Standard Paleo intensity Definitions v1.2.0

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1 Preface

Obtaining reliable paleointensity estimates is an important endeavor for understanding the behaviour of the geomagnetic field, but even after almost 70 years of experimentation, it still remains one of the most challenging aspects of modern paleomagnetism. The problems and pitfalls of paleointensity studies are well documented in the literature and great efforts have been made to solve or circumvent many of these issues. One of main tools in achieving this, is a suite of statistics that we can use to quantify and select what we believe to be the most reliable data.

Sadly, it is all too common, even in the modern literature, to find ambiguous descriptions of selection statistics and procedures, or situations where statistics are inconsistently calculated between different studies. Therefore, in an effort to promote consistency in data analysis and reporting, when have compiled the Standard Paleointensity Definitions (SPD), which presents an extensive list of detailed textual and mathematical definitions for paleointensity statistics to aid analysts. SPD not only lists these statistics, but provides numerical and computational advice on how to appropriately and efficient analyze paleointensity data.

SPD is intended to be a useful reference document for the paleomagnetic community and we hope that the community will take up on SPD and contribute to its development. If readers have comments, suggestions, corrections, or criticisms, we warmly invite them to contact us as we would appreciate all input that can help to further improve our ability to consistently select reliable paleointensity data. Maintenance and updating of SPD, however, is our responsibility and we are accountable for any mistakes and omissions (particularly GAP!). The latest and legacy versions of SPD are available at https://earthref.org/PmagPy/SPD/.

We are also pleased to announce that SPD is now an integral part of paleointensity.org, which is a highly flexible application to analyze paleointensity data produced by different methods and experiments. This easy to use website can handle wide range of data formats and is completely FAIR data compliant (Findable, Accessible, Interoperable, and Reusable), easily allowing your data to be shared or made available through community databases (such as MagIC). More info can be gotten from Beguin et al. (2020; G-Cubed, doi: 10.1029/2019GC008791).

if you do use SPD or in your work (or software that uses SPD), please make it clear what version was used. For example, by stating "paleointensity statistics were calculated following the Standard Paleointensity Definitions v1.2.0" or "paleointensity data were analyzed using software X (citing the appropriate reference), which follows the conventions laid out by the Standard Paleointensity Definitions v1.2.0". We would also be grateful if you cited the publication that introduced the SPD: Paterson, G. A., L. Tauxe, A. J. Biggin, R. Shaar, and L. C. Jonestrask (2014), On improving the selection of Thellier-type paleointensity data, *Geochem. Geophys. Geosyst.*, doi: 10.1002/2013GC005135..

Lastly, throughout our work we have benefited from the ideas and discussions from countless others. All of these have helped to shape SPD in ways both small and large. We want to thank Julie Bowles, Fabio Donadini, Roman Leonhardt, Adrian Muxworthy, Peter Selkin, Hidefumi Tanaka, and Yuhji Yamamoto for their generosity in providing data.

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2 Version history

Please note that the version numbering of the SPD example MATLAB code is not the same as the main SPD reference document.

Pre-v1.0

Draft versions during development.

v1.0 – February 24, 2014

The first finalized version of SPD, published along side Paterson, G. A., L. Tauxe, A. J. Biggin, R. Shaar, and L. C. Jonestrask (2014), On improving the selection of Thellier-type paleointensity data, *Geochem. Geophys. Geosyst.*, doi: 10.1002/2013GC005135..

v1.1 – May 01, 2014 Changes:

- 1. Corrected minor typographical errors and references.
- 2. Added $|\vec{k}'|$, which is the Arai plot curvature of the best-fit line (Section 3).
- 3. Added numerical tip for the calculation of the noncentral t critical value (Section 10).
- 4. Added a note that some version of PmagPy and ThellierGUI call MD_{VDS} "MD(%)", but their calculation is identical.

v1.2 – February 10, 2021 Changes:

- 1. Corrected σ_b to use b and not |b| (Section 3).
- 2. Corrected the anisotropy design matrix in Section 8. The last 2 elements of row 6 were incorrectly labeled as $P_{3,3}$ and $P_{3,1}$. They have now been corrected to $P_{2,2}$ and $P_{2,1}$, respectively.
- 3. Added pTRM check P_{ck} (Section 5).
- 4. Added S, S', and p_{χ^2} , measures of the quality of the Arai plot best-fit line that can be used as a statistical test (Section 3).
- 5. Updated some references.
- 6. Updated SPD web links.
- 7. Updated version numbering to three level semantic versioning (i.e., x.y.z).

3 Arai plot statistics

3.1 A note on data indexing

Statistic: i and n_{max}

The index *i* is used to denote the *i*th temperature step of the paleointensity experiment. *i* is used to index Arai plot data (e.g., x_i , or y_i) and ranges from i = 1 to n_{max} , where n_{max} is the total number of steps on the Arai plot.

Statistic: start and end

start and end denote the i indices of the selected steps used for analyzing the paleointensity results. i = start denotes the first selected data point and i = end denotes the last.

Statistic: T_{min} and T_{max}

The minimum and maximum temperatures used for the best-fit linear segment on the Arai plot, where $T_{min} \equiv T_{i=start}$ and $T_{max} \equiv T_{i=end}$.

Statistic: n

The number of points on an Arai diagram used to estimate the best-fit linear segment and the paleointensity (n = end - start + 1).

3.2 The paleointensity estimate

Statistic: b

Report to 3 d.p.

The slope of the best-fit line of the selected TRM and NRM points on the Arai plot. Determination of the slope uses the standardized major axis form of least squares linear fitting (*York*, 1966; *Coe et al.*, 1978).

$$b = \operatorname{sign}\left\{\sum_{i=start}^{end} (x_i - \bar{x})(y_i - \bar{y})\right\} \left(\frac{\sum_{i=start}^{end} (y_i - \bar{y})^2}{\sum_{i=start}^{end} (x_i - \bar{x})^2}\right)^{\frac{1}{2}}$$

where \bar{x} and \bar{y} are the mean TRM and NRM values of the data selected for the best-fit, that is,

$$\bar{x} = \frac{\sum_{i=start}^{end} x_i}{n},$$

and

$$\bar{y} = \frac{\sum_{i=start}^{ena} y_i}{n}.$$

Statistic: σ_b

Report to 3 d.p.

The standard error on the slope is given by:

$$\sigma_b = \left(\frac{2\sum_{i=start}^{end} (y_i - \bar{y})^2 - 2b\sum_{i=start}^{end} (x_i - \bar{x})(y_i - \bar{y})}{(n-2)\sum_{i=start}^{end} (x_i - \bar{x})^2}\right)^{\frac{1}{2}}$$

Useful Note...

It should be noted that the standard line-fitting routines available in most analysis software (e.g., Excel) do not use the standardized major axis fitting routine, but instead use linear regression (sometimes known as ordinary least-squares), whereby only the y-axis residuals are minimized. Given that accurate estimation of the slope is the objective of paleointensity analysis, standardized major axis, as outlined above, is the most appropriate method (e.g., *Warton et al.*, 2006).

Statistic: B_{Anc} and σ_B , the paleointensity estimate and its error

Report to 1 d.p.

A paleointensity estimate is obtained from $B_{Anc} = |b| \times B_{Lab}$, where B_{Lab} is the strength of the laboratory field. The associated standard error of the estimate is given by $\sigma_B = \sigma_b \times B_{Lab}$.



3.3Arai plot statistics

Figure 1: Schematic illustration of an Arai plot and some quantities used in the calculation of paleointensity statistics.

Statistic: $Y_{Int.}$ The y-axis (NRM) intercept of the best-fit line on the Arai plot.

$$Y_{Int.} = \bar{y} - b\bar{x}$$

Statistic: $X_{Int.}$

The x-axis (TRM) intercept of the best-fit line on the Arai plot.

$$X_{Int.} = \frac{-Y_{Int.}}{b}$$

Statistic: Vector difference sum, VDS

The vector difference sum of the entire NRM vector (\mathbf{NRM}).

$$VDS = |\mathbf{NRM}_{n_{max}}| + \sum_{i=1}^{n_{max}-1} |\mathbf{NRM}_{i+1} - \mathbf{NRM}_i|,$$

where $|\mathbf{NRM}_i|$ denotes the length of the NRM vector at the i^{th} step.

Statistic: x' and y'

x' and y' the x and y points on the Arai plot projected on to the best-fit line. These are used to calculate the NRM fraction and the length of the best-fit line among other parameters. There are multiple ways of calculating x' and y', below is one example.

$$x_i' = \frac{1}{2} \left(x_i + \frac{y_i - Y_{Int.}}{b} \right)$$
$$y_i' = \frac{1}{2} \left(y_i + bx + Y_{Int} \right)$$

Statistic: $\Delta x'$ and $\Delta y'$

 $\Delta x'$ and $\Delta y'$ are TRM and NRM lengths of the best-fit line on the Arai plot, respectively (Figure 1).

$$\Delta x' = \left| \left[\max \left\{ x'_i \right\} - \min \left\{ x'_i \right\} \right]_{i=start,\dots,end} \right|$$
$$\Delta y' = \left| \left[\max \left\{ y'_i \right\} - \min \left\{ y'_i \right\} \right]_{i=start,\dots,end} \right|$$

Statistic: f

Report to 3 d.p.

NRM fraction used for the best-fit on an Arai diagram (*Coe et al.*, 1978).

$$f = \frac{\Delta y'}{|Y_{Int.}|}$$

Statistic: f_{VDS} Report to 3 d.p.

NRM fraction used for the best-fit on an Arai diagram calculated as a vector difference sum (*Tauxe and Staudigel*, 2004).

$$f_{VDS} = \frac{\Delta y'}{VDS}$$

Statistic: FRAC

Report to 3 d.p.

NRM fraction used for the best-fit on an Arai diagram determined entirely by vector difference sum calculation (*Shaar and Tauxe*, 2013).

$$FRAC = \frac{\sum_{i=start}^{end-1} |\mathbf{NRM}_{i+1} - \mathbf{NRM}_i|}{VDS}$$

Statistic: β

Report to 3 d.p.

 β is a measure of the relative data scatter around the best-fit line and is the ratio of the standard error of the slope to the absolute value of the slope (*Coe et al.*, 1978).

$$\beta = \frac{\sigma_b}{|b|}$$

Statistic: g

Report to 3 d.p.

The gap factor (g) is a measure of the average NRM lost between successive temperature steps of the segment chosen for the best-fit line on the Arai plot. The gap reflects the average spacing of the selected Arai plot points along the best-fit line.

$$g = 1 - \frac{\sum_{i=start}^{end-1} (y'_{i+1} - y'_i)^2}{\Delta y'^2}$$

The upper limit of g is dependent on n and occurs when the points on the Arai plot are evenly spaced.

$$g_{lim} = \frac{n-2}{n-1}.$$

Statistic: *GAP-MAX Report to 3 d.p.*

The gap factor defined above is measure of the average Arai plot point spacing and may not represent extremes of spacing. To account for this *Shaar and Tauxe* (2013) proposed *GAP-MAX*, which is the maximum gap between two points determined by vector arithmetic.

$$GAP-MAX = \frac{\max\left\{|\mathbf{NRM}_{i+1} - \mathbf{NRM}_{i}|\right\}_{i=start,...,end-1}}{\sum_{i=start}^{end-1}|\mathbf{NRM}_{i+1} - \mathbf{NRM}_{i}|}$$

Statistic: q

Report to 1 d.p.

The quality factor (q) is a measure of the overall quality of the paleointensity estimate and combines the relative scatter of the best-fit line, the NRM fraction and the gap factor (*Coe et al.*, 1978).

$$q = \frac{|b| fg}{\sigma_b} = \frac{fg}{\beta}$$

Statistic: w

Report to 1 d.p.

Weighting factor of *Prévot et al.* (1985).

$$w = \frac{fg}{s},$$

where s^2 is given by:

$$s^{2} = 2 + \frac{2\sum_{i=start}^{end} (x_{i} - \bar{x})(y_{i} - \bar{y})}{\left(\sum_{i=start}^{end} (x_{i} - \bar{x})^{\frac{1}{2}} \sum_{i=start}^{end} (y_{i} - \bar{y})^{2}\right)^{2}}.$$

It can be noted, however, that w can be more readily calculated as:

$$w = \frac{q}{\sqrt{n-2}}.$$

Statistic: $\left| \vec{k} \right|$ Report to 3 d.p.

The curvature of the Arai plot as determined by the best-fit circle to all of the data (*Paterson*, 2011). To determine the Arai curvature, a best-fit circle of the form $(x-a)^2 + (y-b)^2 = r^2$ is fitted to all of the data using a least-squares approach. For the fitting process, each axis is normalized by the maximum value of the data on that axis such that $0 \leq \text{TRM} \leq 1$, and $0 \leq \text{NRM} \leq 1$, which ensures a consistent comparison between data measured with different B_{Lab} . $|\vec{k}|$ is defined as the reciprocal of the radius (r) of the best-fit circle:

$$\left|\vec{k}\right| = \frac{1}{r}.$$

Curvature can be given a sense of direction by considering the position of the circle center (a, b) with respect to the centroid of all of the data (C_x, C_y) .

$$\vec{k} = \begin{cases} \frac{1}{r} & \text{if } (C_x < a) \text{ and } (C_y < b) \\ -\frac{1}{r} & \text{if } (a < C_x) \text{ and } (b < C_y) \\ 0 & \text{if } (a = C_x) \text{ and } (b = C_y) \end{cases}$$

Numerical Tip...

Standard non-linear line fitting routines can be used for the calculation of the best-fit circle to the Arai plot data, however, the convergence of these algorithms can be poor and they are often numerically inaccurate when the data form a small arc of a much larger circle. This latter point is particularly important for near linear Arai plots as the data represent an increasingly smaller arc of the circle as the linearity increases. *Chernov and Lesort* (2005) developed an algorithm for fitting circles to data that is less affected by both of these issues and should be the preferred method of circle fitting (*Paterson*, 2011). Code for this algorithm in C++ and MATLAB are available from https://earthref.org/PmagPy/SPD/downloads.html.

Statistic: $\left| \vec{k}' \right|$

Report to 3 d.p.

The curvature of the Arai plot as determined by the best-fit circle to the selected best-fit Arai plot segment (*Paterson*, 2011). $|\vec{k'}|$ is calculated in the same fashion as $|\vec{k}|$, but the data are normalized by the respective maximums of the segment NRM and TRM.

Statistic: SSE

Report to 3 d.p.

The quality of the best-fit circle used to determine $|\vec{k}|$ (*Paterson*, 2011).

$$SSE = \sum_{i=1}^{n_{max}} \left(\sqrt{(x_i - a)^2 + (y_i - b)^2} - r \right)^2$$

Where x_i and y_i denote the normalized TRM and NRM, respectively.

Statistic: SCAT

SCAT is a parameter proposed by *Shaar and Tauxe* (2013) in an effort to reduce the number of parameters used to quantify a paleointensity estimate. SCAT is a Boolean operator, which uses the error on the best-fit Arai plot slope to indicate whether the data over the selected range are too scattered or not. This parameter provides a test for the scatter of the points on the Arai plot, pTRM checks, and pTRM tail checks. A schematic illustration of SCAT and some examples are shown in Figure 2.



Figure 2: (a) Schematic illustration of the calculation of the SCAT box. Examples of data that (b) pass and (c) fail SCAT. In all examples the gray shaded area is the SCAT box. In parts (b) and (c) open (closed) circles denote the selected (unselected) points. The triangles and squares represent pTRM checks and pTRM tail checks, respectively.

First, from the chosen the best-fit segment on the Arai plot, the slope (b), the standard error of the slope (σ_b) , and $\beta \left(=\frac{\sigma_b}{|b|}\right)$ are obtained. For a given sample, the threshold value for β that is used to select data $(\beta_{threshold})$ is used to determine an equivalent threshold for σ_b $(\sigma_{threshold} = |b| \beta_{threshold})$.

b and $\sigma_{threshold}$ are then used to determine two lines that pass through the center of mass of the selected Arai plot segment $(\bar{x} \text{ and } \bar{y})$, one with a slope of $b + 2\sigma_{threshold}$, the other with a slope of

 $b-2\sigma_{threshold}$ (Figure 2a). The so called *SCAT* box, is the box that is defined by the four intercepts that the above two lines make with the x- and y-axes (Figure 2a).

If all the data points associated with the chosen Arai plot segment, which includes both pTRM and tail checks, fall within the SCAT box then SCAT is TRUE. If one or more points fall outside the SCAT box then SCAT is FALSE. Samples are accepted only if SCAT is TRUE. pTRM checks and pTRM tail checks are included in the calculation of SCAT if the temperature of the check falls within temperature range of the selected Arai plot segment and the peak temperature before the check was performed is less than or equal to maximum temperature of the selected Arai plot segment. For example, if the temperature range of the selected Arai plot segment was 100–500°C, a pTRM check to 200°C performed after the 400°C step would be included in the calculation of SCAT. However, a pTRM check to 200°C performed after the 540°C would be *not* included in the calculation of SCAT. Examples of samples pass and fail SCAT are shown in Figure 2b and c, respectively.

Statistic: R_{corr}^2

Report to 3 d.p.

The correlation coefficient to estimate the strength of the linear relationship between the NRM and TRM over the best-fit Arai plot segment (the square of the Pearson correlation).

$$R_{corr}^2 = \frac{\left(\sum_{i=start}^{end} (x_i - \bar{x})(y_i - \bar{y})\right)^2}{\sum_{i=start}^{end} (x_i - \bar{x})^2 \sum_{i=start}^{end} (y_i - \bar{y})^2}$$

Statistic: R_{det}^2 Report to 3 d.p.

Coefficient of determination to estimate variance accounted for by the linear model fit.

$$R_{det}^{2} = 1 - \frac{\sum_{i=start}^{end} (y_{i} - y'_{i})^{2}}{\sum_{i=start}^{end} (y_{i} - \bar{y})^{2}}$$

Useful Note...

It should be noted that this is similar to, but strictly not the same as the square of the Pearson correlation coefficient (R_{corr}^2) . For least squares fitting that minimizes the y residuals only, R_{corr}^2 is the same as the coefficient of determination for the model fit. Since Arai plot analysis uses the standardized major axis least-squares variant, this is not the case. For most practical purposes, however, the difference is small, particularly when the chosen Arai plot segment is highly linear with low noise.

Report to 1 d.p.

Z is an Arai plot zigzag parameter defined by Yu and Tauxe (2005).

$$Z = \sum_{i=start}^{end} \frac{x_i \left| \tilde{b}_i - |b| \right|}{|X_{Int.}|},$$

where \tilde{b}_i is the instantaneous slope on the Arai plot determined from the ratio of the NRM lost to the TRM gained at the i^{th} step:

$$\tilde{b}_i = \frac{NRM_{total} - NRM_i}{TRM_i} = \frac{Y_{Int.} - y_i}{x_i}.$$

Since no TRM is gained during the first step $\tilde{b}_1 = 0$. $\tilde{b}_i - |b|$ is a measure of the scatter around the best-fit slope on the Arai plot.

Yu (2012) proposed a modified version, Z^* .

$$Z^* = \frac{1}{n-1} \sum_{i=start}^{end} 100 \times \frac{x_i \left| \tilde{b}_i - |b| \right|}{|Y_{Int.}|}$$

Statistic: S and S'

Report to 2 d.p.

S and S' are two measures of how well the Arai plot best-fit line fits the data and can be used as a goodness-of-fit statistical test (*York*, 1966; *Yu et al.*, 2000). For the standardized major axis least-squares method used for paleointensity analyses, S is the quantity minimized when determining the Arai plot best-fit line (*York*, 1966) and is given as:

$$S = \frac{1}{b^2 \sigma_x^2 + \sigma_y^2} \sum_{i=start}^{end} i (y_i - bx_i - Y_{Int})^2,$$

where σ_x^2 and σ_y^2 are the variances of the x_i and y_i data used for the Arai plot best-fit line, respectively. That is,

$$\sigma_x^2 = \frac{\sum\limits_{i=start}^{end} (x_i - \bar{x})^2}{(n-1)},$$

and

$$\sigma_y^2 = \frac{\sum_{i=start}^{end} (y_i - \bar{y})^2}{(n-1)}$$

Yu et al. (2000) noted that S is χ^2 distributed with n-2 degrees of freedom and an expectation value of n-2. They defined S' as

$$S' = \frac{S}{n-2},$$

which has an expectation value of 1. If S' is larger than one mean that the linear fit is poor.

Statistic: p_{χ^2}

Report to 3 d.p.

The statistic S can be used to perform a χ^2 -goodness-of-fit test (Yu et al., 2000). The probability that our choice linear model fits the data is given by p_{χ^2} :

$$p_{\chi^2} = 1 - F_{\chi^2} \left(S; (n-2) \right),$$

where $F_{\chi^2}()$ chi-squared cumulative distribution function with n-2 degrees of freedom. If $p_{\chi^2} \leq 0.05$ we can reject the null hypothesis that the data follow the best-fit linear line (at the 5% significance level).

Statistic: IZZI_MD

Report to 3 d.p.

 $IZZI_MD$ is a parameter to quantify the zigzagging on an Arai plot (*Shaar et al.*, 2011), which is most pronounced for multidomain grains measured with the IZZI protocol (e.g., *Yu et al.*, 2004).

 $IZZI_MD$ is a measure of the area mapped out on the Arai plot and is determined using all the points on an Arai plot with the exception of the first step, where no TRM is imparted. The calculation is performed after the points have been normalized by the initial NRM, such that

$$x_{(n)i} = \frac{x_i}{y_1}$$
 and $y_{(n)i} = \frac{y_i}{y_1}$.

If we consider the three consecutive Arai plot points illustrated in Figure 3a. The lengths of each side of the triangle formed by these points are given by:

$$L_{1} = \sqrt{\left(x_{(n)i} - x_{(n)i+1}\right)^{2} + \left(y_{(n)i} - y_{(n)i+1}\right)^{2}};$$
$$L_{2} = \sqrt{\left(x_{(n)i+1} - x_{(n)i+2}\right)^{2} + \left(y_{(n)i+1} - y_{(n)i+2}\right)^{2}};$$

and

$$L_3 = \sqrt{\left(x_{(n)i+2} - x_{(n)i}\right)^2 + \left(y_{(n)i+2} - y_{(n)i}\right)^2}.$$

Following the cosine rule, the angle ϕ is

$$\phi = \arccos\left(\frac{L_2^2 + L_3^2 - L_1^2}{2L_2L_3}\right)$$

The height of the triangle can be expressed as

$$H = L_3 \sin\left(\phi\right),$$

and hence the area of the triangle is given by

$$A_i = \frac{L_2 L_3 \sin\left(\phi\right)}{2}.$$

Each A_i is given a sign (\pm) based on whether or not the ZI steps lie above the IZ steps, or *vice versa*. For ZI above IZ, A_i is positive, for IZ above ZI, A_i is negative. For the example in Figure 3, ZI is above IZ and the area is given a positive sign.

To determine the sign, we must determine the relative position of the mid-point of the three consecutive points. First, we calculate the best-fit line through the first and last points and obtain



Figure 3: Illustration of the calculation of $IZZI_MD$. (a) The calculation of the area bounded by three consecutive Arai plot points. (b) The determination of the relative position of the mid-point.

the intercept of the line $(a_1;$ Figure 3b). Using the slope of this best-fit line we determine the intercept of the line (a_2) when the line passes through the mid-point. If a_1 is less than a_2 the mid-point lies above the end points, but if a_2 is less than a_1 the mid-point lies below the end points. In the case where $a_1 = a_2$ all three points lie on a perfect straight line and both the area and the sign are identically zero. The pseudo-code for this is as follows, where S_i denotes the sign of the i^{th} area.

```
for i = 2 \rightarrow i = n_{max} - 2 do
   if a_1 = a_2 then
        S_i = 0
    else
        if Mid-point is a ZI then
            if a_1 < a_2 then
                S_i = 1 \quad \leftarrow \{ \text{Figure 3b falls here} \}
            else
                S_i = -1
            end if
        else if Mid-point is a IZ then
            if a_1 < a_2 then
                S_i = -1
            else
                S_i = 1
            end if
        end if
    end if
end for
```

IZZI_MD is the sum of the signed areas, normalized by length of the line connecting all of the

ZI steps.

$$IZZI_MD = \sum_{i=2}^{n_{max}-2} \frac{S_i A_i}{L_{ZI}}$$

where L_{ZI} is given by:

$$L_{ZI} = \sum_{i \in ZI \text{ points}} \sqrt{\left(x_{(n)i+2} - x_{(n)i}\right)^2 + \left(y_{(n)i+2} - y_{(n)i}\right)^2} \ .$$

N.B. The i + 2 increment assumes alternating ZI and IZ step, whereby if i is a ZI step, i + 1 is an IZ step, and i + 2 is a ZI step.

Numerical Tip...

Calculating the area bounded by a series of points is a common geometric problem. As a consequence most programming languages have functions to perform the calculation either as inbuilt features or as freely available routines. For example, MATLAB has the inbuilt function A=polyarea(p) to return the area, A, bounded by the points given by the two-dimensional matrix p:

$$\mathbf{p} = \begin{pmatrix} x_i & y_i \\ x_{i+1} & y_{i+1} \\ x_{i+2} & y_{i+2} \end{pmatrix}$$

4 Directional statistics

Statistic: Dec and Inc

Report to 1 d.p.

The declination (Dec) and the inclination (Inc) of NRM direction of the paleointensity data over the same range of points used for the paleointensity estimates. *Dec* and *Inc* are calculated using the principal component analysis (PCA) method of *Kirschvink* (1980) and may be obtained from either a free-floating or anchored fit.

The method of applying PCA to paleomagnetic data is as follows. Let $X_{1,i}$, $X_{2,i}$, and $X_{3,i}$ denote the x, y, and z Cartesian coordinates of the NRM vector of the selected points $(i = start \dots end)$. The center of mass of the data is given by the ordinates \bar{X}_1 , \bar{X}_2 , and \bar{X}_3 .

$$\bar{X}_1 = rac{\sum\limits_{i=start}^{end} X_{1,i}}{n}; \quad \bar{X}_2 = rac{\sum\limits_{i=start}^{end} X_{2,i}}{n}; \quad \bar{X}_3 = rac{\sum\limits_{i=start}^{end} X_{3,i}}{n}$$

The coordinates of the NRM vector are then transformed to be centered about \bar{X}_1 , \bar{X}_2 , and \bar{X}_3 .

$$X'_{1,i} = X_{1,i} - \bar{X}_1; \quad X'_{2,i} = X_{2,i} - \bar{X}_2; \quad X'_{3,i} = X_{3,i} - \bar{X}_3,$$

where $X'_{1,i}$, $X'_{2,i}$, and $X'_{3,i}$ are the transformed coordinates. A fit anchored to the origin of the component diagram can be obtained by setting $[\bar{X}_1, \bar{X}_2, \bar{X}_3]$ to [0, 0, 0].

The orientation tensor, \mathbf{T} , of the transformed NRM data is defined as:

$$\mathbf{T} = \begin{pmatrix} \sum X'_{1,i}X'_{1,i} & \sum X'_{1,i}X'_{2,i} & \sum X'_{1,i}X'_{3,i} \\ \sum X'_{1,i}X'_{2,i} & \sum X'_{2,i}X'_{2,i} & \sum X'_{2,i}X'_{3,i} \\ \sum X'_{1,i}X'_{3,i} & \sum X'_{2,i}X'_{3,i} & \sum X'_{3,i}X'_{3,i} \end{pmatrix}.$$

This orientation tensor is usually constructed within sample or geographic coordinates and consists of six independent elements. Typically, none of these elements zero. When the non-diagonal elements of \mathbf{T} are non-zero the vector components described by this coordinate system are not independent, they are correlated. There exists, however, a coordinate system in which the orientation tensor can be expressed in terms of three independent orthogonal components. The axes of this coordinates system are known as the eigenvectors of the matrix and can be expressed in linear algebra form as:

$\mathbf{T}\mathbf{V}=\tau\mathbf{V},$

where **V** is a matrix that contains the three eigenvectors (also know as principal axes) and τ is a diagonal matrix that contains the three eigenvalues. When ranked by τ , such that $\tau_3 < \tau_2 < \tau_1$, the principal axis (i.e., $\mathbf{V}_1 = [V_{1,x}, V_{1,y}, V_{1,z}]$) corresponds to axis of the characteristic paleomagnetic direction.

Numerical Tip...

PCA is a widely used technique and numerous programming languages have inbuilt PCA functions. In MATLAB, for example, the command [V, tau]=eig(T) returns V and τ . The equivalent in Python is tau,V=numpy.linalg.eig(T).

It should be noted, however, that paleomagnetic direction may be either parallel or anti-parallel to \mathbf{V}_1 and the sense of direction must be established. To do this, we define a reference vector (**R**) defined as the difference between the first and last NRM vector measurements:

$$\mathbf{R} = \mathbf{X}'_{i=start} - \mathbf{X}'_{i=end} = \begin{bmatrix} X'_{1,start} - X'_{1,end}, X'_{2,start} - X'_{2,end}, X'_{3,start} - X'_{3,end} \end{bmatrix}.$$

The vector dot product of \mathbf{V}_1 and \mathbf{R} is then determined.

$$dot = \mathbf{V}_1 \cdot \mathbf{R} = V_{1,x} \times R_x + V_{1,y} \times R_y + V_{1,z} \times R_z$$

The range of dot is then truncated to exist only over [-1, 1].

```
if dot < -1 then

dot = -1

else if dot > 1 then

dot = 1

end if
```

The principal paleomagnetic direction $(\mathbf{PD} = [PD_x, PD_y, PD_z]$ can be given a sense of direction along \mathbf{V}_1 as follows.

```
if \arccos (dot) > \frac{\pi}{2}
then
PD = -V_1
else
PD = V_1
end if
```

The declination (Dec) and inclination (Inc) of principal paleomagnetic direction can be calculated.

if
$$PD_x < 0$$
 then
 $Dec = \arctan\left(\frac{PD_y}{PD_x}\right) + 180^\circ$
else if $PD_x > 0$ and $PD_y \le 0$ then
 $Dec = \arctan\left(\frac{PD_y}{PD_x}\right) + 360^\circ$
else
 $Dec = \arctan\left(\frac{PD_y}{PD_x}\right)$
end if

$$Inc = \arctan\left(\frac{PD_z}{\sqrt{PD_x^2 + PD_y^2}}\right)$$

Useful Note...

The choice of free-floating or anchored fit should be stated by using subscripts on Dec and Inc. That is, Dec_{Free} and Inc_{Free} for a free-floating fit and $Dec_{Anc.}$ and $Inc_{Anc.}$ for an anchored fit. Statistic: $MAD_{Anc.}$ and MAD_{Free} Report to 1 d.p.

 $MAD_{Anc.}$ and MAD_{Free} are Maximum Angular Deviation (MAD) of the anchored and freefloating, respectively, directional fits to the paleomagnetic vector on a vector component diagram (*Kirschvink*, 1980). Determined from the paleointensity demagnetization steps.

MAD is calculated as:

$$MAD = \arctan\left(\sqrt{\frac{\tau_2 + \tau_3}{\tau_1}}\right),$$

where $\tau_3 < \tau_2 < \tau_1$ are the eigenvalues of the PCA matrix.

Statistic: α

Report to 1 d.p.

Angular difference between the anchored and free-floating best-fit directions on a vector component diagram.

Numerical Tip...

Most directional parameters are related to the angle between two vectors. The angle between two vectors (denoted \mathbf{a} and \mathbf{b}) can be calculated by a simple rearrangement of the formulation for the dot product:

$$\theta = \arccos\left(\frac{\mathbf{a} \cdot \mathbf{b}}{|\mathbf{a}| |\mathbf{b}|}\right).$$

This approach, however, can be prone to numerical inaccuracies when θ is close to zero or π . For most practical purposes this should not be an issue, but the following alternative formulation can be used to avoid these inaccuracies:

$$heta = \arctan\left(rac{|\mathbf{a} imes \mathbf{b}|}{\mathbf{a} \cdot \mathbf{b}}
ight),$$

where \times denotes the vector cross product. The **atan2** function in most programing languages can be used to determine the appropriate quadrant. N.B. The above fraction is equivalent to $\frac{y}{x}$ in the terminology of most **atan2** functions.

Statistic: α'

Report to 1 d.p.

Angular difference between the anchored best-fit direction from the paleointensity experiment and an independent measure of the paleomagnetic direction (*Kissel and Laj*, 2004). The independent direction can be derived from a separate demagnetization experiment or from a known reference direction.

Statistic: θ

Report to 1 d.p.

The angle between the applied field direction and the ChRM direction of the NRM as determined from the free-floating PCA fit to the selected demagnetization steps of the paleointensity experiment.

Statistic: DANG

Report to 1 d.p.

The Deviation ANGle: the angle between the free-floating best-fit direction and the direction between data center of mass $[\bar{X}_1, \bar{X}_2, \bar{X}_3]$ and the origin of the vector component diagram (*Tanaka* and Kobayashi, 2003; *Tauxe and Staudigel*, 2004).

Statistic: NRM_{dev} Report to 1 d.p.

The intensity deviation of the free-floating principal component from the origin of the vector component diagram, normalized by the total NRM intensity ($Y_{Int.}$; Tanaka and Kobayashi, 2003).

$$NRM_{dev} = \frac{\sin{(DANG)}\sqrt{\bar{X_1}^2 + \bar{X_2}^2 + \bar{X_3}^2}}{|Y_{Int.}|} \times 100$$

Statistic: γ

Report to 1 d.p.

The angle between the pTRM acquired at the last step used for the best-fit segment (i.e., $\mathbf{TRM}_{i=end}$) and the applied field direction (\mathbf{B}_{Lab}). γ can be used as a quick check to assess if a sample is strongly influenced by anisotropic TRM. See Section 8 for further information of measuring and quantifying anisotropy of TRM.

Statistic: CRM(%)Report to 1 d.p.

To detect the potential acquisition of chemical remanent magnetization (CRM), *Coe et al.* (1984) proposed CRM(%) to measure the deflection the NRM vector towards the direction of B_{Lab} as would be expected during the formation of CRM. A schematic illustration of the quantities involved in the calculation of CRM(%) are shown in Figure 4. To determine CRM(%) the characteristic remanent magnetization (ChRM) direction needs to be known. This must be determined from an independent demagnetization experiment.



Figure 4: Illustration of an NRM vector (**NRM**_i) being deflected from the expected ChRM direction toward the applied field direction (**B**_{Lab}) by the acquisition of CRM during laboratory heating.

First, the magnitude of the CRM vector is determined:

$$CRM_{i} = \|\mathbf{NRM}_{i}\| \frac{\sin(\phi_{1})}{\sin(\phi_{2})} = y_{i} \frac{\sin(\phi_{1})}{\sin(\phi_{2})},$$

where $\|\mathbf{NRM}_i\|$ denotes the length of the \mathbf{NRM}_i vector, which is equivalent to y_i . CRM(%) is the maximum value of CRM_i over the selected best-fit Arai segment, normalized by the length of the TRM portion of best-fit Arai line:

$$CRM(\%) = \frac{\max{\{CRM_i\}_{i=start,\dots,end}}}{\Delta x'} \times 100.$$

Useful Note...

In the original work by Coe CRM(%) was originally called "R". We have renamed this statistic to avoid confusion with the "R"'s related to Arai plot linearity and line-fitting $(R_{corr}^2 \text{ and } R_{det}^2)$ and the length of the resultant vector when undertaking a Fisher analysis of paleomagnetic directions. It should be noted that some older studies denote CRM(%) and "R".

5 pTRM check statistics

5.1 pTRM checks

A pTRM check is a repeat TRM acquisition step to test for changes in a specimen's ability to acquire TRM at blocking temperatures below the temperature of the check. The difference between a pTRM check and the original TRM is calculated as the scalar intensity difference. That is,

$$\delta pTRM_{i,j} = pTRM_check_{i,j} - TRM_i = pTRM_check_{i,j} - x_i,$$

where $pTRM_check_{i,j}$ is the pTRM check to the i^{th} temperature step after heating to the j^{th} temperature step. The order of the difference is such that pTRM checks smaller than the original TRM yield negative $\delta pTRM_{i,j}$ and pTRM checks larger than the original TRM give positive $\delta pTRM_{i,j}$. For a pTRM check to be included in the analysis, both T_i and T_j must be less than or equal to T_{max} .



Figure 5: Schematic illustration of pTRM checks on an Arai plot and the quantities used to calculate pTRM check statistics.

Statistic: n_{pTRM}

The number of pTRM checks used to analyze the best-fit segment on the Arai plot (i.e., the number of $pTRM_{i,j}$ with $T_i \leq T_{max}$ and $T_j \leq T_{max}$).

5.2 Maximum pTRM check parameters

Statistic: check(%)Report to 1 d.p.

Maximum absolute difference produced by a pTRM check, normalized by the TRM acquired at that heating step.

$$check(\%) = \max\left\{\frac{|\delta pTRM_{i,j}|}{x_i} \times 100\right\}_{i \le end \text{ and } j \le end}$$

Statistic: δCK

Report to 1 d.p.

Maximum absolute difference produced by a pTRM check, normalized by the total TRM (obtained from the intersection of the best-fit line and the x-axis on an Arai plot; *Leonhardt et al.*, 2004a).

$$\delta CK = \frac{\max\left\{|\delta pTRM_{i,j}|\right\}_{i \le end \text{ and } j \le end}}{|X_{Int.}|} \times 100$$

Statistic: DRAT

Report to 1 d.p.

Maximum absolute difference produced by a pTRM check, normalized by the length of the best-fit line (*Selkin and Tauxe*, 2000).

$$DRAT = \frac{\max\left\{|\delta pTRM_{i,j}|\right\}_{i \le end \text{ and } j \le end}}{L} \times 100,$$

where L is the length of the best-fit line on the Arai plot. L is given by:

$$L = \sqrt{(\Delta x')^2 + (\Delta y')^2},$$

where $\Delta x'$ and $\Delta y'$ are TRM and NRM lengths of the best-fit line on the Arai plot, respectively (Section 3).

Statistic: maxDEV Report to 1 d.p.

Maximum absolute difference produced by a pTRM check, normalized by the length of the TRM segment of the best-fit line on the Arai plot (*Blanco et al.*, 2012).

$$maxDEV = \frac{\max\left\{|\delta pTRM_{i,j}|\right\}_{i \le end \text{ and } j \le end}}{\Delta x'} \times 100$$

Statistic: P_{ck} Report to 1 d.p.

Maximum absolute difference produced by a pTRM check, normalized by the pTRM gained at the maximum temperature used for the best-fit on the Arai diagram (*Carvallo et al.*, 2004).

$$P_{ck} = \frac{\max\left\{|\delta pTRM_{i,j}|\right\}_{i \le end \text{ and } j \le end}}{x_{end}} \times 100$$

Useful Note...

In the original work by Carvallo P_{ck} was originally called "P". We have renamed this statistic to avoid confusion with other "P"'s used in paleomagnetic analyses.

5.3 Cumulative pTRM check parameters

Most cumulative pTRM checks can be calculated in two fashions. The first method, is the summation of the signed pTRM differences (i.e., $\pm \delta pTRM$), the second is to calculate the sum of the absolute pTRM difference (i.e., $|\delta pTRM|$). The convention of the Standard Paleointensity Definition is to denote the second approach with a prime ('). For example, CDRAT is calculated by the first method and CDRAT' by the second.

Statistic: CDRAT

Report to 1 d.p.

Cumulative DRAT (Kissel and Laj, 2004).

$$CDRAT = \frac{\left|\sum_{i=1}^{end} \delta pTRM_{i,j}\right|}{L} \times 100$$
$$CDRAT' = \frac{\sum_{i=1}^{end} |\delta pTRM_{i,j}|}{L} \times 100$$

Statistic: DRATS Report to 1 d.p.

Cumulative pTRM check difference normalized by the pTRM gained at the maximum temperature used for the best-fit on the Arai diagram (*Tauxe and Staudigel*, 2004).

$$DRATS = \frac{\left|\sum_{i=1}^{end} \delta pTRM_{i,j}\right|}{x_{end}} \times 100$$
$$DRATS' = \frac{\sum_{i=1}^{end} |\delta pTRM_{i,j}|}{x_{end}} \times 100$$

Statistic: Mean DRAT

Report to 1 d.p.

The average difference produced by a pTRM check, normalized by the length of the best-fit line.

Mean
$$DRAT = \frac{1}{n_{pTRM}} \frac{\left|\sum_{i=1}^{end} \delta pTRM_{i,j}\right|}{L} \times 100$$

Mean
$$DRAT' = \frac{1}{n_{pTRM}} \frac{\sum_{i=1}^{end} |\delta pTRM_{i,j}|}{L} \times 100$$

Statistic: Mean DEV Report to 1 d.p. Mean deviation of a pTRM check (*Blanco et al.*, 2012).

$$Mean \ DEV = \frac{1}{n_{pTRM}} \frac{\left|\sum_{i=1}^{end} \delta pTRM_{i,j}\right|}{\Delta x'} \times 100$$
$$Mean \ DEV' = \frac{1}{n_{pTRM}} \frac{\sum_{i=1}^{end} |\delta pTRM_{i,j}|}{\Delta x'} \times 100.$$

Statistic: δpal Report to 1 d.p.

A measure of cumulative alteration determined by the difference of the alteration corrected intensity estimate (*Valet et al.*, 1996) and the uncorrected estimate, normalized by the uncorrected estimate (*Leonhardt et al.*, 2004a).

We first calculate the cumulative sum of the pTRM checks up to the i^{th} step of the experiment:

$$\mathbf{C}_{i} = \sum_{l=1}^{l=i} \delta \mathbf{pTRM}_{l,j}, \text{ for } i = 1, \dots, n_{max},$$

where $\delta \mathbf{pTRM}_{l,j}$ is the *vector* difference between \mathbf{TRM}_l and $\mathbf{pTRM}_{-}\mathbf{check}_{l,j}$, i.e.,

$$\delta \mathbf{pTRM}_{l,i} = \mathbf{TRM}_l - \mathbf{pTRM}_c\mathbf{heck}_{l,i}.$$

When no pTRM check is performed $\delta \mathbf{pTRM}_l = [0, 0, 0]$.

The \mathbf{TRM}_i vector is then corrected by adding the cumulative effect of the alteration, \mathbf{C}_i :

$$\mathbf{TRM}_i^* = \mathbf{TRM}_i + \mathbf{C}_i, \text{ for } i = 1, \dots, n_{max}.$$

Since no pTRM check is performed at the first step:

$\mathbf{TRM}_1^* = \mathbf{TRM}_1.$

The corrected TRM values on the Arai plot (x_i^*) can be calculated by determining the vector lengths of **TRM**_i^{*}. The corrected slope on the Arai plot (b^*) can be calculated using the selected points and the standard approach outlined in Section 3. δpal is then given by:

$$\delta pal = \left|\frac{b - b^*}{b}\right| \times 100.$$

6 pTRM tails check statistics

6.1 pTRM tail checks

A pTRM tail check is a repeat demagnetization step to test for changes in a specimen's magnetization carried in the blocking temperature range above the temperature of the check. The difference between the first NRM measurement and the pTRM tail check is calculated as the scalar intensity difference:

$$\delta tail_i = tail_check_i - NRM_i = tail_check_i - y_i,$$

where $tail_check_i$ is the pTRM tail check to the i^{th} temperature step. The order of the difference is such that tail checks smaller than the original NRM yield negative $\delta tail_i$ and tail checks larger than the original NRM give positive $\delta tail_i$. For a pTRM tail check to be included in the analysis, T_i must be less than or equal to T_{max} .



Figure 6: Schematic illustration of pTRM tail checks on an Arai plot and the quantities used to calculate pTRM tail check statistics.

Statistic: n_{Tail}

The number of pTRM tail checks conducted below the maximum temperature used for the best-fit segment on the Arai plot (i.e., the number of pTRM tail checks used to analyze the best-fit segment on the Arai plot).

Statistic: $DRAT_{Tail}$ Report to 1 d.p.

Maximum absolute difference produced by a pTRM tail check, normalized by the length of the best-fit line (*Biggin et al.*, 2007).

$$DRAT_{Tail} = \frac{\max\left\{|\delta tail_i|\right\}_{i=1,\dots,end}}{L} \times 100$$

Statistic: δTR Report to 1 d.p. Maximum absolute difference produced by a pTRM tail check, normalized by the NRM (obtained from the intersection of the best-fit line and the y-axis on an Arai plot; *Leonhardt et al.*, 2004a).

$$\delta TR = \frac{\max\left\{|\delta tail_i|\right\}_{i=1,\dots,end}}{|Y_{Int.}|} \times 100$$

Statistic: MD_{VDS} Report to 1 d.p.

Maximum absolute difference produced by a pTRM tail check, normalized by the vector difference sum of the NRM (*Tauxe and Staudigel*, 2004).

$$MD_{VDS} = \frac{\max\left\{|\delta tail_i|\right\}_{i=1,\dots,end}}{VDS} \times 100$$

Useful Note...

Some versions of PmagPy and ThellierGUI use a pTRM tail check statistic called MD(%). This is identical to MD_{VDS} , but the change in name emphasizes its calculation method.

Statistic: δt^*

Report to 1 d.p.

The extent of a pTRM tail after correction for angular dependence (*Leonhardt et al.*, 2004a,b).

The applied laboratory field vector (\mathbf{B}_{Lab}) is typically applied along a principle axis in the sample coordinate system (i.e., $\pm x$, $\pm y$, or $\pm z$). Therefore, for simplicity, δt^* should be calculated in the sample coordinate system only. Figure 7 is a schematic illustration of aspects of the calculation of δt^* .



Figure 7: Schematic illustration of an NRM vector (\mathbf{NRM}_i) and a pTRM tail check vector $(\mathbf{tail_check}_i)$ for a sample that exhibits a pTRM tail. Modified after *Leonhardt et al.* (2004b).

Let $N_{x,i}$, $N_{y,i}$, and $N_{z,i}$ denote the Cartesian coordinates of the NRM vector at step i (i.e., $\mathbf{NRM}_i = [N_{x,i}, N_{y,i}, N_{z,i}]$). Similarly, let $T_{x,i}$, $T_{y,i}$, and $T_{z,i}$ denote the Cartesian coordinates of the repeat demagnetization vector at step i (i.e., **tail_check**_i = [$T_{x,i}$, $T_{y,i}$, $T_{z,i}$]).

Assuming that \mathbf{B}_{Lab} is applied along the z-axis, the difference in the horizontal (δH_i) and vertical components (δZ_i) between **NRM**_i and **tail_check**_i (Figure 7) are given by:

$$\delta H_i = \sqrt{N_{x,i}^2 + N_{y,i}^2} - \sqrt{T_{x,i}^2 + T_{y,i}^2}$$

and

 $\delta Z_i = N_{z,i} - T_{z,i}.$

pTRM tails have an angular dependence and the calculation of δt^* requires two angular differences. Let $\Delta \theta_i$ denote the angle between \mathbf{B}_{Lab} and \mathbf{NRM}_i (see Section 4 for advice on calculating the angle between two vectors). Let δInc_i denote the difference in inclinations between the \mathbf{B}_{Lab} and \mathbf{NRM}_i :

$$\delta Inc_i = Inc(\mathbf{B}_{Lab}) - Inc(\mathbf{NRM}_i) = \arctan\left(\frac{B_{Lab,z}}{\sqrt{B_{Lab,x}^2 + B_{Lab,y}^2}}\right) - \arctan\left(\frac{N_{z,i}}{\sqrt{N_{x,i}^2 + N_{x,i}^2}}\right).$$

In the ThellierTool software (v4.22 and previous) \mathbf{B}_{Lab} is determined from each \mathbf{TRM}_i . Given that \mathbf{B}_{Lab} is almost always known, the convention of SPD is to use the known \mathbf{B}_{Lab} and not as estimated from \mathbf{TRM}_i , which may suffer from the effects of experimental noise.

As will be seen below, the calculation of δt^* requires $\frac{1}{\tan(\Delta \theta_i)}$. As $\Delta \theta_i$ approaches zero or 180° this fraction tends to infinity. To tackle this, δt^* is calculated in a piecewise fashion that depends on upper and lower angular limits (Lim_{upper} and Lim_{lower} , respectively). Below is pseudo-code that describes the logic of the calculation procedure.

```
 \begin{array}{ll} \mbox{if } \Delta \theta_i \ < \ Lim_{upper} \ \mbox{and } \Delta \theta_i \ > \ Lim_{lower} \\ \mbox{then} \\ \mbox{if } \delta Inc_i > 0 \ \mbox{then} \\ t_i^* = 100 \times |b| \ \frac{-\delta Z_i + \frac{\delta H_i}{\tan{(\Delta \theta_i)}}}{|Y_{Int.}|} \\ \mbox{else} \\ t_i^* = 100 \times |b| \ \frac{\delta Z_i - \frac{\delta H_i}{\tan{(\Delta \theta_i)}}}{|Y_{Int.}|} \\ \mbox{end if} \\ \mbox{else} \\ \mbox{if } \Delta \theta_i \le Lim_{lower} \ \mbox{then} \\ t_i^* = 0 \\ \mbox{else if } \Delta \theta_i \ge Lim_{upper} \ \mbox{then} \\ t_i^* = 100 \times \frac{-\delta Z_i}{|X_{Int.}| + |Y_{Int.}|} \\ \mbox{end if} \\ \mbox{end if} \\ \mbox{end if} \end{array}
```

In v4.22 of the ThellierTool $Lim_{lower} = 0.175 \ (\approx 10^{\circ})$ radians and $Lim_{upper} = 2.968 \text{ radians} (\approx 170^{\circ})$. These are adopted here.

 δt^* is then calculated as:

$$\delta t^* = \begin{cases} \max \{t_i^*\}_{i=1,\dots,end} & \text{if } (\max \{t_i^*\}_{i=1,\dots,end} > 0) \\ 0 & \text{if } (\max \{t_i^*\}_{i=1,\dots,end} < 0) \end{cases}$$

Only positive values of t^* and δt^* can be attributed to the effects of pTRM tails, hence δt^* is calculated as the maximum of t^* and not the maximum of $|t^*|$.

It should be noted that an implicit assumption in the above calculations is that \mathbf{B}_{Lab} is applied along the z-axis. In situations where \mathbf{B}_{Lab} is applied along the x- or y-axes, the definition of "horizontal" and "vertical" can be redefined such that \mathbf{B}_{Lab} is applied in the "vertical" direction. For example, if \mathbf{B}_{Lab} is along the x-axis, δH_i and δZ_i can be defined as:

$$\delta H_i = \sqrt{N_{y,i}^2 + N_{z,i}^2} - \sqrt{T_{y,i}^2 + T_{z,i}^2}$$

and

$$\delta Z_i = N_{x,i} - T_{x,i},$$

and

$$\delta Inc_i = \arctan\left(\frac{B_{Lab,x}}{\sqrt{B_{Lab,y}^2 + B_{Lab,z}^2}}\right) - \arctan\left(\frac{N_{x,i}}{\sqrt{N_{y,i}^2 + N_{z,i}^2}}\right).$$

The remaining calculations can proceed as described above.

7 Additivity check statistics

7.1 Additivity checks

An additivity check is a repeat demagnetization step to test the validity of Thellier's law of additivity (*Krása et al.*, 2003). In the course of a paleointensity experiment, a pTRM at temperature T_j is imparted, pTRM(T_j , T_0), where T_0 is room temperature. An additivity check demagnetizes pTRM(T_j , T_0) by heating to T_i , where $T_i < T_j$. The remaining pTRM (pTRM(T_j , T_i)) is subtracted from the previous pTRM acquisition step, pTRM(T_j , T_0), to estimate pTRM*(T_i , T_0). That is

$$pTRM^*(T_i, T_0) = pTRM(T_j, T_0) - pTRM(T_j, T_i)$$

where * denotes an estimated value. This estimated value can be compared with a previously observed value of $pTRM(T_i, T_0)$ that was measured earlier in the experiment. The difference between the estimated and observed pTRMs is a measure of the violation of additivity between T_i and T_0 . The additivity check difference $(AC_{i,j})$ is the scalar intensity difference between the two pTRMs:

$$AC_{i,j} = pTRM^*(T_i, T_0) - pTRM(T_i, T_0).$$

For an additivity check to be included in the analysis, both T_i and T_j must be less than or equal to T_{max} .



Figure 8: Schematic illustration of additivity checks on an Arai plot and the quantities used to calculate additivity check statistics.

Statistic: n_{Add}

The number of additivity checks used to analyze the best-fit segment on the Arai plot (i.e., the number of $AC_{i,j}$ with $T_i \leq T_{max}$ and $T_j \leq T_{max}$).

Statistic: δAC

Report to 1 d.p.

The maximum absolute additivity check difference normalized by the total TRM (obtained from the intersection of the best-fit line and the x-axis on an Arai plot; *Leonhardt et al.*, 2004a).

$$\delta AC = \frac{\max\left\{|AC_{i,j}|\right\}_{i \le end \text{ and } j \le end}}{|X_{Int.}|} \times 100.$$

8 Anisotropic TRM

8.1 The basic procedure

Correction of paleointensity results for anisotropic TRM is based on the premise that in a weak magnetic field a TRM vector (\mathbf{TRM}) is related to the applied field vector (\mathbf{B}) by:

$$\mathbf{TRM} = \chi_{TRM} \mathbf{B},$$

where χ_{TRM} is the TRM anisotropy tensor, which is assumed to be temperature invariant (*Veitch et al.*, 1984; *Selkin et al.*, 2000). The anisotropy tensor can be experimentally determined by giving a specimen a full TRM or ARM in 6 different directions $(\pm x, \pm y, \pm z)$.

To correct a paleointensity estimate, a unit vector in the direction of the ancient field $(\hat{\mathbf{B}}_{Anc};$ the hat denotes a unit vector) must be determined. Given a unit vector in the direction of the characteristic NRM direction $(\hat{\mathbf{M}}_{ChRM})$, $\hat{\mathbf{B}}_{Anc}$ can be calculated from:

$$\hat{\mathbf{B}}_{\mathbf{Anc}} = \frac{\chi_{TRM}^{-1} \hat{\mathbf{M}}_{\mathbf{ChRM}}}{\left|\chi_{TRM}^{-1} \hat{\mathbf{M}}_{\mathbf{ChRM}}\right|}.$$

 $\hat{\mathbf{M}}_{\mathbf{ChRM}}$ is determined from the free-floating PCA fit to the NRM steps from the selected Arai plot segment.

The paleointensity correction factor, c, which is the ratio of a magnetization gained in the direction of $\hat{\mathbf{B}}_{\text{Lab}}$ to a magnetization gained in the direction of $\hat{\mathbf{B}}_{\text{Anc}}$ can then be calculated as:

$$c = \frac{\left|\chi_{TRM}\hat{\mathbf{B}}_{Lab}\right|}{\left|\chi_{TRM}\hat{\mathbf{B}}_{Anc}\right|}.$$
(1)

The anisotropy corrected paleointensity estimate is simply given by:

$$B_{Anc} = cB_{Lab} \left| b \right|.$$

Useful Note...

Two methods to correct for anisotropy have been outlined in the literature (*Veitch et al.*, 1984; *Selkin et al.*, 2000). The method outlined above is that derived from *Veitch et al.* (1984), but both methods yield identical paleointensity estimates. The method of *Selkin et al.* (2000), however, can have a detrimental effect of some selection statistics and the method outlined above should be the preferred approach (*Paterson*, 2013).

8.2 Calculation of χ_{TRM}

The anisotropy tensor, χ_{TRM} , can be mathematically represented by a 3×3 matrix, which has 6 independent elements.

$$\chi_{TRM} = \begin{pmatrix} s_1 & s_4 & s_6 \\ s_4 & s_2 & s_5 \\ s_6 & s_5 & s_3 \end{pmatrix}$$

For convenience, we can define a column matrix, **s**, which contains the 6 independent elements, $s_j, j = 1 \dots 6$.

$$\mathbf{TRM}_i = \mathbf{A}_{i,j} \mathbf{s}_j,$$

where *i* denotes the i^{th} measurement position and **A** is know as the design matrix and depends on the experimental design (i.e., the sequence of axes along which the field is applied).

The typical 6 positions of measurement of χ_{TRM} or χ_{ARM} ($\pm x, \pm y, \pm z$) can be represented by the matrix **P**, which contains the unit vectors of the axes along which B_{Lab} is applied.

$$\mathbf{P} = \begin{pmatrix} P_{1,1} & P_{1,2} & P_{1,3} \\ P_{2,1} & P_{2,2} & P_{2,3} \\ \vdots & \vdots & \vdots \\ P_{6,1} & P_{6,2} & P_{6,3} \end{pmatrix}$$

For each element $P_{i,j}$ of **P**, *i* denotes the *i*th measurement position and j = 1...3, denotes the Cartesian coordinates of the unit vector along which the remanence is acquired (i.e., j = 1 = x, j = 2 = y, j = 3 = z). The design matrix of such a routine is given by:

$$\mathbf{A} = \begin{pmatrix} P_{1,1} & 0 & 0 & P_{1,2} & 0 & P_{1,3} \\ 0 & P_{1,2} & 0 & P_{1,1} & P_{1,3} & 0 \\ 0 & 0 & P_{1,3} & 0 & P_{1,2} & P_{1,1} \\ P_{2,1} & 0 & 0 & P_{2,2} & 0 & P_{2,3} \\ 0 & P_{2,2} & 0 & P_{2,1} & P_{2,3} & 0 \\ 0 & 0 & P_{2,3} & 0 & P_{2,2} & P_{2,1} \\ P_{3,1} & 0 & 0 & P_{3,2} & 0 & P_{3,3} \\ 0 & P_{3,2} & 0 & P_{3,1} & P_{3,3} & 0 \\ 0 & 0 & P_{3,3} & 0 & P_{3,2} & P_{3,1} \\ P_{4,1} & 0 & 0 & P_{4,2} & 0 & P_{4,3} \\ 0 & 0 & P_{4,3} & 0 & P_{4,2} & P_{4,1} \\ P_{5,1} & 0 & 0 & P_{5,2} & 0 & P_{5,3} \\ 0 & 0 & P_{5,2} & 0 & P_{5,1} & P_{5,3} & 0 \\ 0 & 0 & P_{5,3} & 0 & P_{5,2} & P_{5,1} \\ P_{6,1} & 0 & 0 & P_{6,2} & 0 & P_{6,3} \\ 0 & 0 & P_{6,2} & 0 & P_{6,1} & P_{6,3} & 0 \\ 0 & 0 & P_{6,3} & 0 & P_{6,2} & P_{6,1} \end{pmatrix}$$

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Numerical Tip...

To allow flexibility, the below pseudo-code can be used to easily generate \mathbf{A} , when \mathbf{P} is variable.

for $i = 1 \rightarrow 6$ do $index = (3 \times (i-1)) + 1$ A(index,1)=P(i,1)A(index,2)=P(i,2)A(index,3) = P(i,3)A(index+1,1)=P(i,1)A(index+1,2)=P(i,2)A(index+1,3)=P(i,3)A(index+2,1)=P(i,1)A(index+2,2)=P(i,2)A(index+2,3)=P(i,3)

end for

The best-fit values for \mathbf{s} for the measured data can be obtained through the linear relationship:

$$\mathbf{s} = \left(\mathbf{A}^T \mathbf{A}\right)^{-1} \mathbf{A}^T \mathbf{T} \mathbf{R} \mathbf{M},$$

where T and $^{-1}$ denote the matrix transpose and inverse, respectively. These best-fit value then be used to construct χ_{TRM} and hence determine c.

Numerical Tip...

Calculating the inverse of the anisotropy tensor or $(\mathbf{A}^T \mathbf{A})$ is not strictly necessary and can be inefficient and inaccurate. An alternative approach can be used if it is recognized that $\chi^{-1}_{TRM} \hat{\mathbf{M}}_{ChRM}$ and $(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T$ are linear problems of the form:

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}.$$

Many programming languages support tools that allow the solving of such linear systems without having to calculate the matrix inverse. For example, the MATLAB command $x=A\b$ or the Python command x=linalg.solve(A,b) are solutions that do not need to calculate the inverse of A. Such approaches are numerically efficient and more stable and should be used where available.

 χ_{TRM} is expressed in core coordinates, but much like the analysis of paleomagnetic directions, there exists an alternate coordinate system that that allows χ_{TRM} to be expressed in terms of principal components. That is,

$\chi_{TBM} \mathbf{V} = \tau \mathbf{V}$

where V is a matrix that contains the three eigenvectors (principal axes) and τ is a diagonal matrix that contains the three eigenvalues. The eigenvalues of the anisotropy tensor can then be used to characterize the anisotropy behaviour of a specimen (e.g., degree of anisotropy etc.) See Tauxe (2010) for further details.

8.3 Test for alteration after measurement of χ_{TRM}

Statistic: δTRM_{Anis} Report to 1 d.p.

To test for alteration during the measurement of χ_{TRM} a repeat remagnetization step is performed to the first treatment position. δTRM_{Anis} is the difference between the intensities of the TRM acquired during first heating in position 1 (TRM_{P1}) and the second heating in position 1 (TRM'_{P1}) normalized by TRM_{P1} .

 $\delta TRM_{Anis} = \frac{|TRM_{P1} - TRM'_{P1}|}{TRM_{P1}} \times 100$

9 Non-linear TRM

9.1 Theory and correction

An implicit assumption in paleointensity experiments is that TRM acquisition is linearly proportional to the applied field. However, both single domain theory (*Néel*, 1949) and experimental data (*Selkin et al.*, 2007) provide evidence that some particles can acquire TRM in a non-linear fashion in applied fields that are typical of the geomagnetic field intensity (i.e., $\leq 100 \ \mu$ T).

Single domain theory (*Néel*, 1949) predicts that TRM is proportional the hyperbolic tangent of the applied field (B) as described by:

$$TRM = M_{rs} \tanh\left(\frac{VM_s(T_b)B}{kT_b}\right),$$

where M_{rs} is the saturation remanent magnetization, V is the grain volume, $M_s(T_b)$ is the saturation magnetization at the blocking temperature (T_b) , and k is the Boltzmann constant. For weak fields the linear approximation generally holds true for most SD grains and is a result of the approximation that $tanh(x) \approx x$ for small x.

In comparison to Néel theory, *Selkin et al.* (2007) proposed that non-linear TRM acquisition be approximated by:

$$TRM = A_1 \tanh\left(A_2B\right),$$

where A_1 and A_2 are scaling coefficients. It can be noted that the linear approximation is valid in the limit as A_2 tends to zero and that linearity of magnetization with applied field is a special case of the more general non-linear form. This simple approximation assumes that A_2 is temperature invariant, which in the strictest sense is not true $\left(A_2 \propto \frac{M_s(T_b)}{T_b}\right)$. This approximation, however, has been demonstrated to fit real data well (*Selkin et al.*, 2007; *Shaar et al.*, 2010) and the assumption of a temperature invariant A_2 is equivalent to assuming that the degree of non-linearity is identical for all pTRMs and total TRMs.

The non-linear behaviour of a specimen can be determined by imparting the specimen with TRMs acquired in a range of applied field. A best-fit hyperbolic tangent model (of the form given above) can then be fitted to the data to determine the A_1 and A_2 coefficients. For an SD specimen in the absence of chemical alteration (i.e., the coefficients A_1 and A_2 do not change), the slope of the line (|b|) on an Arai plot can be described by:

$$|b| = \frac{NRM_{Anc}}{TRM_{Lab}} = \frac{\tanh\left(A_2B_{Anc}\right)}{\tanh\left(A_2B_{Lab}\right)}.$$

and B_{Anc} is:

$$B_{Anc} = \frac{\tanh^{-1}\left(|b|\tanh\left(A_2B_{Lab}\right)\right)}{A_2}.$$

9.2 Combined anisotropic and non-linear TRM correction

Many natural specimens, particularly archeological materials, suffer from both anisotropic and nonlinear TRM and therefore must be corrected for both. This can be achieved using:

$$B_{Anc} = \frac{\tanh^{-1} \left(c \left| b \right| \tanh \left(A_2 B_{Lab} \right) \right)}{A_2}$$

9.3 Test for alteration after measurement of A_1 and A_2

Statistic: δTRM_{NLT} Report to 1 d.p.

To test for alteration during the measurement TRM acquisition a repeat remagnetization step is performed in the first laboratory field, which is typically the same field as used for the paleointensity experiment (i.e., the final TRM acquisition in B_{Lab}). δTRM_{NLT} is the difference between the TRM acquire in first heating in B_{Lab} ($TRM_{B_{Lab}}$) and the second heating in B_{Lab} ($TRM'_{B_{Lab}}$) normalized by $TRM_{B_{Lab}}$.

 $\delta TRM_{NLT} = \frac{\left| TRM_{B_{Lab}} - TRM'_{B_{Lab}} \right|}{TRM_{B_{Lab}}} \times 100$

10 The statistics of multiple paleointensity estimates

10.1 Averaging and weighting

Statistic: N

The number of paleointensity estimates to be analyzed.

Statistic: B_j The value of the j^{th} paleointensity estimate, where j = 1 to N.

Statistic: m

Report to 1 d.p.

The arithmetic mean of the N paleointensity estimates

$$m = \frac{\sum_{j=1}^{N} B_j}{N}.$$

Statistic: s

Report to 1 d.p.

The standard deviation of the N paleointensity estimates.

$$s = \left(\frac{\sum_{j=1}^{N} (B_j - m)^2}{N - 1}\right)^{\frac{1}{2}}$$

Statistic: m_w Report to 1 d.p.

The weighted mean of the N paleointensity estimates.

$$m_w = \frac{\sum\limits_{j=1}^N W_j B_j}{\sum\limits_{j=1}^N W_j},$$

where W_j is the weight on the j^{th} paleointensity estimate.

Statistic: s_w

Report to 1 d.p.

The weighted standard deviation of the N paleointensity estimates (*Heckert and Filliben*, 2003).

$$s_w = \left(\frac{N\sum_{j=1}^{N} W_j (B_j - m_w)^2}{(N-1)\sum_{j=1}^{N} W_j}\right)^{\frac{1}{2}}$$

Useful Note...

Several options are available to act as weights. Two options that have been used in the literature are the quality and weighting factors, q and w, respectively. Their use as weighting factors, however, are not appropriate. As is outlined in Section 3, $w \propto q$, which itself is a function of the Arai plot slope (|b|). Hence, both q and w are proportional to the paleointensity estimate. If q or w are used as a weight (W_j) then $W_j \propto B_j$ (i.e., higher paleointensity estimates will tend to have larger weights), which can bias the weighted mean to higher values. Such dependencies should be carefully considered when deciding on the choice of which statistic to use for weighting.

10.2 Measures of scatter

Statistic: $\delta B(\%)$ Report to 1 d.p.

The standard deviation as a percentage of the mean value. Often referred to as the scatter.

$$\delta B(\%) = \frac{s}{m} \times 100$$

Statistic: $\delta B_N(\%)$

Report to 1 d.p.

When dealing with small numbers of data (i.e., small N), both m and s are inherently uncertain and these uncertainties propagate into measures of scatter. To account for this, *Paterson et al.* (2010a) proposed an adjustment to $\delta B(\%)$ to determine the upper 95% confidence interval ($\delta B_N(\%)$). Using this approach we can say, with 95% confidence, that the true scatter of the data is less than $\delta B_N(\%)$. This allows for a fairer comparison of data sets with different N.

$$\delta B_N(\%) = \left| \frac{\sqrt{N}}{t_{nc_{\left(1-\alpha; (N-1); \frac{m\sqrt{N}}{s}\right)}}} \right| \times 100,$$

where t_{nc} is the noncentral t critical value for the $(1 - \alpha)$ confidence level for (N - 1) degrees of freedom and with noncentrality parameter $\frac{m\sqrt{N}}{s}$.

Numerical Tip...

Different software packages use different conventions for the input of $(1 - \alpha)$ into the calculation of the noncentral t critical value. For example, the MATLAB command nctinv() takes $\alpha = 0.95$, while other function may use $\alpha = 0.05$. For N - 1 = 1 and a noncentrality parameter of unity (1) the noncentral t critical value at the 95% confidence level is -1.193.

10.3 Statistical tests for scatter

Statistic: $p_{\delta B}$

Report to 3 d.p.

An alternative approach is to determine the probability that the scatter (i.e., $\delta B(\%)$) is less than some critical value, δB_{max} (*Paterson et al.*, 2010a). By adopting this approach, selection based on scatter can be performed as a statistical test, whereby we test the null hypothesis that our measured scatter is less than or equal to δB_{max} . The probability $p_{\delta B}$ that this is the case is given by

$$p_{\delta B} = F_{nct} \left(\frac{\sqrt{N}}{\delta B_{max}}; \ (N-1) \ ; \ \frac{m\sqrt{N}}{s} \right),$$

where $F_{nct}()$ is the noncentral t cumulative distribution function and δB_{max} is given as a fraction and not a percentage (e.g., 0.25 as opposed to 25%). If $p_{\delta_B} \leq 0.05$ we cannot reject the null hypothesis that our measured scatter is less than or equal to δB_{max} (at the 5% significance level). If, however, $p_{\delta_B} > 0.05$ we can reject the null hypotheses and our measured scatter is most likely greater than δB_{max} . This approach is identical to $\delta B_N(\%)$, where $\delta B_N(\%)$ is the value of δB_{max} that yields $p_{\delta_B} = 0.05$.

Statistic: p_s

Report to 3 d.p.

Some studies prefer to select data using an absolute limit on the standard deviation (s_{max}) of an average paleointensity estimate, most notably when the estimate is low and the relative scatter may therefore be high. Given that, under the assumption of normality, estimated variance follows a scaled chi-squared distribution, the probability that s is less than or equal to s_{max} is given by:

$$p_s = F_{\chi^2} \left(\frac{(N-1)s_{max}^2}{s^2}; \ (N-1) \right),$$

where $F_{\chi^2}()$ chi-squared cumulative distribution function with N-1 degrees of freedom. If $p_s \leq 0.05$ we cannot reject the null hypothesis that our measured scatter is less than or equal to s_{max} (at the 5% significance level). If, however, $p_s > 0.05$ we can reject the null hypotheses and our measured scatter is most likely greater than s_{max} .

It should be noted that, in cases where $\delta B_{max} = \frac{s_{max}}{m}$, p_s is always less than $p_{\delta B}$. This is due to fact that $p_{\delta B}$ accounts for sample size related uncertainty in both m and s, but p_s accounts for sample size uncertainty in only s.

11 Calibration data set

To be able to test and calibrate the calculation of paleointensity data across multiple platforms and software bases, we have provided a data set of 20 paleointensity specimens with which users can test their software. The data set can be downloaded from https://earthref.org/PmagPy/SPD/ downloads.html in either MagIC or ThellierTool format.

11.1 Data sources

The 20 specimens used for testing and calibration are drawn from the compilation of data from historical location compiled by *Paterson et al.* (2014). The data are from 12 studies and represent 15 localities or laboratory experiments (16 unique heating events). The methods and materials used are outlined in Table 1. The data from *Selkin et al.* (2000) and *Shaar et al.* (2010) are corrected for the effects of non-linear TRM and/or anisotropy. The non-linear TRM coefficients (i.e., A_1 and A_2) as well as the six unique elements required to construct the anisotropy tensors (i.e., $s_{j=1...6}$) are available as text files to download from https://earthref.org/PmagPy/SPD/downloads.html.

11.2 Standardized parameters

A complete table of the standard paleointensity statistics and the details of the chosen best-fits for the above 20 specimens is available for download from https://earthref.org/PmagPy/SPD/ downloads.html. For the calculation of statistics that make a comparison with a known reference direction (e.g., α' , or CRM(%)) an arbitrary reference direction is used. This direction is Dec. = 90° , Inc = 45° , or $\left[0, \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right]$ in Cartesian coordinates. For the calculation of SCAT, $\beta_{threshold}$ is taken to be 0.1.

	Table 1:	Meta-data of the 20) specime	ens provide	to test and	d calibrate	with SPD.	
Sample	Reference	Location	pTRM	pTRM tail	Additivity	Method	Material	Comment
name			checks	checks	checks			
187A	Biggin et al. (2007)	Mt. Etna, 1979	$\mathbf{Y}_{\mathbf{es}}$	Yes	No	MW Coe	Basaltic lava	With AF cleaning
283A	$Biggin \ et \ al. \ (2007)$	Mt. Etna, 1950	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	MW Coe	Basaltic lava	
A-3-3	Yamamoto et al. (2003)	Hawaii, 1960	$\mathbf{Y}_{\mathbf{es}}$	No	N_{O}	Coe	Basaltic lava	
AL2770-3b	$Bowles \ et \ al. (2006)$	East Pacific Rise,	\mathbf{Yes}	No	N_{O}	Coe	SBG	
		1991/92				i		
BR06-4F	Donadini et al. (2007)	Helsinki, 1906	Yes	${ m Yes}$	No	Coe	Brick	
C-4-4L	Yamamoto et al. (2003)	Hawaii, 1960	$\mathbf{Y}_{\mathbf{es}}$	No	No	Coe	Basaltic lava	With low temperature
								demagnetization
HEL2-2d	$Donadini \ et \ al. \ (2007)$	Helsinki, 1906	$\mathbf{Y}_{\mathbf{es}}$	\mathbf{Yes}	N_{O}	Coe	Brick	
KF-3-1	$Tanaka \ et \ al. \ (2012)$	Krafia, 1984	Y_{es}	N_{O}	N_{O}	Coe	Basaltic lava	
LV6C3A	Paterson et al. (2010b)	Láscar, 1993	$\mathbf{Y}_{\mathbf{es}}$	\mathbf{Yes}	N_{O}	Coe	Andesitic	
							clast	
m428b1	Selkin et al. (2000)	Stillwater Complex	Y_{es}	No	N_{O}	Coe	An orthosite	Anisotropic TRM corrected
MCT	Muxworthy (1998)	N/A	N_{O}	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	Coe	Synthetic	Average grain size 22.5 μm
							magnetite	
MSH6E13	Paterson et al. (2010b)	Mt. St. Helens,	\mathbf{Yes}	\mathbf{Yes}	\mathbf{Yes}	Coe	Dacitic clast	
		1980						
P1MY	Muxworthy et al. (2011)	Parícutin, 1943	$\mathbf{Y}_{\mathbf{es}}$	$\mathbf{Y}_{\mathbf{es}}$	N_{O}	Coe	Basaltic lava	
RD2358-4f	Pick and Tauxe (1993)	East Pacific Rise,	$\mathbf{Y}_{\mathbf{es}}$	\mathbf{Yes}	N_{O}	Coe	SBG	
		1990						
RS25b	Shaar $et al. (2010)$	N/A	\mathbf{Yes}	N_{O}	N_{O}	IZZI	${\rm Remelted}$	Anisotropic and non-linear
							copper slag	TRM corrected
m RS26a	$Shaar \ et \ al. \ (2010)$	N/A	\mathbf{Yes}	N_{O}	N_{O}	IZZI	Remelted	Anisotropic and non-linear
							copper slag	TRM corrected
m RS26e	$Shaar \ et \ al. \ (2010)$	N/A	\mathbf{Yes}	No	No	IZZI	Remelted	Anisotropic and non-linear
							copper slag	TRM corrected
TS01-20A-2	Yamamoto and Hoshi (2008)	Sakurajima, 1914	\mathbf{Yes}	\mathbf{Yes}	No	Coe	Andesitic lava	
VM1F	Muxworthy et al. (2011)	Vesuvius, 1940	$\mathbf{Y}_{\mathbf{es}}$	\mathbf{Yes}	N_{O}	Coe	Basaltic lava	
W3	Krása et al. (2003)	N/A	\mathbf{Yes}	\mathbf{Yes}	\mathbf{Yes}	Coe	$\mathbf{Synthetic}$	Average grain size $0.5 \ \mu m$
							magnetite	

11.3 Averaging and descriptive statistics

The paleointensity estimates along with the average, standard deviation and scatter for the 20 specimens are given in Table 2. This table is also available for download as an Excel spreadsheet from https://earthref.org/PmagPy/SPD/downloads.html.

Sample name	B_{Exp} (μT)	$B_{Anc} (\mu T)$
187A	44.1	45.2
283A	43.3	116.3
A-3-3	36.2	35.6
AL2770-3b	35.8	26.0
BR06-4F	49.6	41.9
C-4-4L	36.2	36.1
HEL2-2d	49.6	49.5
KF-3-1	52.1	44.1
LV6C3A	24.0	23.4
m428b1	25.0	25.1
MCT	55.6	71.7
MSH6E13	55.6	22.1
P1MY	45.0	30.1
RD2358-4f	37.0	85.7
RS25b	30.0	29.3
RS26a	60.0	63.3
RS26e	60.0	58.1
TS01-20A-2	45.7	50.1
VM1F	44.0	74.0
W3	60.0	60.7
	N	20
	$m~(\mu T)$	49.4
	$s \; (\mu T)$	24.2

Table 2: Paleointensity estimates and average value with associated descriptive statistics.

11.3.1 Weighting

An example of the calculation of weighted average and standard deviation of the above described data set is given in the Excel spreadsheet available for download from https://earthref.org/PmagPy/SPD/downloads.html. In this example, the weights are taken to be $\frac{1}{(\sigma_B)^2}$.

 $\delta B(\%)$

 $\delta B_N(\%)$

48.9

66.3

References

- Biggin, A. J., M. Perrin, and M. J. Dekkers (2007), A reliable absolute palaeointensity determination obtained from a non-ideal recorder, *Earth Planet. Sci. Lett.*, 257, 545–563, doi:10.1016/j.epsl.2007. 03.017.
- Blanco, D., V. A. Kravchinsky, J.-P. Valet, A. Ali, and D. K. Potter (2012), Does the Permo-Triassic geomagnetic dipole low exist?, *Phys. Earth Planet. Inter.*, 204-205, 11–21, doi:10.1016/ j.pepi.2012.06.005.
- Bowles, J., J. S. Gee, D. V. Kent, M. R. Perfit, S. A. Soule, and D. J. Fornari (2006), Paleointensity applications to timing and extent of eruptive activity, 9°–10°N East Pacific Rise, *Geochem. Geophys. Geosyst.*, 7, Q06006, doi:10.1029/2005gc001141.
- Carvallo, C., Ö. Özdemir, and D. J. Dunlop (2004), Palaeointensity determinations, palaeodirections and magnetic properties of basalts from the Emperor seamounts, *Geophys. J. Int.*, 156, 29–38, doi:10.1111/j.1365-246X.2004.02110.x.
- Chernov, N., and C. Lesort (2005), Least squares fitting of circles, J. Math. Imaging Vis., 23, 239–252, doi:10.1007/s10851-005-0482-8.
- Coe, R. S., S. Grommé, and E. A. Mankinen (1978), Geomagnetic paleointensities from radiocarbondated lava flows on Hawaii and the question of the Pacific nondipole low, J. Geophys. Res., 83, 1740–1756, doi:10.1029/JB083iB04p01740.
- Coe, R. S., S. Grommé, and E. A. Mankinen (1984), Geomagnetic paleointensities from excursion sequences in lavas on Oahu, Hawaii, J. Geophys. Res., 89, 1059–1069, doi:10.1029/ JB089iB02p01059.
- Donadini, F., M. Kovacheva, M. Kostadinova, L. Casas, and L. J. Pesonen (2007), New archaeointensity results from Scandinavia and Bulgaria: Rock-magnetic studies inference and geophysical application, *Phys. Earth Planet. Inter.*, 165, 229–247, doi:10.1016/j.pepi.2007.10.002.
- Heckert, N. A., and J. J. Filliben (2003), NIST Handbook 148: Dataplot Reference Manual, Volume
 2: Let Subcommands and Library Functions, National Institute of Standards and Technology
 Handbook Series.
- Kirschvink, J. L. (1980), The least-squares line and plane and the analysis of palaeomagnetic data, *Geophys. J. R. Astr. Soc.*, 62, 699–718, doi:10.1111/j.1365-246X.1980.tb02601.x.
- Kissel, C., and C. Laj (2004), Improvements in procedure and paleointensity selection criteria (PICRIT-03) for Thellier and Thellier determinations: Application to Hawaiian basaltic long cores, *Phys. Earth Planet. Inter.*, 147, 155–169, doi:10.1016/j.pepi.2004.06.010.
- Krása, D., C. Heunemann, R. Leonhardt, and N. Petersen (2003), Experimental procedure to detect multidomain remanence during Thellier-Thellier experiments, *Phys. Chem. Earth*, 28, 681–687, doi:10.1016/S1474-7065(03)00122-0.
- Leonhardt, R., C. Heunemann, and D. Krása (2004a), Analyzing absolute paleointensity determinations: Acceptance criteria and the software ThellierTool4.0, *Geochem. Geophys. Geosyst.*, 5, Q12016, doi:10.1029/2004GC000807.

- Leonhardt, R., D. Krása, and R. S. Coe (2004b), Multidomain behavior during Thellier paleointensity experiments: A phenomenological model, *Phys. Earth Planet. Inter.*, 147, 127–140, doi:10.1016/j.pepi.2004.01.009.
- Muxworthy, A. R. (1998), Stability of magnetic remanence in multidomain magnetite, Ph.D. thesis.
- Muxworthy, A. R., D. Heslop, G. A. Paterson, and D. Michalk (2011), A Preisach method for estimating absolute paleofield intensity under the constraint of using only isothermal measurements:
 2. Experimental testing, J. Geophys. Res., 116, B04103, doi:10.1029/2010jb007844.
- Néel, L. (1949), Théorie du traînage magnétique des ferromagnétiques en grains fins avec applications aux terres cuites, Ann. Géophys., 5, 99–136.
- Paterson, G. A. (2011), A simple test for the presence of multidomain behaviour during paleointensity experiments, J. Geophys. Res., 116, B10104, doi:10.1029/2011JB008369.
- Paterson, G. A. (2013), The effects of anisotropic and non-linear thermoremanent magnetizations on Thellier-type paleointensity data, *Geophys. J. Int.*, 193, 694–710, doi:10.1093/gji/ggt033.
- Paterson, G. A., D. Heslop, and A. R. Muxworthy (2010a), Deriving confidence in paleointensity estimates, *Geochem. Geophys. Geosyst.*, 11, Q07Z18, doi:10.1029/2010gc003071.
- Paterson, G. A., A. R. Muxworthy, A. P. Roberts, and C. Mac Niocaill (2010b), Assessment of the usefulness of lithic clasts from pyroclastic deposits for paleointensity determination, J. Geophys. Res., 115, B03104, doi:10.1029/2009JB006475.
- Paterson, G. A., L. Tauxe, A. J. Biggin, R. Shaar, and L. C. Jonestrask (2014), On improving the selection of thellier-type paleointensity data, *Geochem. Geophys. Geosyst.*, 15, 1180–1192, doi:10.1002/2013GC005135.
- Pick, T., and L. Tauxe (1993), Geomagnetic palaeointensities during the Cretaceous normal superchron measured using submarine basaltic glass, *Nature*, 366, 238–242, doi:10.1038/366238a0.
- Prévot, M., E. A. Mankinen, R. S. Coe, and C. S. Grommé (1985), The Steens Mountain (Oregon) geomagnetic polarity transition: 2. Field intensity variations and discussion of reversal models, J. Geophys. Res., 90, 10,417–10,448, doi:10.1029/JB090iB12p10417.
- Selkin, P. A., and L. Tauxe (2000), Long-term variations in palaeointensity, Phil. Trans. R. Soc. London, 358, 1065–1088, doi:10.1098/rsta.2000.0574.
- Selkin, P. A., W. P. Meurer, A. J. Newell, J. S. Gee, and L. Tauxe (2000), The effect of remanence anisotropy on paleointensity estimates: A case study from the Archean Stillwater Complex, *Earth Planet. Sci. Lett.*, 183, 403–416, doi:10.1016/S0012-821X(00)00292-2.
- Selkin, P. A., J. S. Gee, and L. Tauxe (2007), Nonlinear thermoremanence acquisition and implications for paleointensity data, *Earth Planet. Sci. Lett.*, 256, 81–89, doi:10.1016/j.epsl.2007.01.017.
- Shaar, R., and L. Tauxe (2013), Thellier gui: An integrated tool for analyzing paleointensity data from thellier-type experiments, *Geochem. Geophys. Geosyst.*, 14, 677–692, doi:10.1002/ggge. 20062.
- Shaar, R., H. Ron, L. Tauxe, R. Kessel, A. Agnon, E. Ben-Yosef, and J. M. Feinberg (2010), Testing the accuracy of absolute intensity estimates of the ancient geomagnetic field using copper slag material, *Earth Planet. Sci. Lett.*, 290, 201–213, doi:10.1016/j.epsl.2009.12.022.

- Shaar, R., H. Ron, L. Tauxe, R. Kessel, and A. Agnon (2011), Paleomagnetic field intensity derived from non-SD: Testing the Thellier IZZI technique on MD slag and a new bootstrap procedure, *Earth Planet. Sci. Lett.*, 310, 213–224, doi:10.1016/j.epsl.2011.08.024.
- Tanaka, H., and T. Kobayashi (2003), Paleomagnetism of the late Quaternary Ontake Volcano, Japan: directions, intensities, and excursions, *Earth Planets Space*, 55, 189–202.
- Tanaka, H., Y. Hashimoto, and N. Morita (2012), Palaeointensity determinations from historical and Holocene basalt lavas in Iceland, *Geophys. J. Int.*, 189, 833–845, doi:10.1111/j.1365-246X. 2012.05412.x.
- Tauxe, L. (2010), Essentials of Paleomagnetism, University of California Press, Berkeley.
- Tauxe, L., and H. Staudigel (2004), Strength of the geomagnetic field in the Cretaceous Normal Superchron: New data from submarine basaltic glass of the Troodos Ophiolite, *Geochem. Geophys. Geosyst.*, 5, Q02H06, doi:10.1029/2003GC000635.
- Valet, J.-P., X. Quidelleur, E. Tric, P. Y. Gillot, J. Brassart, I. Le Meur, and V. Soler (1996), Absolute paleointensity and magnetomineralogical changes, J. Geophys. Res., 101, 25,029–25,044.
- Veitch, R. J., I. G. Hedley, and J.-J. Wagner (1984), An investigation of the intensity of the geomagnetic-field during roman times using magnetically anisotropic bricks and tiles, Arch. Sci., 37, 359–373.
- Warton, D. I., I. J. Wright, D. S. Falster, and M. Westoby (2006), Bivariate line-fitting methods for allometry, *Biol. Rev.*, 81, 259–291, doi:10.1017/s1464793106007007.
- Yamamoto, Y., and H. Hoshi (2008), Paleomagnetic and rock magnetic studies of the Sakurajima 1914 and 1946 and esitic lavas from Japan: A comparison of the LTD-DHT Shaw and Thellier paleointensity methods, *Phys. Earth Planet. Inter.*, 167, 118–143, doi:10.1016/j.pepi.2008.03.006.
- Yamamoto, Y., H. Tsunakawa, and H. Shibuya (2003), Palaeointensity study of the Hawaiian 1960 lava: implications for possible causes of erroneously high intensities, *Geophys. J. Int.*, 153, 263– 276, doi:10.1046/j.1365-246X.2003.01909.x.
- York, D. (1966), Least-squares fitting of a straight line, Can. J. Phys., 44, 1079–1086, doi:10.1139/ p66-090.
- Yu, Y. (2012), High-fidelity paleointensity determination from historic volcanoes in Japan, J. Geophys. Res., 117, B08,101, doi:10.1029/2012jb009368.
- Yu, Y., D. J. Dunlop, L. Pavlish, and M. Cooper (2000), Archeomagnetism of Ontario potsherds from the last 2000 years, J. Geophys. Res., 105, 19419–19433, doi:10.1029/2000JB900177,
- Yu, Y. J., and L. Tauxe (2005), Testing the IZZI protocol of geomagnetic field intensity determination, Geochem. Geophys. Geosyst., 6, Q06H11, doi:10.1029/2004GC000840.
- Yu, Y. J., L. Tauxe, and A. Genevey (2004), Toward an optimal geomagnetic field intensity determination technique, *Geochem. Geophys. Geosyst.*, 5, Q02H07, doi:10.1029/2003GC000630.