

Lecture 13:

Tensors in paleomagnetism

- What are tensors anyway?
- anisotropy of magnetic susceptibility (AMS)
 - how to find it
 - what to do with it
- anisotropy of magnetic remanence

What is a tensor?

- "An array of numbers (or functions) that transform according to certain rules under a change of coordinates."
- "Tensors are geometric entities [that] extend the notion of scalars, vectors and matrices."

Examples

- We used a tensor of direction cosines to transform coordinate systems:

$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

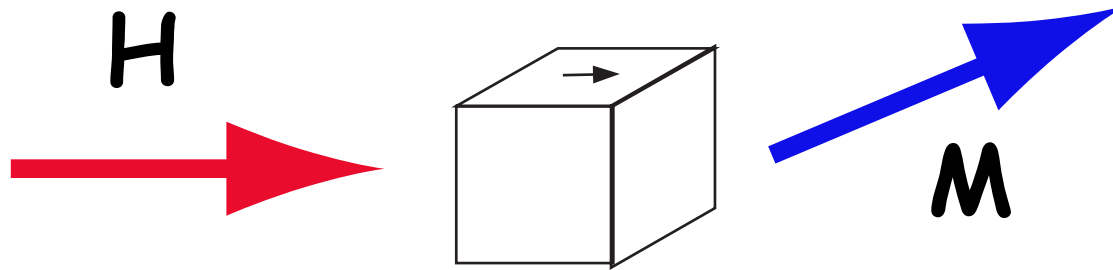
- The stress and strain tensor in structural geology
- The optical indicatrix in mineralogy
- AND in paleomagnetism, the relationship between the induced field and the applied field (sometimes not parallel!)

Anisotropy of Magnetic Susceptibility

In Chapter I, we learned that $M_I = \chi H$

where χ was just a number (scalar) and so M was parallel to and proportional to H

But what if the magnetic minerals in the rock are not isotropic?



$$M_1 = \chi_{11}H_1 + \chi_{12}H_2 + \chi_{13}H_3$$

$$M_2 = \chi_{21}H_1 + \chi_{22}H_2 + \chi_{23}H_3$$

$$M_3 = \chi_{31}H_1 + \chi_{32}H_2 + \chi_{33}H_3$$

or: $M_i = \chi_{ij}H_j$

$$M_i = \chi_{ij} H_j$$

For convenience (later on), we can remap this to:

$$s_1 = \chi_{11}$$

$$s_2 = \chi_{22}$$

$$s_3 = \chi_{33}$$

$$s_4 = \chi_{12} = \chi_{21}$$

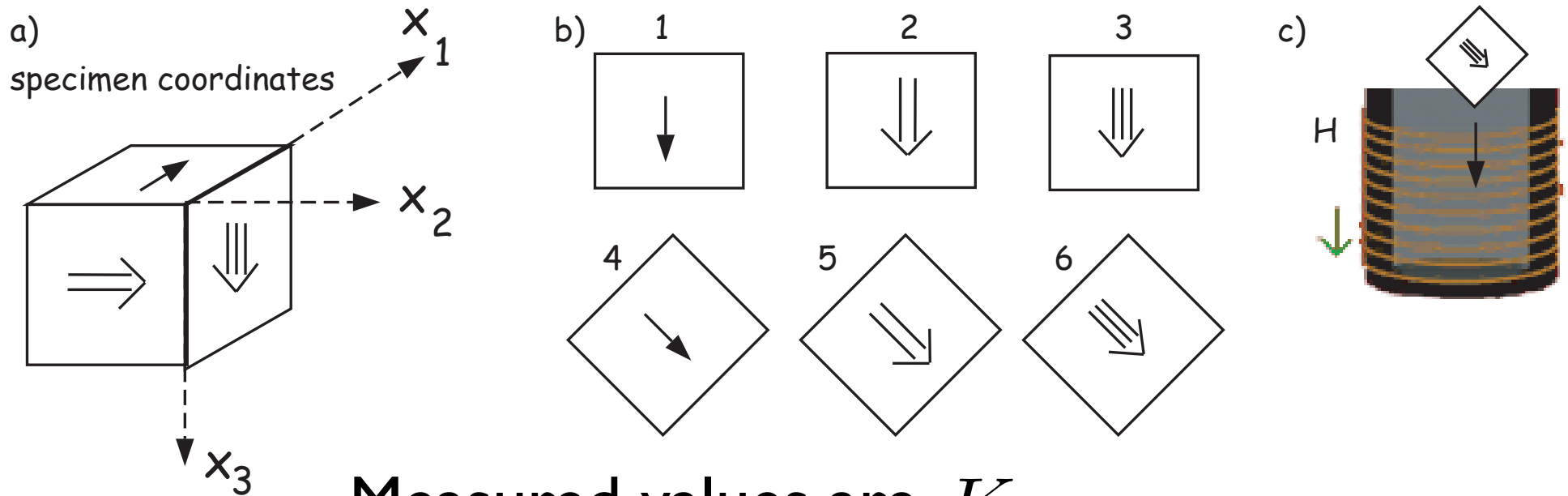
$$s_5 = \chi_{23} = \chi_{32}$$

$$s_6 = \chi_{13} = \chi_{31}$$

Applications of Anisotropy of magnetic susceptibility

- Related to distribution of magnetic minerals (easy axes) in rocks
- Can use it to understand rock fabrics
 - Sedimentary processes
 - Igneous processes
 - Metamorphic processes

need to measure in at least 6 positions to determine AMS tensor



Measured values are K_i

$$K_1 = s_1$$

$$K_2 = s_2$$

$$K_3 = s_3$$

BUT

$$K_4 = \frac{1}{2}(s_1 + s_2) + s_4$$

$$K_5 = \frac{1}{2}(s_2 + s_3) + s_5$$

$$K_6 = \frac{1}{2}(s_1 + s_3) + s_6$$

$$\begin{aligned}
K_1 &= s_1 & K_4 &= \frac{1}{2}(s_1 + s_2) + s_4 \\
K_2 &= s_2 & K_5 &= \frac{1}{2}(s_2 + s_3) + s_5 \\
K_3 &= s_3 & K_6 &= \frac{1}{2}(s_1 + s_3) + s_6
\end{aligned}$$

can be re-written as: $K_i = A_{ij}s_j$ where A_{ij} is:

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ .5 & .5 & 0 & 1 & 0 & 0 \\ 0 & .5 & .5 & 0 & 1 & 0 \\ .5 & 0 & .5 & 0 & 0 & 1 \end{pmatrix}$$

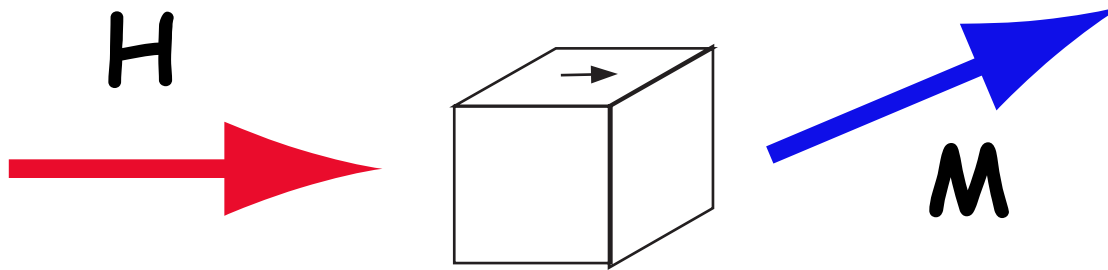
So how do we get the s_i from the measurements K_i ?

need tricks from linear algebra:

$$\bar{s} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{K} \quad \text{or} \quad \bar{s}_i = B_{ij} K_j$$

where \mathbf{B} is:

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ -0.5 & -0.5 & 0 & 1 & 0 & 0 \\ 0 & -0.5 & -0.5 & 0 & 1 & 0 \\ -0.5 & 0 & -0.5 & 0 & 0 & 1 \end{pmatrix}$$



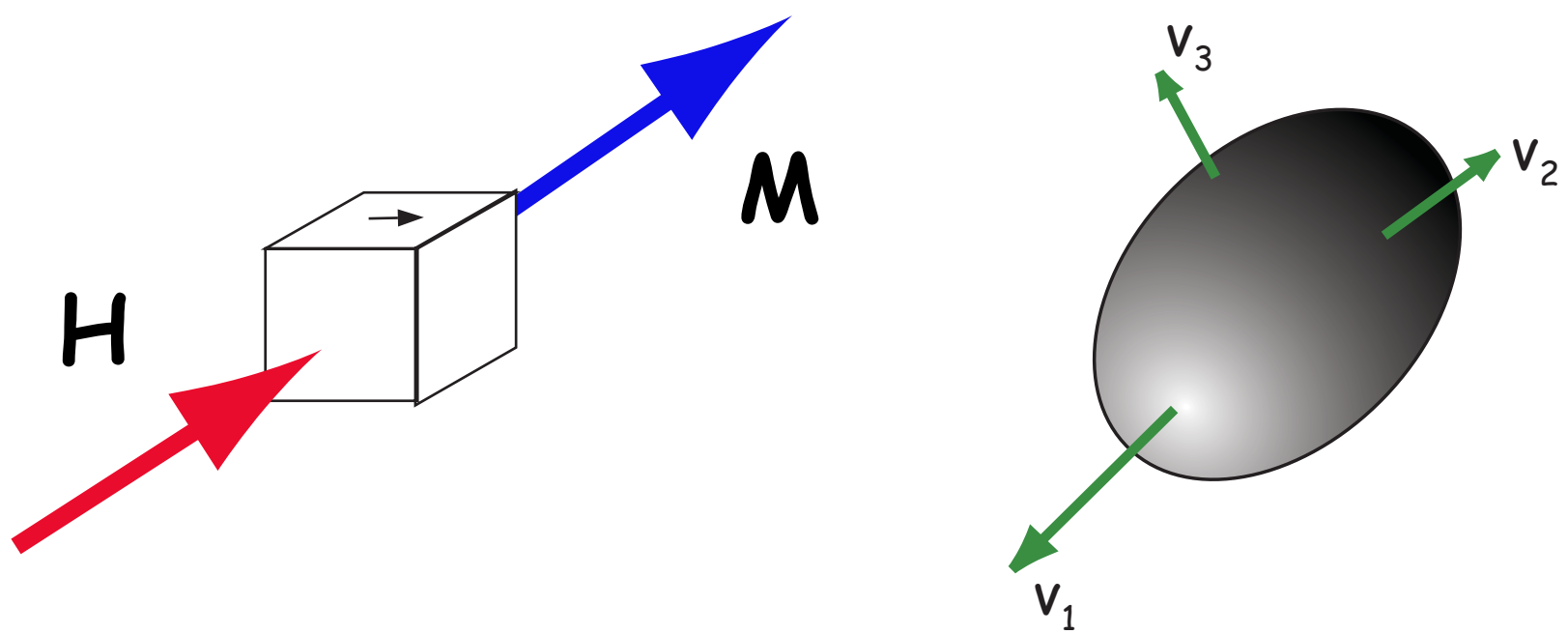
$$M_1 = \chi_{11}H_1 + \chi_{12}H_2 + \chi_{13}H_3$$

$$M_2 = \chi_{21}H_1 + \chi_{22}H_2 + \chi_{23}H_3$$

$$M_3 = \chi_{31}H_1 + \chi_{32}H_2 + \chi_{33}H_3$$

In general, M is not parallel to H because
coefficients

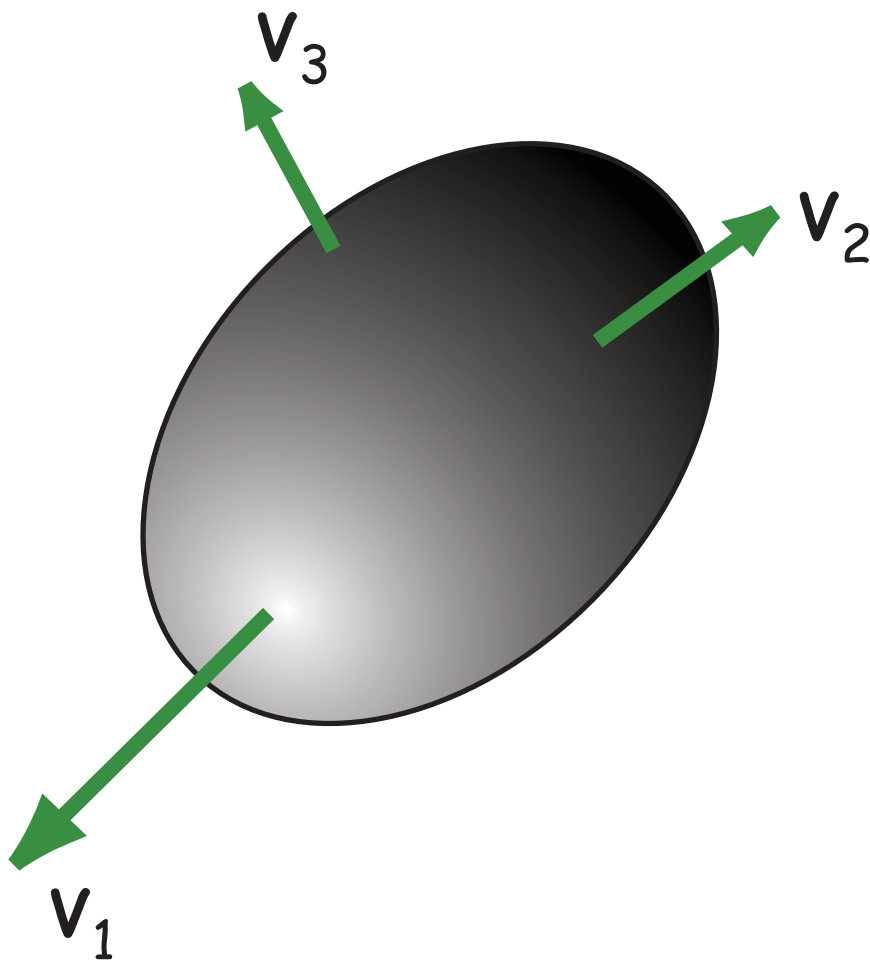
$$\chi_{ij} \neq 0 \text{ when } i \neq j$$



But there exists a coordinate system where M IS parallel to H .
Axes are called “eigenvectors”: V_i

so when $H \parallel V_1, M \parallel V_1$

and the magnitude corresponds to: $M = \chi_{max}H$



see Appendix
A.3.5.4 for how to find
eigenvectors and
eigenvalues

$$\chi_{max} \propto \tau_1$$

$$\chi_{int} \propto \tau_2$$

$$\chi_{min} \propto \tau_3$$

where τ_i are the eigenvalues

$$\tau_1 > \tau_2 > \tau_3$$

What is AMS good for?

- Is a particular fabric axis parallel to some direction? e.g., is the minimum axes parallel to the vertical as expected for sediments?
- Are two sets of axes distinct? e.g., has strain rotated the rock fabric?
- What is the shape of the ellipsoid? Is it "pancake" (as exp. for sediments?) or "hot-dog" (as exp. for deformed fabrics?)

These all require statistical test! - poor statistics with just 6 measurements

Statistics for AMS data

- Every measurement has some uncertainty:

$$K_i = A_{ij}\bar{s}_j + \delta_i$$

- Remember we get average tensor elements::

$$\bar{s}_i = B_{ij}K_j$$

- Now we could calculate what measurements we “should” have gotten, assuming that the average tensor was correct:

$$\bar{K}_i = A_{ij}\bar{s}_j$$

- and calculate the deviations (residuals) between “average” and actual measurements:

$$\delta_i = K_i - \bar{K}_i$$

The fun begins

- A measure of the uncertainty is the sum of squared residuals:

$$S_o = \sum_i \delta_i^2$$

- From this we get the estimated variance:

$$\sigma^2 = S_o/n_f$$

- But what is “degrees of freedom”, n_f ? It is the number of measurements less the number of things you are estimating

$$N_{meas} - 6$$

- So we need more than six measurements to do any cool statistics

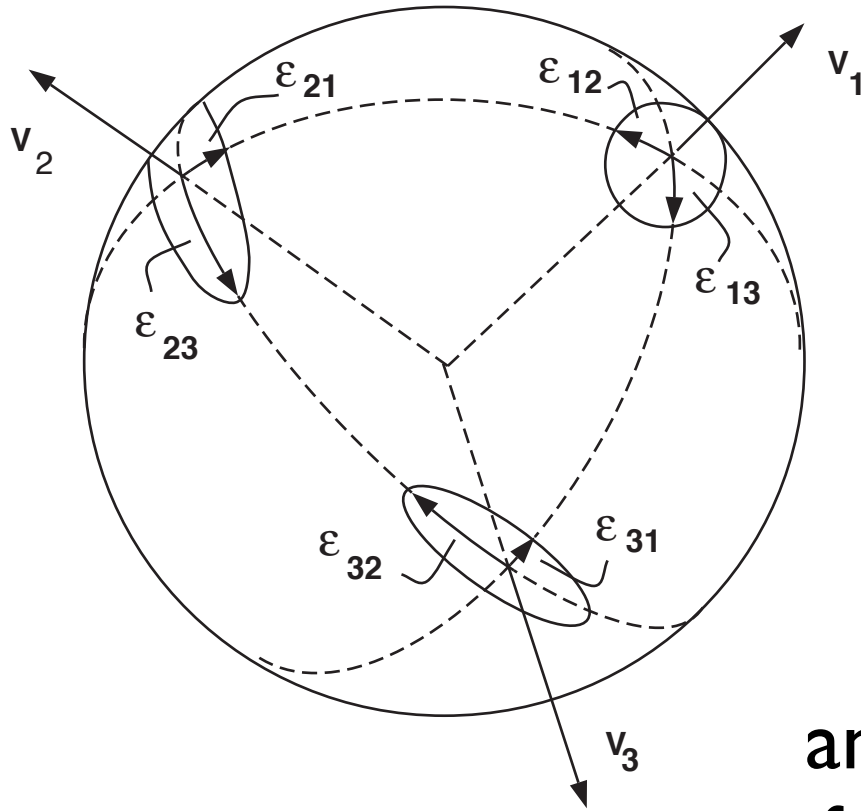
Measurement of AMS in practice

- Lots of different protocols (7, 12, 15, “many” measurement protocols)
- In Appendix D.1 you will find descriptions of two commonly used ones: 15 and the “spinning” protocols
- Each protocol has its own “design matrix”, B from which all data reduction flows

Parametric significance tests

- Analogous to Fisher statistics – known as Hext statistics
- Allows calculation of uncertainties in eigenvector directions
- Allows tests for shape (are the eigenvalues significantly different so what shape do we have?)

Uncertainty ellipses of eigenvectors: the parametric approach of Hext



$$\epsilon_{12} = \tan^{-1} [f\sigma/2(\tau_1 - \tau_2)]$$

$$\epsilon_{23} = \tan^{-1} [f\sigma/2(\tau_2 - \tau_3)]$$

$$\epsilon_{13} = \tan^{-1} [f\sigma/2(\tau_1 - \tau_3)]$$

$$\epsilon_{21} = \epsilon_{12}$$

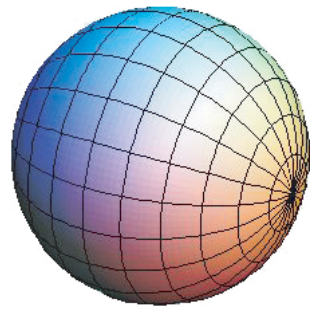
$$\epsilon_{32} = \epsilon_{23}$$

$$\epsilon_{31} = \epsilon_{13},$$

where $f = \sqrt{2(F_{(2, n_f); (1-p)})}$

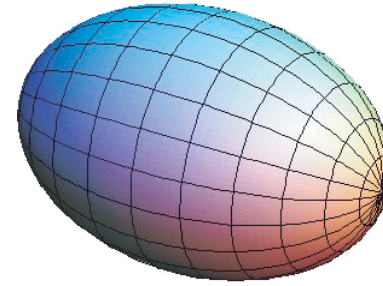
and this F is another critical value for significance at a given level of confidence (e.g., $p = .05$ for 95%) look up F on an “F-table”

Shapes - the Hext way: but first some definitions



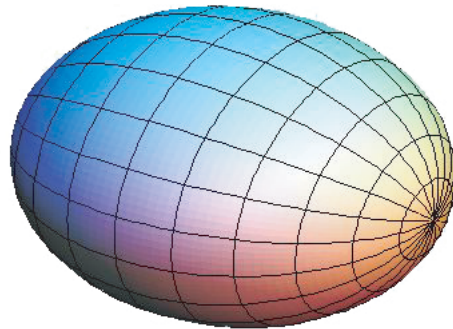
sphere

$$\tau_1 \simeq \tau_2 \simeq \tau_3$$



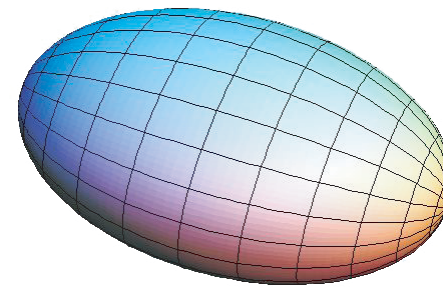
prolate

$$\tau_1 > \tau_2 \simeq \tau_3$$



oblate

$$\tau_1 \simeq \tau_2 > \tau_3$$



triaxial

$$\tau_1 > \tau_2 > \tau_3$$

F statistics for shapes

but first, bulk susceptibility:

$$\chi_b = (\bar{s}_1 + \bar{s}_2 + \bar{s}_3)/3$$

Are your data spherical? Must have F
less than this:

$$F = 0.4 (\tau_1^2 + \tau_2^2 + \tau_3^2 - 3\chi_b^2) / \sigma^2$$

Are they oblate?

$$F_{23} = 0.5((\tau_2 - \tau_3)/\sigma)^2$$

Are they prolate?

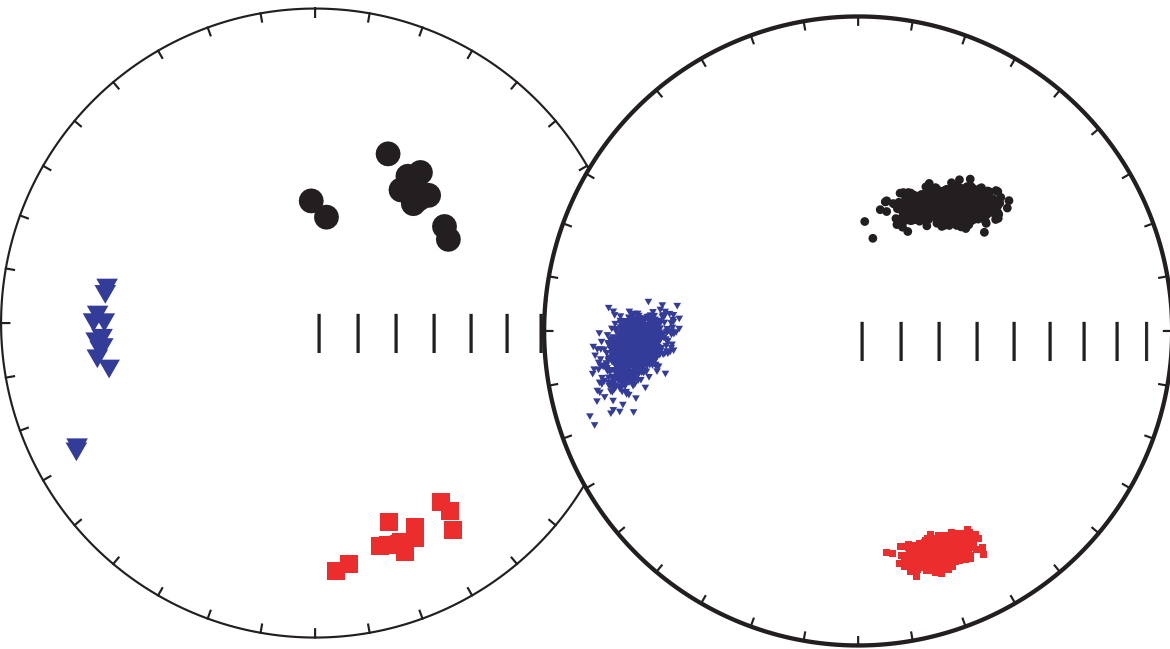
$$F_{12} = 0.5((\tau_1 - \tau_2)/\sigma)^2$$

But when do Hext statistics apply?

- When errors are small
- When they have zero mean
- AND when they are normally distributed
- oops - only for measurement data - rarely for data from multiple specimens
- maybe you should bootstrap?

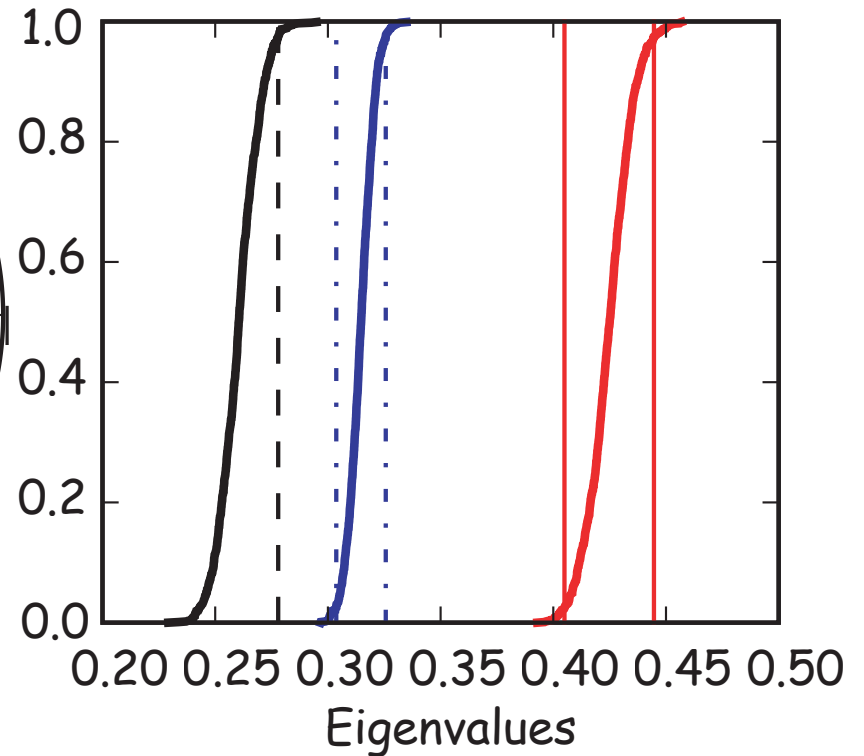
Bootstrapping for AMS

works just like for directions: see Chapter 12



Data

bootstrapped
eigenvectors

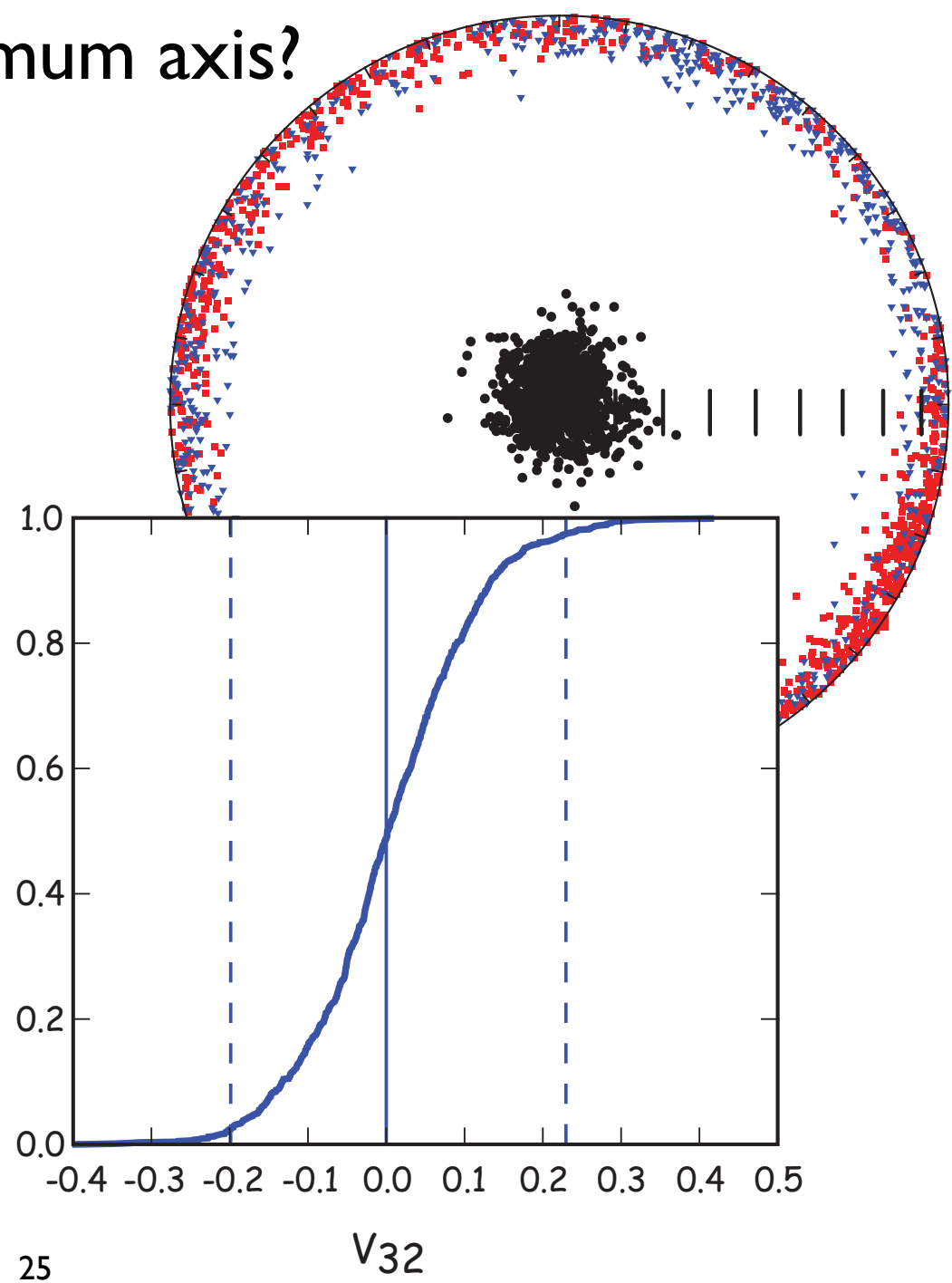
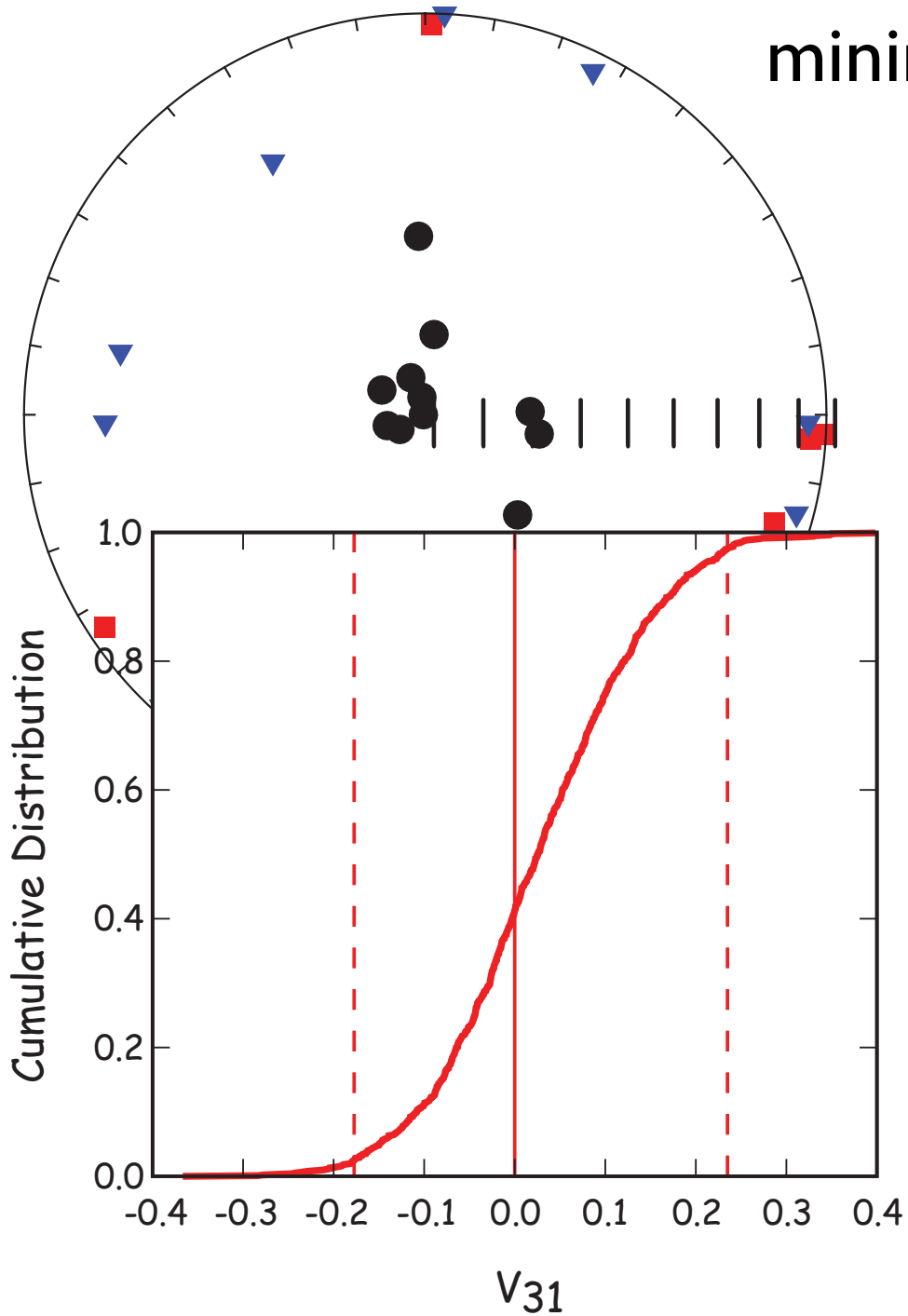


bootstrapped
eigenvalues

What is AMS good for?

- Is a particular fabric axis parallel to some direction? e.g., is the minimum axes parallel to the vertical as expected for sediments?
- Are two sets of axes distinct? e.g., has strain rotated the rock fabric?
- What is the shape of the ellipsoid? Is it “pancake” (as exp. for sediments?) or “hot-dog” (as exp. for deformed fabrics?)

Do these data have a vertical minimum axis?



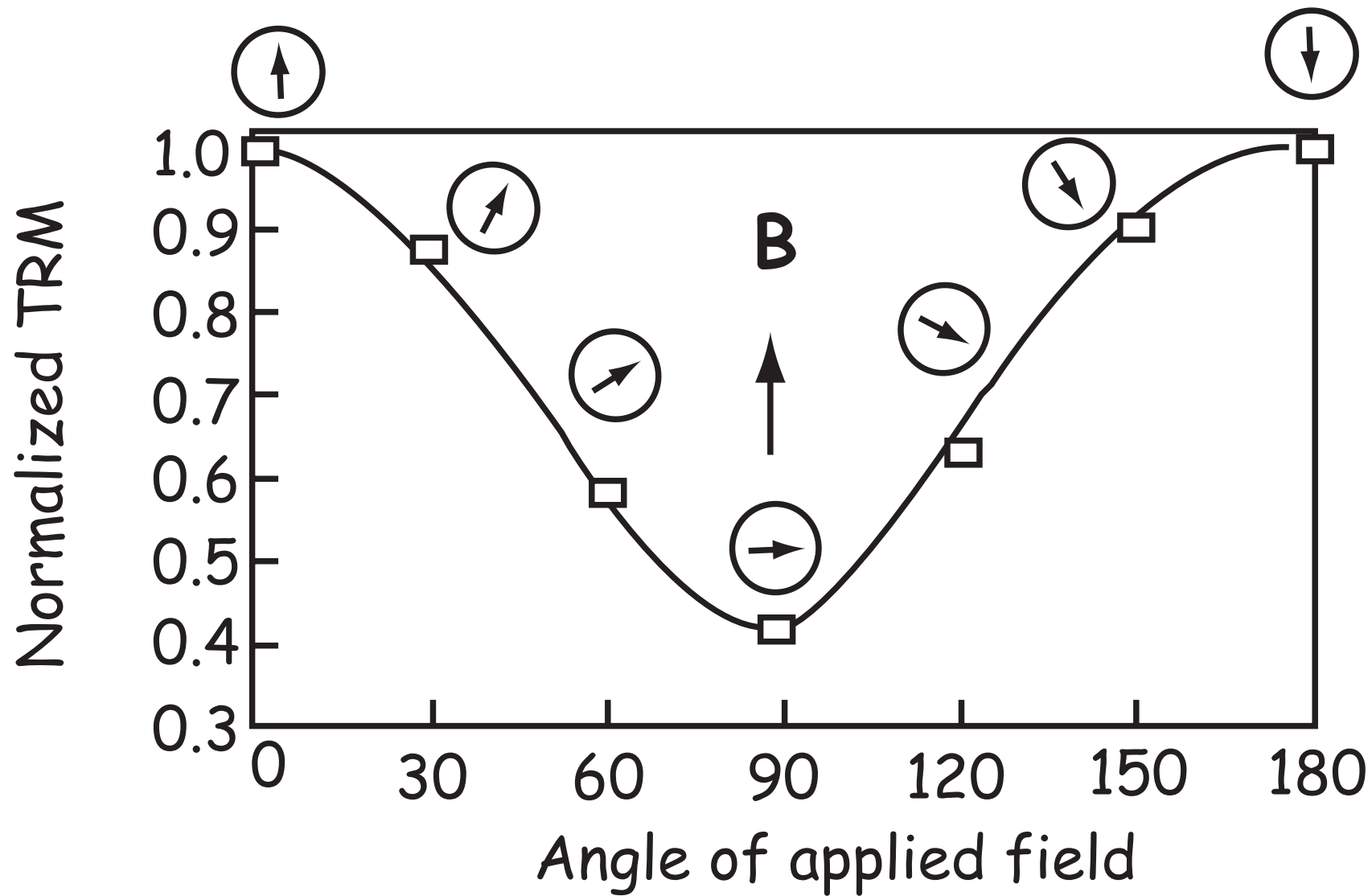
Some comments on AMS statistics

- There are many ways of repackaging the eigenvalues into anisotropy statistics (see Table 13.1)
- Typical examples boil the eigenvalues into an anisotropy factor (P or P') and one that reflects shape (e.g., shape factor, T)
- this allows only “prolate” and “oblate” – what about spherical and triaxial? you lose info
- also none incorporate any uncertainty estimate

Anisotropy of magnetic remanence

- Different than susceptibility because can measure three components at a time - not just one.
- So degrees of freedom are calculated differently ($nf = 3 \times N_{\text{meas}} - 6$ instead of $N_{\text{meas}} - 6$)
- Useful for:
 - correcting absolute paleointensity data for anisotropy of TRM
 - correcting sedimentary data for inclination error

Remember from Chapter 10
that TRM can be anisotropic:



Analogous to induced magnetization and susceptibility:

$$\mathbf{M}_I = \chi \mathbf{H}$$

Remanent vectors can be related to the applied field by:

$$\mathbf{M}_R = \chi^R \mathbf{H}$$

Where χ^R is the anisotropy of remanence tensor

Knowing that, we could get:

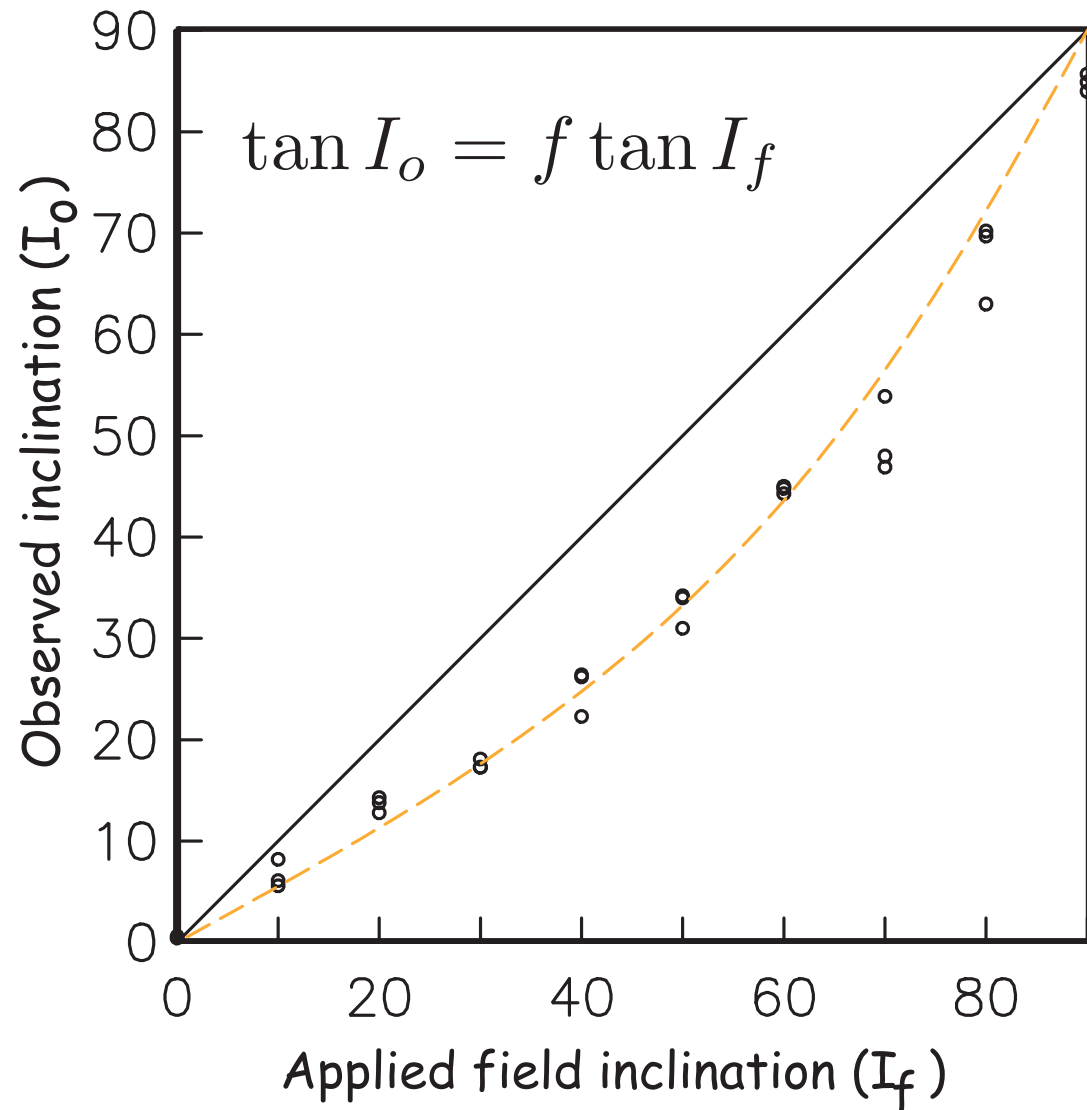
$$\mathbf{H}_{anc} = \mathbf{M} \cdot \chi_R^{-1}$$

(see Chapter 13 for tricky bits regarding normalization)

TRM anisotropy in practice

- Give specimens total TRM in at least 3 (usually 6 or 9 for uncertainties) directions.
- Calculate remanence anisotropy tensor using the design matrix for your protocol.
- Because each additional heating step can produce additional alteration, some substitute ARM anisotropy for TRM anisotropy
- With ARMs, you have to AF demagnetize completely between each ARM acquisition step

Remember
inclination
“error” (Chapter 7)?



DRM anisotropy may help us to correct for this

As before: $\mathbf{M}_{DRM} = \chi^{DRM} \mathbf{H}$

But in the book, we use terminology of Jackson et al. (1991) or: $\mathbf{M}_d = \mathbf{k}_d \mathbf{H}$

DRM anisotropy tensor has eigenvalues:

$$\kappa_{d1} > \kappa_{d2} > \kappa_{d3}$$

flattening factor, f , gotten by: $f = \frac{\kappa_{d3}}{\kappa_{d1}}$

so all we need is the DRM anisotropy tensor!

But, this is not so easy.....

How to get DRM anisotropy tensor:

- Redeposition? NOT
- Measure proxy? ARM?
 - needs to reflect same grains as those carrying the DRM
 - needs to take into account “particle anisotropy”
 - what about the effect of flocculation?
 - for details see Chapter 13 and Appendix D.3

Assignment

- Problem 13.3