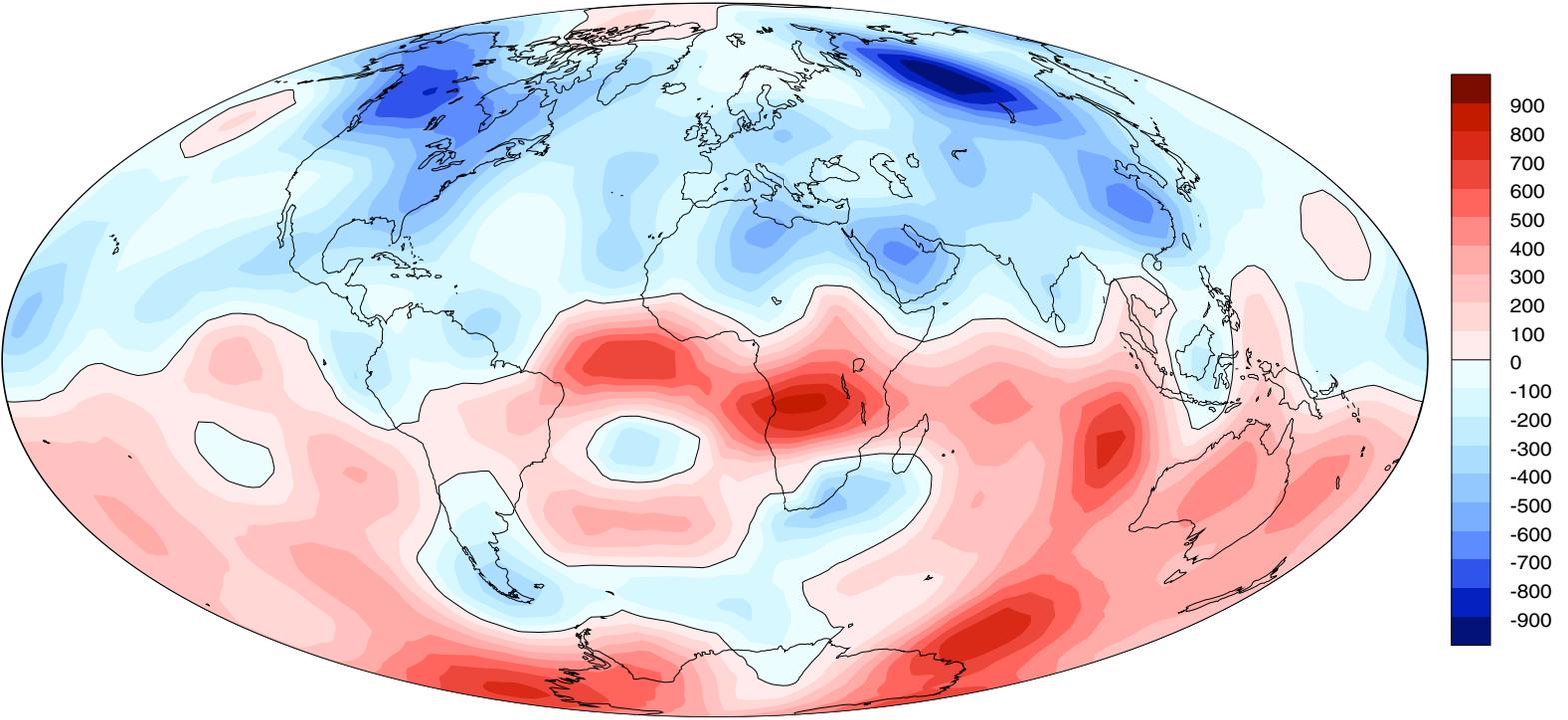
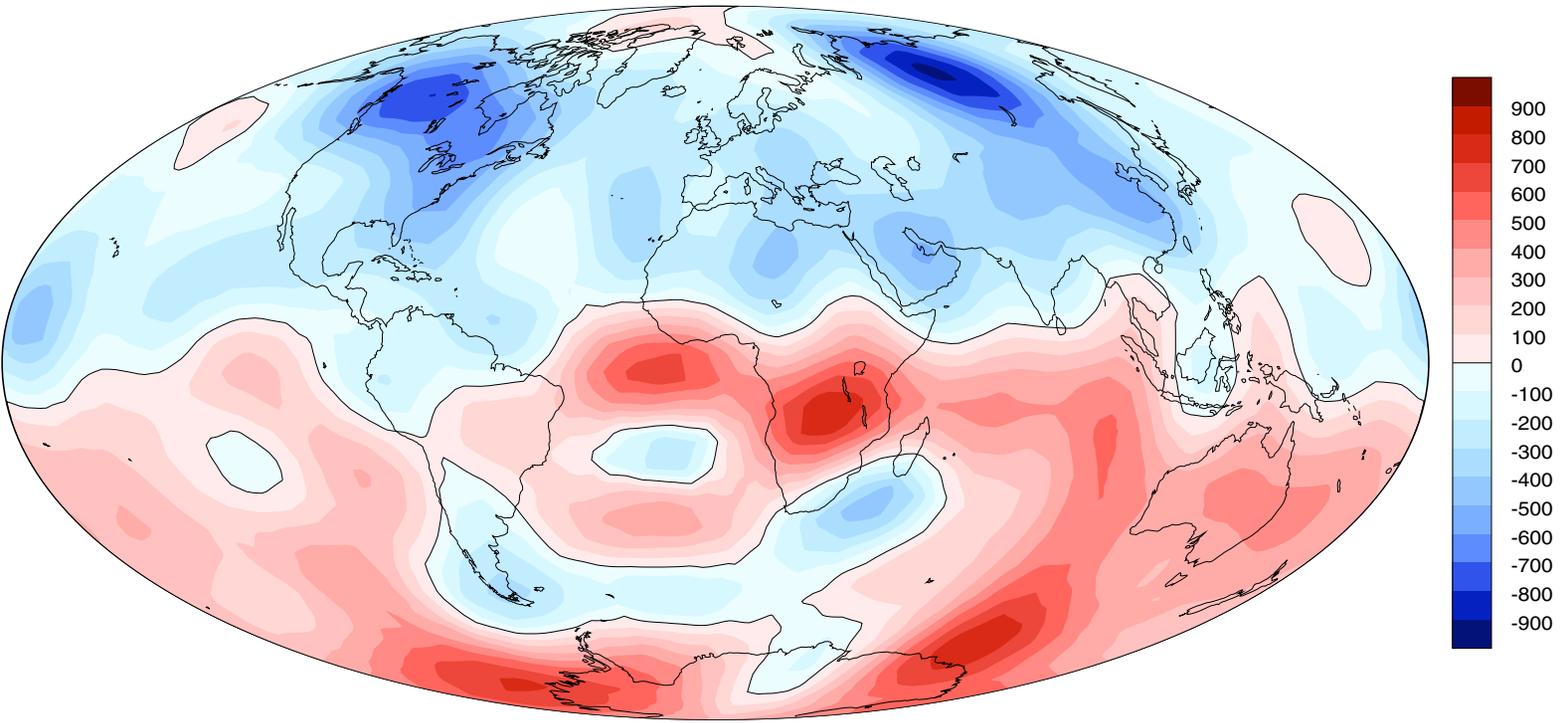


Fig. 4



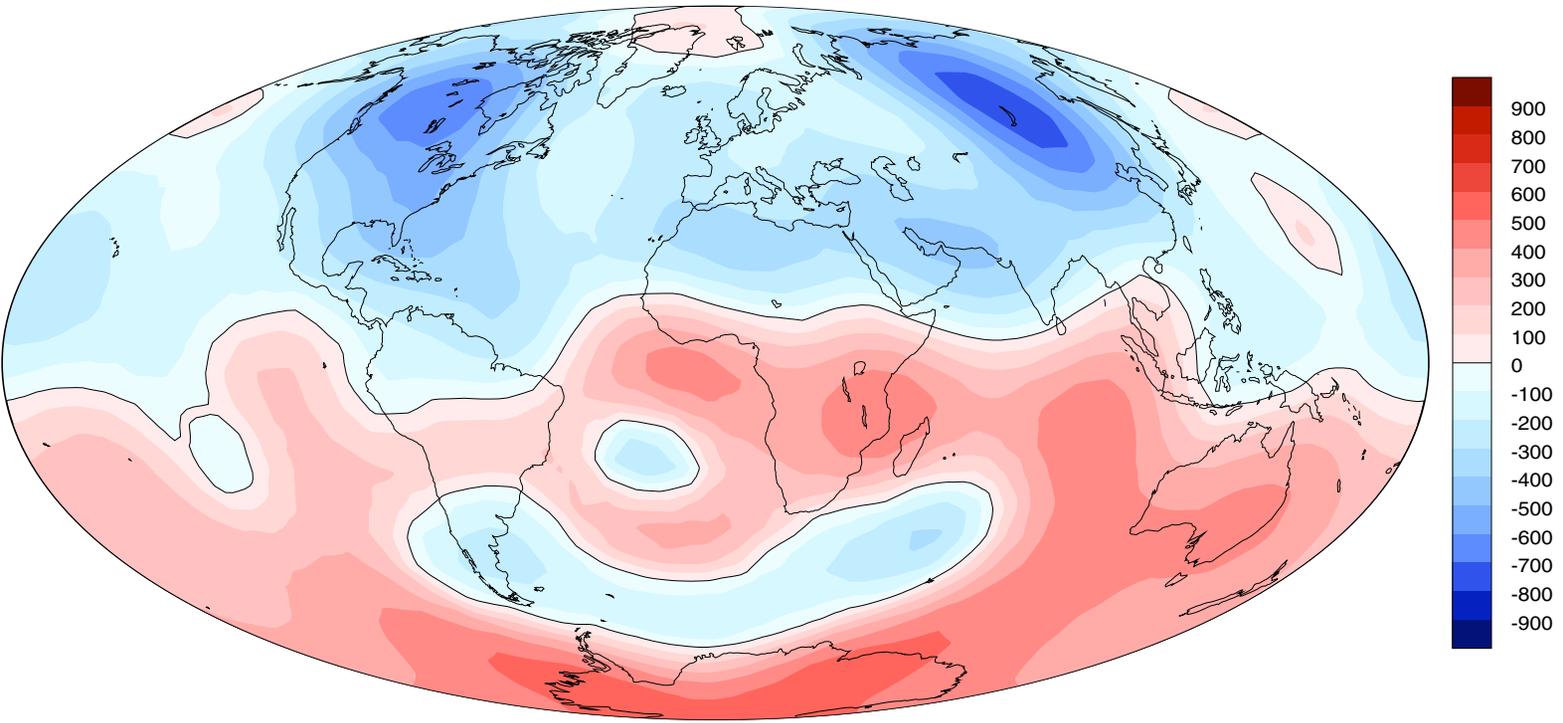
2000 Reference Model

Fig. 5



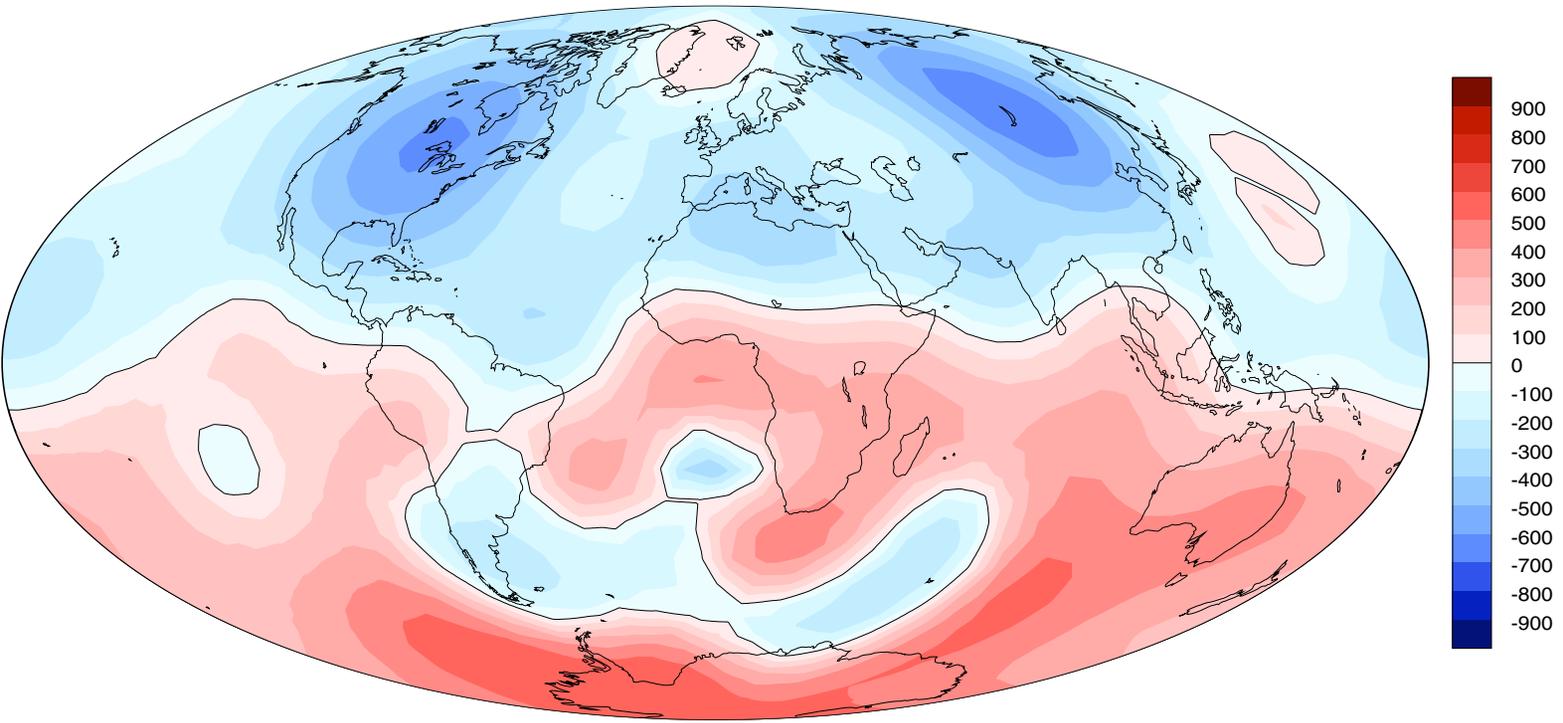
1980

Fig. 6



1945

Fig. 7



1915

Fig. 8

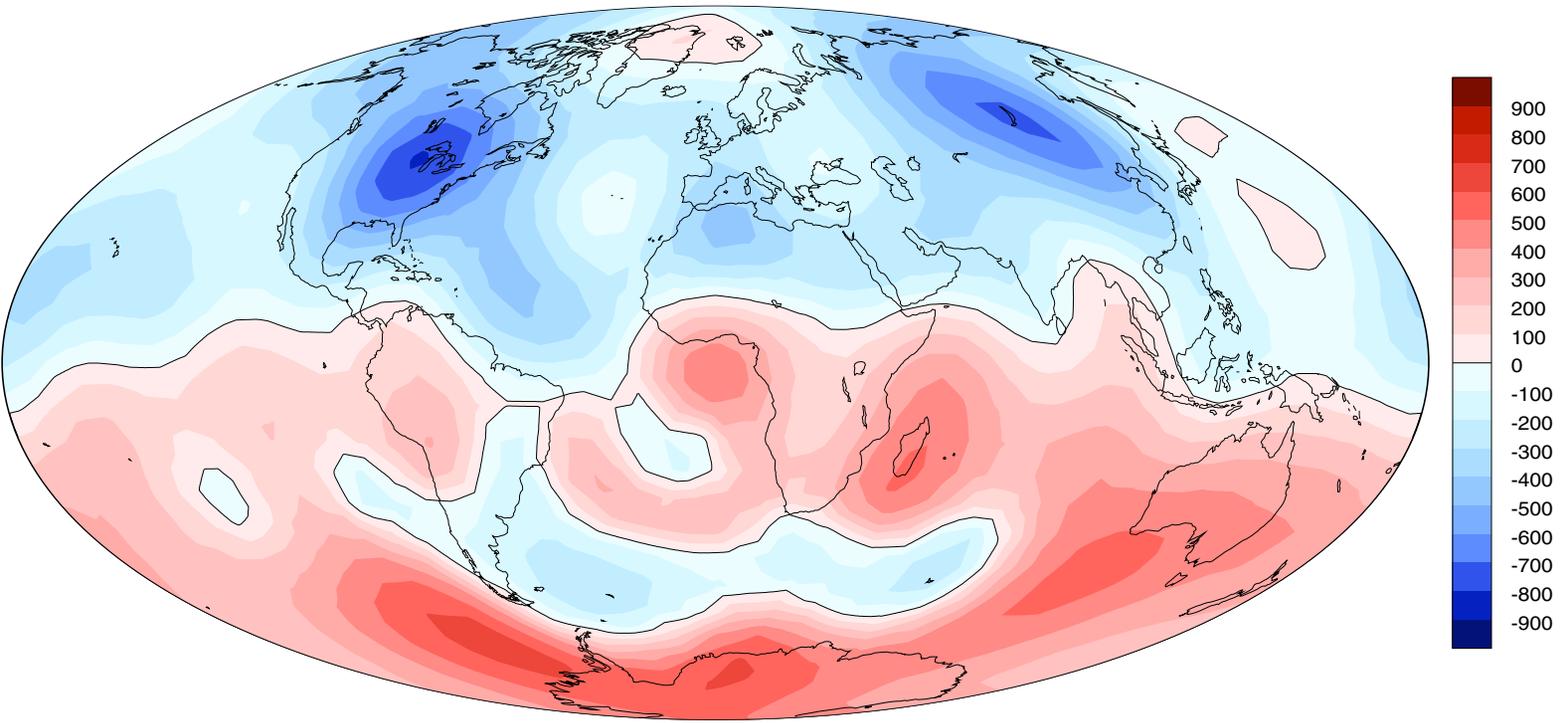


Fig. 9

1882

Spherical Harmonic Power Spectra of Magnetic Field Models

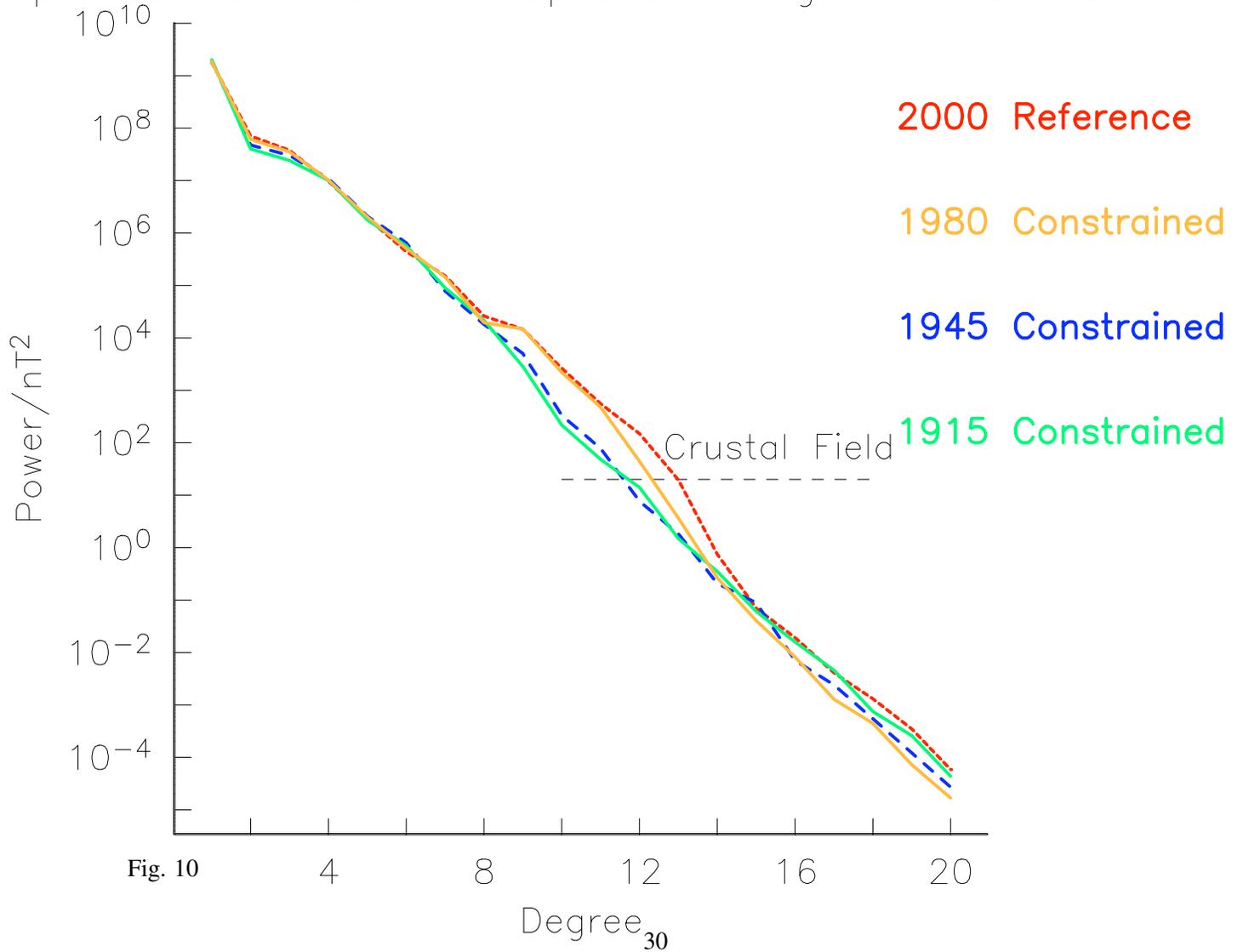


Fig. 10