

1 **MCADAM: A continuous paleomagnetic dipole**
2 **moment model for at least 3.7 billion years**

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6 **Key Points:**

- 7 • Continuous dipole moment models for the past 3.7-4.2 billion years are presented
- 8 • Our model reproduces salient features of the paleomagnetic dipole field
- 9 • Paleomagnetosphere estimates suggest Precambrian atmospheric shielding was much
- 10 weaker than present day
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Abstract

Understanding the evolution of Earth's magnetic field can provide insights into core processes and can constrain plate tectonics and atmospheric shielding. The PINT absolute paleointensity database provides a curated repository of site mean, i.e., cooling unit, estimates of the strength of the magnetic field. We present a minor update to the PINT database to version 8.1 by adding 199 records from 25 studies published primarily from 2019 through 2022. The PINT database is used to define a continuous model of the dipole field, using an approach combining non-parametric and Monte Carlo resampling termed MCADAM. Three dipole field strength models spanning 50 ka to 3.7-4.2 Ga (MCADAM.1a-c) are presented, reflecting three tiers of increasingly more stringent data selection thresholds. The MCADAM dipole field models allow for the estimation of the magnetic stand-off distance, constraining the shielding of Earth's atmosphere against solar wind erosion provided by the geodynamo.

1 Introduction

The evolution of Earth's deep interior since core formation (Nimmo, 2015) > 4 billion years ago (Ga) remains a topic of considerable study. Obtaining information of the deep interior is generally restricted to present-day observations (e.g., seismic tomography). Alternatively, insights on processes occurring before the modern era require sampling geologic materials which formed at, or were transported to, Earth's surface. However, the geomagnetic field is generated in the liquid fraction of Earth's core through the geodynamo, and changes in the morphology, strength and variability in the geodynamo may reflect the evolution of core processes and the pattern of heat flux at the core-mantle boundary (CMB). The geomagnetic field is also a critical component for Earth's habitability (Rodríguez-Mozos & Moya, 2017) due to the protective envelope provided by the magnetosphere against atmospheric erosion by charged solar particles. It is speculated that changes in the paleomagnetosphere may have contributed to substantial changes in the evolution of life (e.g., Meert et al., 2016).

Paleomagnetic studies offer the potential to help close this gap: when rocks bearing magnetic carriers form the geomagnetic field imparts a remanence magnetization that, with ideal conditions and carriers, can be robustly preserved on the order of billions of years. The strength of the geodynamo can be described by the magnitude of the dipole moment, the first-degree spherical harmonic component of the field, which should reflect

44 a long-term (time-averaged) trend and describes $\sim 90\%$ of the recent geomagnetic field.
 45 A fundamental question regarding Earth’s dynamo is how the dipole moment has changed
 46 over long timescales (\gg millions of years). Individual paleomagnetic specimens can be
 47 measured in the laboratory to quantify the strength of the remanent magnetization, or
 48 paleointensity, imparted into the specimen. Paleointensities measured from the same ge-
 49 ologic time (e.g., from the same cooling-unit, referred to as a “site”) can be related to
 50 paleointensities from other locations by transforming the paleointensity (B) into a vir-
 51 tual (axial) dipole moment (V(A)DM) using the following equation (Merrill et al., 1996)

$$VDM = \frac{4\pi r_E^3}{2\mu_0} B(1 + 3\cos^2 I)^{0.5} \quad (1)$$

52 where r_E is Earth’s radius, μ_0 is vacuum permeability, and I is the inclination of the site
 53 derived from paleomagnetic directional measurements (there is an equivalent transfor-
 54 mation to VADM using site paleolatitude). Virtual dipole moment transformations as-
 55 sert that the mean paleointensity measured at the site level can be entirely described by
 56 the dipole field, this simplification allows for comparisons from globally distributed ob-
 57 servations on the strength of the field.

58 Characterizing the time-varying paleomagnetic field can be approached using sev-
 59 eral different methods. On geologically recent timescales (< 100 thousand years, kyr),
 60 spherical harmonic models describe the morphology and strength of the field (e.g., Panovska
 61 et al., 2018). For the past 2 Myr, a continuous axial dipole moment model (Ziegler et
 62 al., 2011) can be constructed using relative paleointensity data from stacked sedimen-
 63 tary records combined with absolute paleointensity estimates, generally from volcanic
 64 sources. For longer timescales ($\gg 2$ million years), dipole moment descriptions are sub-
 65 stantially less well resolved. Tauxe and Staudigel (2004) reported a mean value for the
 66 0-300 Ma interval, whereas Ingham et al. (2014) and Kulakov et al. (2019) applied a more
 67 complex reversible-jump Markov Chain Monte Carlo approach to define Mesozoic trends.
 68 Other approaches recently applied to the Precambrian field include binned data (e.g.,
 69 Biggin et al., 2015), a low-degree polynomial fit (e.g., Bono et al., 2019), or sliding win-
 70 dow average (e.g., Tarduno et al., 2020). These meta-analyses have proven important
 71 in providing observational constraints on dynamo and core evolution models (e.g., Big-
 72 ggin et al., 2015; Driscoll, 2016; Bono et al., 2019) and time-averaged and time-varying
 73 field estimates (e.g., Selkin & Tauxe, 2000; Ziegler et al., 2011).

74 The PINT database (<http://www.pintdb.org/>; Biggin et al., 2009; Bono et al.,
 75 2022) is a curated repository of absolute paleointensity records derived from volcanic sources
 76 and reported at the site mean level with associated meta-data, which makes it well-suited
 77 for paleointensity meta-analyses. In this study, we provide an incremental update to the
 78 PINT database (v8.1) that we use as the basis for a dipole moment evolution model (Sec-
 79 tion 2). In Section 3, we introduce a modeling framework, MCADAM (Monte Carlo Ax-
 80 ial Dipole Average Model), that uses a combination of non-parametric site resampling,
 81 Monte Carlo simulations, and time-adaptive locally-weighted smoothing to produce a
 82 posterior distribution of field strength estimates from which a median dipole strength
 83 and associated predictive interval can be determined. Using the MCADAM framework
 84 and *three* filtered datasets from the PINT database that apply increasingly more strin-
 85 gent selection criteria, we present a suite of dipole moment evolution models that yield
 86 continuous predictions of the time-average (paleomagnetic) dipole moment extending back
 87 to the oldest paleomagnetic records from > 4 Ga, and compare these models with other
 88 time-average descriptions of field strength in deep time (Section 4) and the associated
 89 impact on the paleomagnetosphere (Section 5).

90 2 Updates to PINT v8.1

91 The PINT database underwent a significant update to version 8.0, and we refer read-
 92 ers to Bono et al. (2022) who describe the current structure of the database and broadly
 93 summarizes the distribution and quality of the paleointensity dataset. The most salient
 94 changes in PINT v8.0 with respect to prior versions of the PINT database (Biggin et al.,
 95 2015) are the inclusion of new paleointensity data published through the end of 2019,
 96 the removal of demonstrably biased paleointensity records (so-called “auto-zeros”), and
 97 the integration of Q_{PI} assessments for over 90% of the database. Q_{PI} (Quality of Pa-
 98 leointensity; Biggin & Paterson, 2014) is a semi-quantitative framework to describe the
 99 reliability of a site mean paleointensity record, and we again refer readers to Bono et al.
 100 (2022) for a complete description of Q_{PI} implementation in PINT v8.0.

101 In this study, we include a minor version update of PINT to v8.1 (1 that includes
 102 paleointensity records published in 2020 through July 2022. Included studies are not ex-
 103 haustive of entire paleointensity dataset published during this interval, however, it rep-
 104 represents a good-faith effort to identify as many relevant studies as possible. In total, **230**
 105 new sites from **29** studies have been added to the PINT v8.1 database. These data in-

106 clude contributions constraining the field during the Cambrian/Ediacaran (e.g., Thallner,
 107 ner, Biggin, & Halls, 2021; Thallner, Biggin, McCausland, & Fu, 2021) and Neoproterozoic
 108 zoic (e.g., Lloyd et al., 2021), which remain under-sampled relative to other geologic in-
 109 tervals.

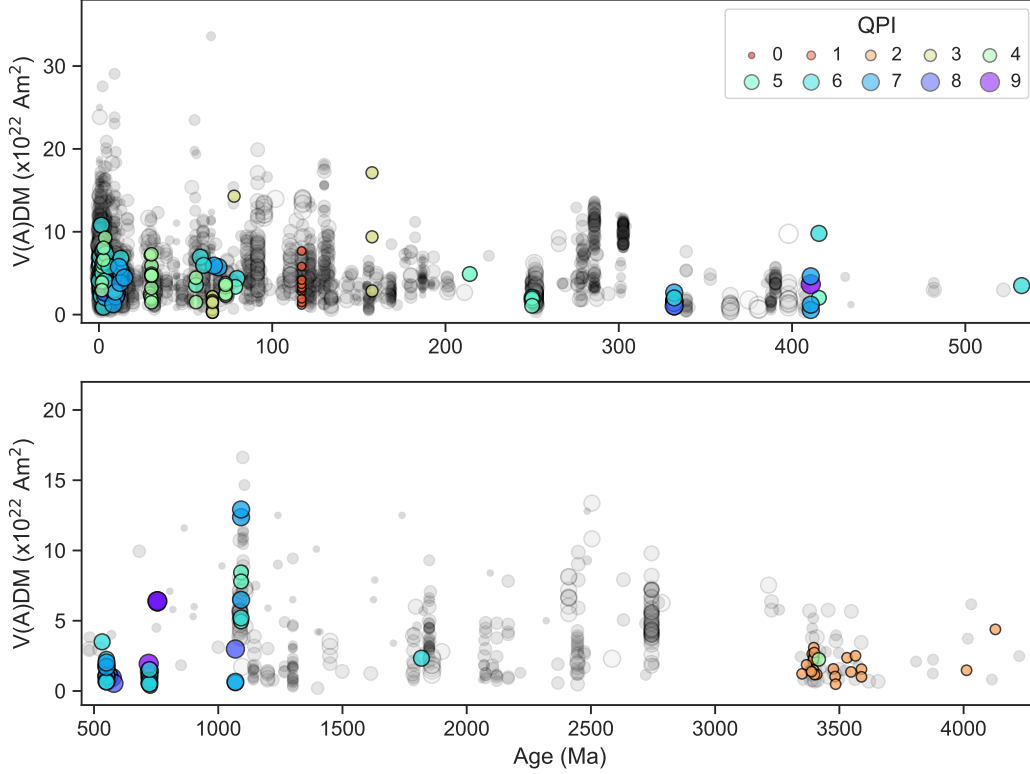


Figure 1. PINT v8.1 absolute paleointensity database. Colored circles show site mean records added since v8.0 (Bono et al., 2022); grey circles are data in PINT v8.0. Symbol size and color shows Q_{PI} score. Top: Phanerozoic; bottom: Precambrian.

110 The Q_{PI} criteria (Biggin & Paterson, 2014) serve two main purposes: first to act
 111 as a framework in the design and implementation of paleointensity research. Each in-
 112 dividual criterion represents a pillar of best practice in paleointensity experimental de-
 113 sign. While satisfying any or all the individual criterion does not guarantee that a given
 114 site mean paleointensity result reflects the true field strength, our confidence in the data
 115 should improve with increasing Q_{PI} criteria achievement (i.e., the data take into account
 116 key factors that are known to detrimentally impact the fidelity of the paleointensity record-
 117 ing). A second purpose of the Q_{PI} criteria is to facilitate the meta-analysis of the PINT
 118 database and site mean paleointensity-data. In particular, to enable more nuanced anal-

119 ysis outside of the paleomagnetic community, where the familiarity and historical con-
120 text of how paleointensity data are collected and reported is less understood. Q_{PI} cri-
121 teria allow for a semi-quantitative, objective definition of requirements to filter data from
122 the PINT database, with the goal of improving the robustness of meta-analyses.

123 Field strength estimates are inherently challenging to extract from the rock record.
124 Paleointensity specimens may be compromised by the presence of non-ideal magnetic recorders
125 (e.g., multidomain grains) and/or laboratory alteration. The potential for remanences
126 to be reset by thermal or chemical over-printing after emplacement must also be excluded
127 before accepting a measured paleointensity as valid and meaningfully linked to the em-
128 placement age. Several efforts have been made to identify useful heuristics and minimum
129 analytical requirements to separate robust individual paleointensity results from suspect
130 ones (e.g., Selkin & Tauxe, 2000; Kissel & Laj, 2004; Biggin et al., 2007). However, there
131 is no clear consensus on what should be the minimum acceptable thresholds for paleoin-
132 tensity data at the specimen level, and propagating those decisions through to the site
133 mean level often requires substantial discipline expertise and specific geologic context
134 for the locality. Thus, applying a consistent data treatment in meta-analyses is often not
135 feasible. While the Q_{PI} criteria do not reflect all aspects of paleointensity data (see Smirnov
136 et al. (2016)), by providing individual Q_{PI} criteria and descriptive notes explaining the
137 scoring process, the framework allows for informed decision making in the data selection
138 process.

139 Since the data may reflect some non-ideal paleointensity biases, some fraction of
140 the site mean data should be excluded from analyses in order to improve the robustness
141 of any resulting conclusions drawn from using the PINT database. However, paleointen-
142 sity data are sparse and imperfect individual records may still yield meaningful infer-
143 ences in aggregate. Thus it is crucial to define selection criteria that balance data qual-
144 ity with data availability, specifically for the development of time-averaged and time-evolution
145 field descriptions on million-to-billion-year timescales. Meta-analyses considering other
146 topics will, of course, result in different optimal selection criteria choice.

147 Three different selection criteria are employed as examples for how the PINT database
148 can be interrogated for meta-analysis, which have been previously presented in Bono et
149 al. (2022). The first two filters are as follows: all data, and site mean $Q_{PI} \geq 3$. The third
150 filter (introduced by Kulakov et al., 2019) prioritizes certain criteria and requires that

151 site mean data meet the QAGE, QALT, and QMD criteria, thus requiring evidence that
 152 the site age is well constrained and the primary remanence is associated with the age
 153 estimate (QAGE) and there were experimental controls to limit the influence of labo-
 154 ratory alteration (QALT) and non-ideal (i.e., multidomain) magnetic carriers (QMD).
 155 We note that Smirnov et al. (2016) and Bono et al. (2019) previously identified PINT
 156 data which potentially under (over) estimate field strength fitting the shallow (steep) com-
 157 ponents of two-slope or concave Arai diagrams. Since this level of analysis was not ap-
 158 plied to all records within PINT v8.1, we have not excluded the identified sites *a pri-*
 159 *ori*, however, we distinguish sites which may be biased in Figure 2b and all but two sites
 160 are excluded using our “strict” prioritized Q_{PI} selection criteria. In addition to the above
 161 selection criteria, sites explicitly described as having a transitional polarity were excluded.

162 **3 Time-varying paleofield models with uncertainties**

163 Here, we consider whether a continuous time-varying field model can be realized
 164 for the entire paleointensity record. Ideally, a time-varying dipole field strength model
 165 should take several factors into consideration. We chose to focus on the following require-
 166 ments:

- 167 1. Data selection should balance quality with availability of data.
- 168 2. The model should not be overly sensitive to any given data point due to the sparse
 169 and non-uniform distribution of paleointensity site means in the PINT database.
- 170 3. The model should reflect the uncertainty of individual site mean estimates in both
 171 age and field strength.
- 172 4. The model should seek to average secular variation, taking into account the in-
 173 creasing sparsity of data going further back into geologic time.

174 To meet these requirements, we employ a combination of techniques, which we re-
 175 fer to as a Monte Carlo Axial Dipole Average Model (MCADAM). A complete descrip-
 176 tion of the modelling approach is provided in Supplementary Text S1. In summary, a
 177 resampling algorithm using non-parametric resampling of site mean records combined
 178 with Monte Carlo realizations derived from site mean intensity, inclination and age (and
 179 associated uncertainties). An individual set of realizations are time-averaged using a LOWESS
 180 (local regression) approach (Cleveland, 1979), applying an adaptive kernel defined used
 181 a minimum number of sites (5) and reasonable age bounds (spanning 250 kyr to ~ 76

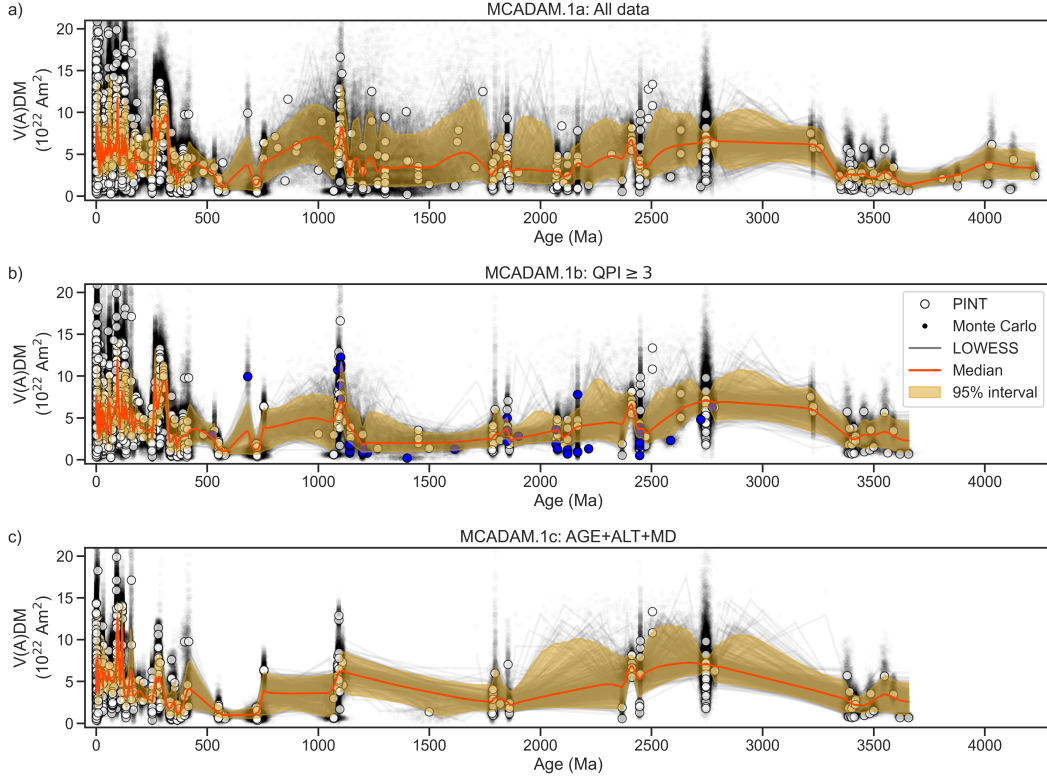


Figure 2. MCADAM time-varying model of paleofield strength for the past 3.7 to 4.2 billion years from PINT data. White circles: selected PINT data; black points, Monte Carlo realizations of PINT data; grey lines, individual MCADAM realizations; orange line, median time-varying model from MCADAM with shaded 95% interval. A) MCADAM.1a, all data in PINT v8.1; B) MCADAM.1b, $Q_{PI} \geq 3$, blue circles mark sites which may be biased as identified by Smirnov et al. (2016) or Bono et al. (2019); C) MCADAM.1c, prioritized Q_{PI} : QAGE, QALT and QMD must be equal to 1.

182 Myr). The MCADAM modeling framework was tested using a synthetic data set with
 183 a known “true” dipole moment and a temporal distribution derived from PINT v8.1 (Sup-
 184 plementary Fig. S1).

185 **4 Comparing MCADAM to other compilations**

186 Applying the MCADAM approach with the PINT v8.1 dataset restricted by the
 187 three selection filters previously discussed (all data, $Q_{PI} \geq 3$, and prioritized selection
 188 QAGE+QALT+QMD), the resulting time-varying models (MCADAM.1a-c) are presented
 189 in Figure 2. Our preferred model is MCADAM.1b, which uses the a moderately restric-

190 tive dataset requiring that paleointensity site records meet at least three of the Q_{PI} cri-
 191 teria. In general, this model reproduces several characteristic features previously observed
 192 in the paleofield (Figure 3 and Supplementary Figures S2-S3), such as rise in field strength
 193 from the Matuyama to Brunhes chrons, intervals of high field strength during the Cre-
 194 taceous Normal Superchron preceded by a weaker field (cf. the binned PINT analysis
 195 of Kulakov et al. (2019)), and a high field during the Kiaman Superchron (e.g., Cottrell
 196 et al., 2008) preceded by sustained weak field during the Devonian (Hawkins et al., 2019).
 197 For the 50 kyr to 2 Ma interval, there is good agreement between our model and that
 198 of PADM2M (Ziegler et al., 2011). Given the denser temporal sampling during the Phanero-
 199 zoic, more variation in the field can be resolved with a smaller confidence interval for the
 200 resulting model relative to the Precambrian.

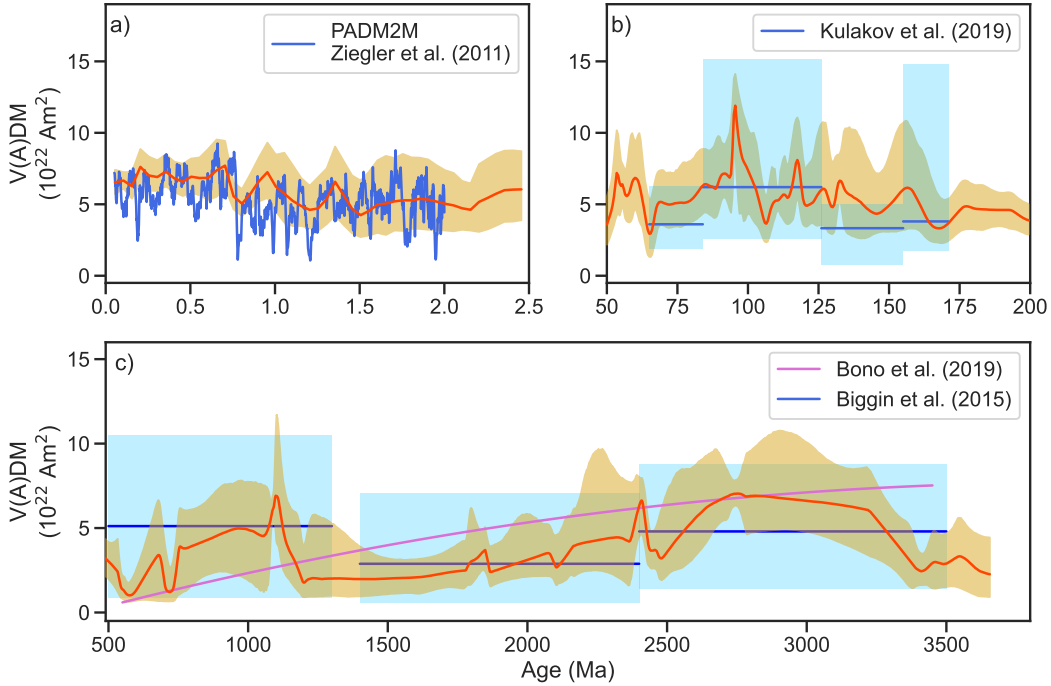


Figure 3. MCADAM.1b time-varying model of paleofield strength for the past 3.7 billion years from PINT v8.1 data meeting $Q_{PI} \geq 3$ criteria. In all panels, the orange line represents the median time-varying model from MCADAM.1b with shaded 95% interval. A) Quaternary; blue line shows PADM2M model (Ziegler et al., 2011); b) Mesozoic; blue line and field shows median and 95% interval of Q_{PI} binned following (Kulakov et al., 2019); C) Precambrian; purple line shows polynomial fit of Bono et al. (2019), blue lines show bin medians with shaded 95% confidence intervals of Biggin et al. (2015).

201 The Paleozoic through the Precambrian poses the greatest challenge for charac-
 202 terizing the time-varying field due to large gaps in the PINT database. In our model,
 203 we use a linear interpolation between sampling, however given that intervals spanning
 204 ~ 100 Myr may not sample the field at all, it is almost certain there are field variations
 205 that are not captured in our model. Given the combination of non-parametric resam-
 206 pling for data selection and using a Monte Carlo sampler based the selected data in the
 207 realization, we feel that the MCADAM models represent an overly smoothed description
 208 of the time varying field, particularly where the data are are sparse. We note that the
 209 oldest field records of the Archean are dominated by the Thellier-Coe zircon experiments
 210 of Tarduno et al. (2015, 2020), which due to their lack of orientation, represent a source
 211 of uncertainty in our model during the Eoarchean/Hadean. The fall and rise in field strength
 212 during the Mid- to Late- Proterozoic (as suggested by Biggin et al. (2015)) is supported
 213 by our model, as well as the drop in field strength at the end of the Proterozoic reported
 214 in Bono et al. (2019).

215 There are some general differences in the analyses of Biggin et al. (2015), Bono et
 216 al. (2019) and our study that can explain the apparent disagreement in estimated field
 217 trends. First, there are differences in the data sets used between both analyses, as sum-
 218 marized by Bono et al. (2019). Second, Biggin et al. (2015) divided the data sets into
 219 Early, Mid and Late Proterozoic bins and summarized the statistical properties of PINT
 220 records within each bin. Bono et al. (2019) focused *a priori* on estimates from slow-cooling
 221 intrusives (or select sites demonstrating time-averaged statistics) resulting in a substan-
 222 tially reduced data set compared to either this study or that of Biggin et al. (2015), and
 223 from this restricted data set estimated a 2^{nd} degree polynomial fit. In this study, we forgo
 224 both dividing the data into prescribed bins or focusing *a priori* on intrinsically time-averaged
 225 records. Our study uses a broader dataset, supplemented by new data published since
 226 the prior studies, that result in more variation in the interpreted dipole field strength
 227 relative to prior work.

228 **5 Implications for the paleomagnetosphere**

229 The geodynamo and the associated magnetic field extending into space provides
 230 shielding of Earth's atmosphere and surface water from erosion due to solar wind (Tarduno
 231 et al., 2014). In addition to increasing erosion of the atmosphere, reductions in magnetic
 232 shielding can drive breakdown of atmospheric ozone, which limits penetration of UVB

233 radiation (Glassmeier & Vogt, 2010). Because of the protective ability of magnetospheres,
 234 a long-lived, robust magnetic field has been identified as one of the prerequisites for a
 235 habitable planet (Rodríguez-Mozos & Moya, 2017). Therefore, the strength and evolu-
 236 tion of the magnetosphere is of critical interest. Currently, modelling the magnetosphere
 237 (or paleomagnetosphere) in detail requires fully coupled dynamo and solar activity sim-
 238 ulations beyond the scope of what is available. However, a first-order approximation can
 239 be estimated using a series of reasonable simplifications, chiefly that the field is dipole-
 240 dominated (supported by Biggin et al., 2020) and that magnetic shielding can be approx-
 241 imated by the magnetic standoff distance, or magnetopause, where solar wind pressure
 242 is balanced by the repelling force of a dipole field (Siscoe & Sibeck, 1980). The present-
 243 day magnetopause is $\sim 10 R_E$ (Earth radii) and will fluctuate on annual timescales as
 244 the magnetic pole moves about the spin axis (Shue et al., 1997).

245 Following the approach of Tarduno et al. (2010), the magnetic standoff distance,
 246 $r_s(t)$ for a given time t , can be estimated as the balance point between solar wind pres-
 247 sure and the dipole magnetic field (described by Siscoe & Chen, 1975),

$$r_s(t) = \left[\frac{\mu_0^2 f_0^2 M_E(t)^2}{4\pi^2 (2\mu_0 P_{SW}(t) + B_{IMF}^2)} \right] \quad (2)$$

248 where μ_0 is vacuum permeability, f_0 is a field shape parameter for the magnetosphere
 249 (1.16 for present day Earth, Voigt (1995), held constant here), and B_{IMF} is the inter-
 250 planetary field (which is neglected in our calculations since it is small, $\ll 10$ nT). M_E
 251 is the (paleo)magnetic dipole moment as a function of time. P_{SW} is the solar wind ram
 252 pressure, which is dependent on the mass loss rate of the sun and velocity of solar wind
 253 as a function of time. Extrapolating present day P_{SW} (~ 1.9 nPa, Shue et al. (1997)) back
 254 through time can be done with power-law model $(t/t_0)^{-2.33}$ based on solar analogs (e.g.,
 255 Wood et al., 2005), at least until the young Hadean sun.

256 Using the MCADAM.1b model, the magnetic standoff distance from 50 ka to 3.7
 257 Ga can be estimated (Figure 4 and Supplementary Figures S4-S5). The magnetopause
 258 responds rapidly to changes in either solar wind activity or the geomagnetic field and
 259 will vary by 1-2 R_E during typical space weather (Voigt, 1995). Coronal mass ejections
 260 and solar flares can suppress the standoff distance by half (e.g., the Halloween 2003 event
 261 was observed to reduce the magnetopause to $\sim 5 R_E$, Rosenqvist et al. (2005)). While
 262 short term reductions (\ll millions of years) in magnetic shielding are unlikely to impact
 263 the biosphere significantly, protracted intervals of reduced shielding may have affected

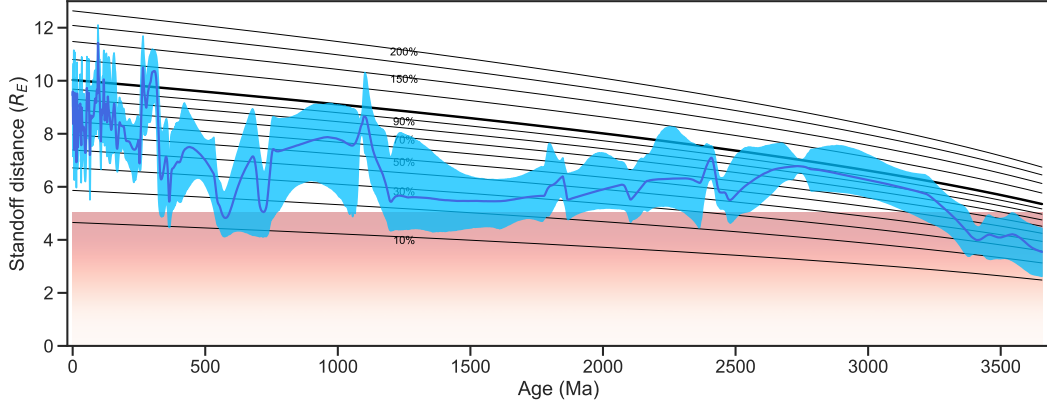


Figure 4. Magnetopause standoff distance estimate using equation 2 and the MCADAM.1b modeled dipole moment curve with PINT v8.1 data meeting $Q_{PI} \geq 3$ criteria. Blue curve is the predicted median dipole moment and blue field is the 95% predicted interval. Contour lines show standoff distance relative to the present day. Red gradient shows standoff distance associated with the Halloween 2003 solar storm (Rosenqvist et al., 2005).

264 evolutionary processes (e.g., Meert et al., 2016; van der Boon et al., 2022). Our analy-
 265 sis suggests that the combination of the generally weaker Precambrian field and the in-
 266 creased solar wind associated with a younger, more active sun resulted in a long-term
 267 average standoff of $\sim 5 R_E$, which is about half the present-day magnetopause and con-
 268 sistent with the single zircon crystal estimates of Tarduno et al. (2010) for the early Archean.
 269 Individual snapshots or time-average estimates (on million-year or shorter timescales)
 270 suggest there were intervals with even further reduced standoff distances, e.g., the Edi-
 271 acaran. These values represent a baseline standoff distance, which could be further re-
 272 duced due to internal changes in the field (reduction or loss of dipolarity) or increases
 273 in solar wind activity (coronal mass ejections, solar flares). This implies that during the
 274 Proterozoic and Archean, atmospheric shielding by the magnetic field was potentially
 275 tenuous despite the robust, albeit weaker than present day, dipole field.

276 6 Conclusions

277 Using an updated PINT database, we have developed a new continuous dipole field
 278 modelling approach (MCADAM). Based on three approaches of selection data using Q_{PI} cri-
 279 teria, our MCADAM models can robustly recover the average dipole field strength and
 280 captures key features previously identified in the Quaternary (Ziegler et al., 2011), the

281 Mesozoic (Kulakov et al., 2019), and the Precambrian (Biggin et al., 2015; Bono et al.,
282 2019).

283 We estimate the magnetic standoff distance from our preferred model MCADAM.1b
284 and show that the after the earliest Archean, the standoff distance was less than half the
285 present day distance of ~ 10 Earth radii due to the strong solar wind pressure stream-
286 ing off the younger Sun. Following the Young Sun lows, the paleomagnetosphere experi-
287 enced a protracted (~ 20 – 100 Myr) minima during the Ediacaran, followed by a highly
288 variable but generally increasing standoff distance in the Phanerozoic.

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294 available at <http://www.pintdb.org/>. MCADAM.1a-c model outputs are available on
295 the Earth Ref Data Archive, ERDA, at <http://www.earthref.org/ERDA/>.

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