

Parte 3

Determination of the coefficients $C_l^{\mu\nu}$

pole figures $P_{\pi_i}(\bar{y})$ are developed in a series of spherical surface harmonics.

$$P_{\pi_i}(\bar{y}) = \sum_{l=0}^{\infty} \sum_{\nu=1}^{M(l)} F_l^{\nu}(\pi_i) \hat{P}_l^{\nu}(\bar{y}) \quad [21]$$

with the coefficients:

$$F_l^{\nu}(\pi_i) = \frac{4\pi}{(2l+1)} \sum_{\mu=1}^{M(l)} C_l^{\mu\nu} \hat{P}_l^{i*\mu}(\pi_i) \quad [22]$$

Coefficients $F_l^{\nu}(\pi_i)$ are known for a certain number I_p of pole figures

→ [22] represents a system of I_p linear equations with $M(l)$ unknowns.
(only even values of l)

In order to have a (unique) solution:

$$M(l) \leq I_p$$

If there are more equations than unknowns
→ it can occur there is no solution?

However! physics of the phenomenon

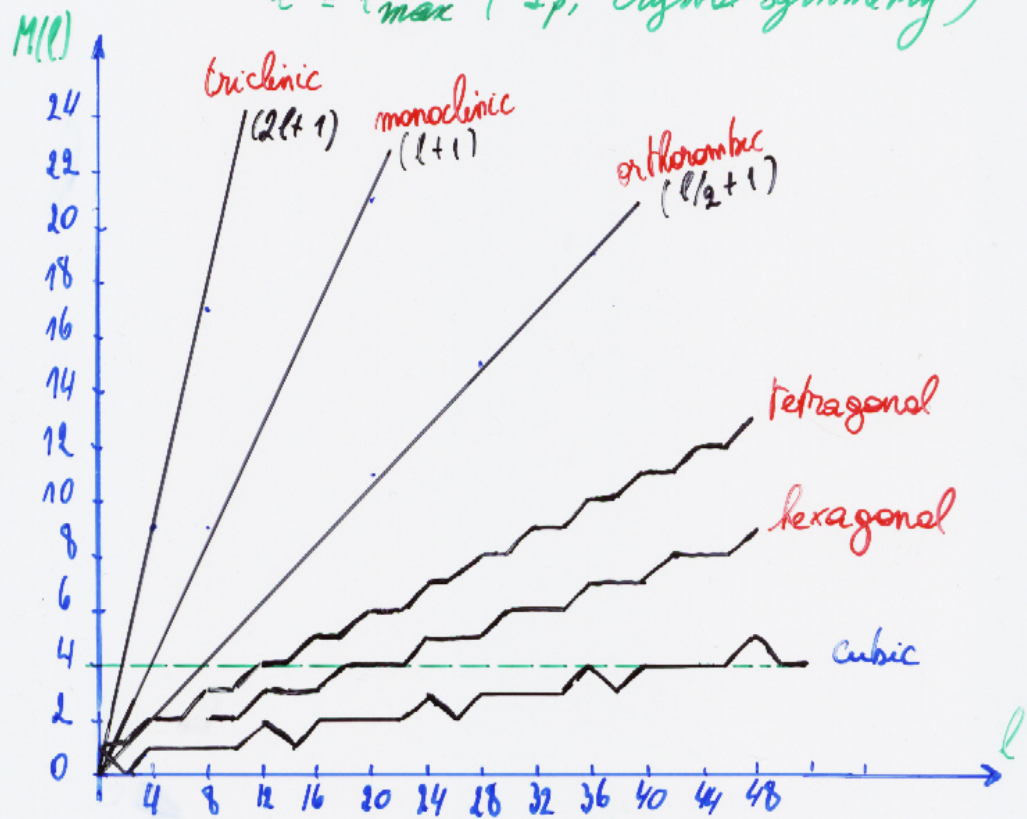
→ $F_l^{\nu}(\pi_i)$ must be conditioned such that Eq [22] has a unique solution because some ODF $f(\bar{g})$ and thus a set of C coeff^{ts} must exist.

→ if Eq. [22] has more equations than unknowns \Rightarrow equations are not linear independent \Rightarrow \neq eq's can be reduced \rightarrow unique solution.

$M(l)$ is increasing function of l (depending on crystal symmetry)

\Rightarrow For a given number of pole figures the system of eq's [22] can only be solved uniquely up to a certain degree

$$l = l_{\max} (I_p, \text{crystal symmetry})$$

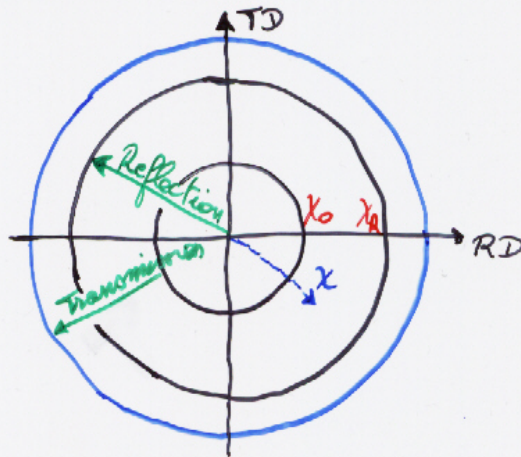


l_{\max} = the resolving power

Determination of the coefficients $C_p^{\mu\nu}$ from incompletely measured pole figures.

In practice (e.g. due to the defocussation effect) pole figures can only be measured over a limited range: $\chi_{\min} < \chi < \chi_{\max}$

(for the backreflection method with X-ray diffraction: $\chi_{\min} = 0$ and $\chi_{\max} \approx 80^\circ$).



Suppose: pole figure can only be measured over an area B for which: $\chi_0 < \chi < \chi_a$

→ previous method for calculating coeff⁰ can no longer be used because orthogonality relations are no longer valid:

$$\oint \hat{e}^{\mu} \hat{e}^{\nu} (\bar{y}) d\bar{y} = \delta_{\mu\alpha} \delta_{\nu\beta}$$

↳ integration over the entire surface of the pole figure.

Alternative method:

Suppose $C_l^{\mu\nu}$ are known

→ we can calculate pole fig. $P_{ti}(\bar{y})_{cal}$ using

$$P_{ti}(\bar{y})_{cal} = \int_{\ell, \nu} \left[\frac{4\pi}{2\ell+1} \sum_{\mu} C_l^{\mu\nu} \hat{p}_\ell^{j\mu}(t_i) \right] \hat{p}_\ell^{\nu}(\bar{y}) \quad [23]$$

Suppose: measured (and normalized) pole figures $P_{ti}(\bar{y})_{obs}$

Now, we set as a condition:

$$\sum_i w_i \int_B [P_{ti}(\bar{y})_{obs} - P_{ti}(\bar{y})_{cal}]^2 d\bar{y} = \min \quad [24]$$

w_i = weight factors which reflect the accuracy of the pole figure measurement.

In general: measured pole figure
→ unknown normalization

$$\hat{P}_{ti}(\bar{y})_{obs} = \frac{1}{N_i} P_{ti}(\bar{y})_{obs} \quad [25]$$

with N_i = the unknown normalization factor.
Substituting eq. [23] in condition [24], considering eq. [25]

$$\sum_i w_i \int_B \left[N_i \hat{P}_{ti}(\bar{y})_{obs} - \int_{\ell, \nu} \frac{4\pi}{2\ell+1} C_l^{\mu\nu} \hat{p}_\ell^{j\mu} \hat{p}_\ell^{\nu}(\bar{y}) \right]^2 \times d\bar{y} = \min \quad [26]$$

Expression [26] must be minimized w.r.t. the unknowns $C_l^{\mu\nu}$ and M_i

→ differentiation w.r.t. $C_l^{\mu\nu}$

$$\sum_i \omega_i \int_B h_l^{i+\mu}(h_i) h_{l'}^{i+\nu}(\bar{y}) \quad [27]$$

$$\times \left[M_i \bar{P}_{h_i}(\bar{y})_{obs} - \sum_{l, l'} \frac{4\pi}{2l+1} C_l^{\mu\nu} h_l^{i+\mu}(h_i) h_{l'}^{i+\nu}(\bar{y}) \right] d\bar{y} = 0$$

→ differentiation w.r.t. M_i :

$$\int_B \bar{P}_{h_i}(\bar{y}_{obs}) \left[M_i \bar{P}_{h_i}(\bar{y})_{obs} - \sum_{l=0}^L \sum_{\mu=1}^{M(l)} \sum_{\nu=1}^{M(l)} \frac{4\pi}{(2l+1)} C_l^{\mu\nu} h_l^{i+\mu}(h_i) h_{l'}^{i+\nu}(\bar{y}) \right] d\bar{y} = 0 \quad [28]$$

We introduce the following quantities:

$$\int_B [\bar{P}_{h_i}(\bar{y})_{obs}]^2 = P_i$$

$$h_l^{i+\mu}(h_i) \int_B \bar{P}_{h_i}(\bar{y})_{obs} h_{l'}^{i+\nu}(\bar{y}) d\bar{y} = a_l^{i+\nu}(h_i)$$

$$\int_B h_l^{i+\nu}(\bar{y}) h_{l'}^{i+\nu'}(\bar{y}) d\bar{y} = \xi_{ll'}$$

$$\sum_i \omega_i \frac{4\pi}{2l+1} h_l^{i+\mu}(h_i) h_{l'}^{i+\mu'}(h_i) = \alpha_{ll'}$$

⇒ the equations [27] and [28] can be written as

$$\sum_{l=0}^L \sum_{\mu=1}^{H(l)} \sum_{\nu=1}^{N(l)} C_l^{\mu\nu} \sum_{l'l'} \sum_{\mu'\nu'} a_{l'l'}^{\mu\nu} = \sum_i \omega_i N_i a_{l'}^{\mu\nu}(k_i) \quad [29]$$

$$\sum_{l=0}^L \sum_{\mu=1}^{H(l)} \sum_{\nu=1}^{N(l)} C_l^{\mu\nu} \frac{4\pi}{2l+1} a_l^{\mu\nu}(k_i) = N_i P_i \quad [30]$$

If the normalization factor N_i is expressed according to eq. [30] and substituted in eq. [29]

$$\sum_{l=0}^L \sum_{\mu=1}^{H(l)} \sum_{\nu=1}^{N(l)} C_l^{\mu\nu} \left[\sum_{l'l'} \sum_{\mu'\nu'} a_{l'l'}^{\mu\nu} - \frac{4\pi}{2l+1} \sum_i \frac{\omega_i}{P_i} a_l^{\mu\nu}(k_i) a_{l'}^{\mu'\nu'}(k_i) \right] = 0 \quad [31]$$

This is a system of linear eq^s with as many eq^s as unknowns.

Eq. [31] is a homogeneous system \Rightarrow the coeff^s are only determined up to a common factor.
 \rightarrow can be derived from the normalization cond.

$$C_0^{11} = 1$$

The texture index

$$J = \oint [f(g)]^2 dg = \text{texture index}$$

Because of the orthogonality rules:

$$J = \sum_{l,\mu,\nu} \frac{1}{(2l+1)} |C_l^{\mu\nu}|^2$$

For a random texture: $J_r = 1$

For a single crystal texture: $J_{g_0} \rightarrow +\infty$

Ghost Correction.

Pole figures are independent of $C_l^{\mu\nu}$ coeff^{ts} of odd (uneven) rank because of the centro-symmetry property of pole figures.

⇒ classical pole figure inversion only yields even $C_l^{\mu\nu}$ coeff^{ts}.

However, it does NOT mean that the uneven C coeff^{ts} are zero!

$$f(g) = \sum_{l=0}^{\infty} \sum_{\mu=1}^{M(l)} \sum_{\nu=1}^{M(l)} C_l^{\mu\nu} j_l^{\mu\nu}(g) \quad [1]$$

If uneven C coeff^{ts} are ignored
→ $f(g) =$ the reduced ODF.

The reduced ODF may contain peaks (positive + negative) which do not represent real existing texture comp^{ts}
⇒ GHOST peaks.

In literature: various ghost correction procedures:

- positivity method (Bunge and Delmas)
- quadratic method (Van Houtte)
- exponential method (Van Houtte).

All these methods use the positivity property $f(g) \geq 0$ of the texture function.

The ODF is a statistical distribution function
→ physical it must be non-negative everywhere.

Construction of a general algorithm.

Suppose: $f_{\text{true}}(g)$ = the true texture function.
In reality we only know an approximation $f(g)$ of $f_{\text{true}}(g)$.

↳ represented by a finite nr. of parameters such as:

1] $C_{\mu\nu}$ coeff^{PS}

2] $f_{\text{true}}(g)$ on a 5° grid in Euler space.

Suppose:

x_i ($i=1, \dots, n$) is one set of values to

y_j ($j=1, \dots, m$) is another set. ^{reconstruct $f_{\text{true}}(g)$}

Obviously, there must exist a relation between both sets of parameters:

$$y_j = Y_j(x_i) \text{ and } x_i = X_i(y_j)$$

Assume x_i 's are known, and y_j 's not but we know y_{0j} as a first estimate for y_j .

$$\Rightarrow x_{0i} = X_i(y_{0j}) \quad [2]$$

We can now expand the function $Y_j(x_i)$ as a Taylor series:

$$Y_j(x_i) = y_{0j} + \sum_{i=1}^n \frac{\partial Y_j(x_{0k})}{\partial x_i} (x_i - x_{0i}) + \dots \quad [3]$$

If we neglect the higher order terms:
 → a new estimate of y_j can now be made:

$$y_{1j} = y_{0j} + \sum_{i=1}^n \frac{\partial Y_j(x_{0k})}{\partial x_i} (x_i - x_{0i}) \quad [4]$$

The old y_{0j} are replaced by the new y_{1j} and the algorithm can be executed again starting with Eq. [2].

These algorithms can only be used if the functions $X_i(y_j)$ are known and if the values $\frac{\partial Y_j(x_{0k})}{\partial x_i}$ can be evaluated.

The positivity method.

The two parameter sets are chosen as follows:

$$\begin{aligned} y_j &= \text{odd C-coeff}^{\text{ts}} \quad (\# = m) \\ x_i &= \text{values of the complete ODF on a } 5^\circ \\ &\quad \text{grid in Euler space. } (\# = n) \end{aligned}$$

The function $X_i(y_j)$ is given by:

$$x_i = X_i(y_j) = \tilde{f}(g_i) + \sum_{l=1}^L \sum_{\mu, \nu} \tilde{f}^{\mu\nu}(g_i) y_j \quad [5]$$

with g_i = crystal orientations at the grid points
 and \tilde{f} = the reduced ODF (even C-coeff^{ts})

Relation between C coeff^{ts} and $f(g)$:

$$C_l^{\mu\nu} = (2l+1) \oint f(g) \ddot{T}_l^{\mu\nu}(g) dg.$$

In a discrete way:

$$C_l^{\mu\nu} = (2l+1) \sum_{i=1}^n w_i f(g_i) \ddot{T}_l^{\mu\nu}(g_i) \quad [6]$$

with w_i = weight factors suitable for numerical integration.
With above conventions for x_i and y_j :

$$y_j(x_i) = (2l+1) \sum_{i=1}^n w_i \ddot{T}_l^{\mu\nu}(g_i) x_i \quad [7]$$

$$\Rightarrow \frac{\partial y_j}{\partial x_i} = (2l+1) w_i \ddot{T}_l^{\mu\nu}(g_i) \quad [8]$$

The initial estimate y_{0j} of the odd C-coeff^{ts} = 0
 \Rightarrow the x_{0i} coeffs = the reduced ODF \tilde{f}

In certain negative regions (i.e. regions for which $\tilde{f}(g_i) \leq 0$) \rightarrow the true value should be zero.

\Rightarrow we assume:

$$[7 \text{ bis}] \quad x_i = x_{0i} = \tilde{f}(g_i) \text{ if } x_{0i} \geq \alpha \quad \rightarrow \text{range } Z_+ \\ x_i = 0 \text{ if } x_{0i} \leq \alpha \quad \rightarrow \text{range } Z_-$$

Using eq. [4] and eq. [8]

$$y_{1j} = y_{0j} - (2l+1) \sum_{g \in Z_-} w_i \ddot{T}_l^{\mu\nu}(g_i) x_{0i}$$

- In the next iteration step \rightarrow values y_{0j} are substituted by values y_{1j} in eq. [4] and eq. [5] is used to calculate the new x_{0i} being x_{1i} , whereas [7bis] is used to find the new values for x_i .
- The algorithm can be repeated as often as necessary. At each stage x_{0i} is the current estimate of the complete ODF.
- Convergence can be monitored by some error criterion:

$$-\int_{\Sigma} f(g) < \epsilon$$

ϵ \rightarrow small positive value.

Exponential Method.

Idea to produce a strictly non-negative ODF with C-coeffts $C_{\mu\nu}$.

\rightarrow even C-coeffts should be equal to those obtained after classical pole figure inversion $C_{\mu\nu}$

$$[8] \quad \Delta = \sum_{l=0(2)}^L \sum_{\mu, \nu} \frac{1}{(2l+1)} [C_{\mu\nu}^{sl} - C_{\mu\nu}^l]^2 \rightarrow 0$$

Strictly positive ODF: $f_s(g) = e^{h(g)}$ [9]

Choice of parameter sets:

$y_j = h(g_j)$ on grid points g_j in Σ . S.

$x_i = C_{\mu\nu}^{sl}$ (even + unaven).

y_{0j} is estimated from other method (e.g. positivity).

$$X_i(y_j) = (2l+1) \sum_{j=1}^m w_j \dot{T}_l^{\mu\nu}(g_j) e^{y_j}$$

and using eq. [4] and eq. [9]

$$\Rightarrow e^{y_j} = \sum_{l=0}^L \sum_{\mu, \nu} x_i \dot{T}_l^{\mu\nu}(g_j)$$

i corresponds to a comb. of l, μ, ν

Differentiation w.r.t. x_i

$$e^{y_j} \frac{\partial y_j}{\partial x_i} = \dot{T}_l^{\mu\nu}(g_j)$$

$$\frac{\partial y_j}{\partial x_i} = e^{-y_j} \dot{T}_l^{\mu\nu}(g_j)$$

The general algorithm can now be applied with eq. [8] as convergence criterion.

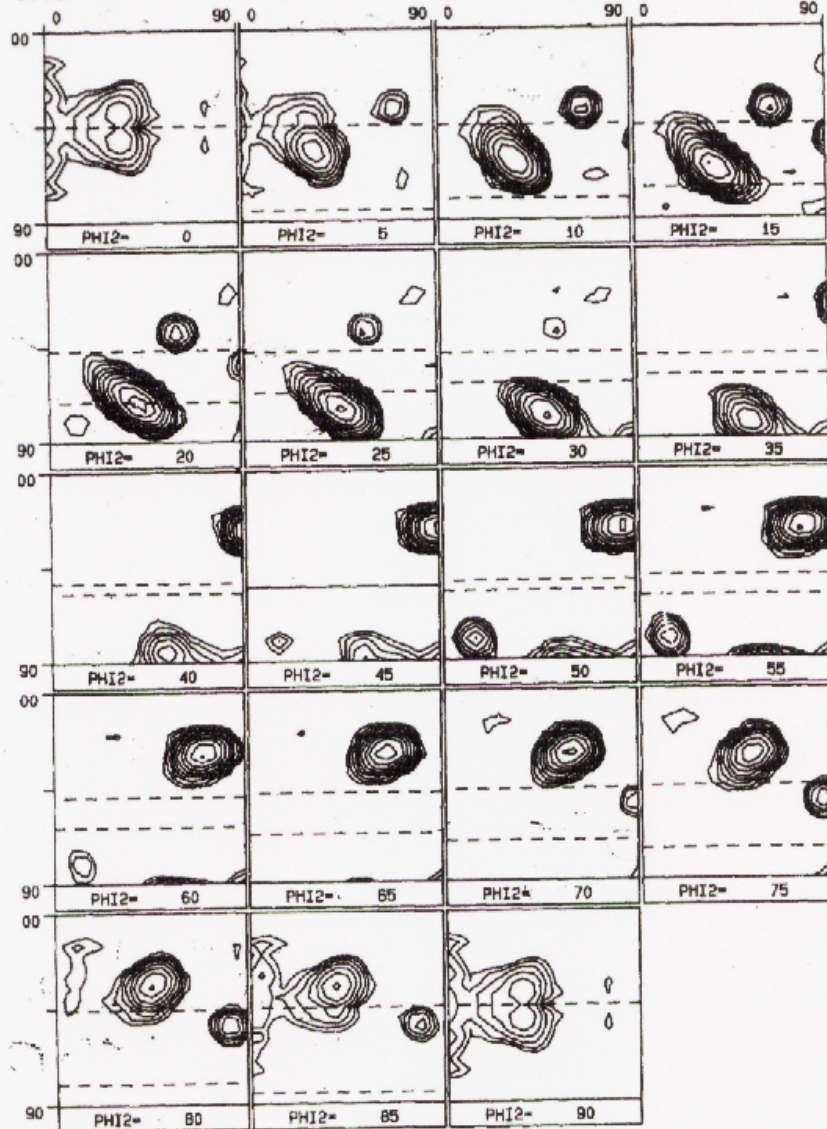
Texture before ghost correction

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Texture after ghost correction

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Discretization of continuous ODFs.

Literature: L.S. Tóth and P. Van Houtte,
Textures and Microstructures, Vol. 19, pp. 229-244.

Approximation of the ODF by a discrete set of orientations is required in many modelling studies of texture dependent physical properties.

1. Discretization on the basis of a random set of orientations.

Suppose: g_i = set of N orientations randomly distributed in orientation space (N can be chosen freely).

With each orientation a volume v_i of Euler space is associated such that:

$$\sum_{i=1}^N v_i = \text{the total volume of E.S.} \\ = 1 \quad (\text{if the Bunge normalization is accepted})$$

$f(g)$ is specified by its set of $C_l^{\mu\nu}$ coeffs

$$\Rightarrow f(g_i) = \sum_{l, \mu, \nu}^{L, M, N} C_l^{\mu\nu} \frac{j_l^{\mu\nu}(g_i)}{j_l^{\mu\nu}}$$

can be calculated for any arbitrary orientation.

and $v_i f(g_i) =$ an approximation of the volume fraction of orientation g_i

⇒ The set of orientations g_i with volume fractions w_i given by

$$w_i = v_i f(g_i)$$

can be regarded as a discrete representation of $f(g)$.

2. Discretisation on the basis of the "cumulative" ODF statistical technique.

For some applications it is desirable to produce a discrete grain orientation distribution in which each grain has the same volume fraction.

↳ can be obtained by employing the cumulative distribution function.

The cumulative ODF is defined by:

$$F(G) = \int_{G_0}^G f(g) dg \quad [4]$$

with $G =$ an arbitrary crystal orientation the integral follows an arbitrary integration path that covers the whole orientation space.

The integration of eq. [1] will be carried out in a discrete way over a number of boxes in Euler space.

Suppose, we use the 5° -grid in E.S.

→ partitions E.S. in N boxes with sequence numbers i ($i=1, \dots, N$).

The centre of such a box is orient. g_i .
→ the discrete version of integral [1] is given by:

$$F_j = \sum_{i=1}^N f_i \quad [2]$$

with $f_i =$ the integral of $f(g)$ over box i :

$$f_i = \int_{\text{box } i} f(g) dg.$$

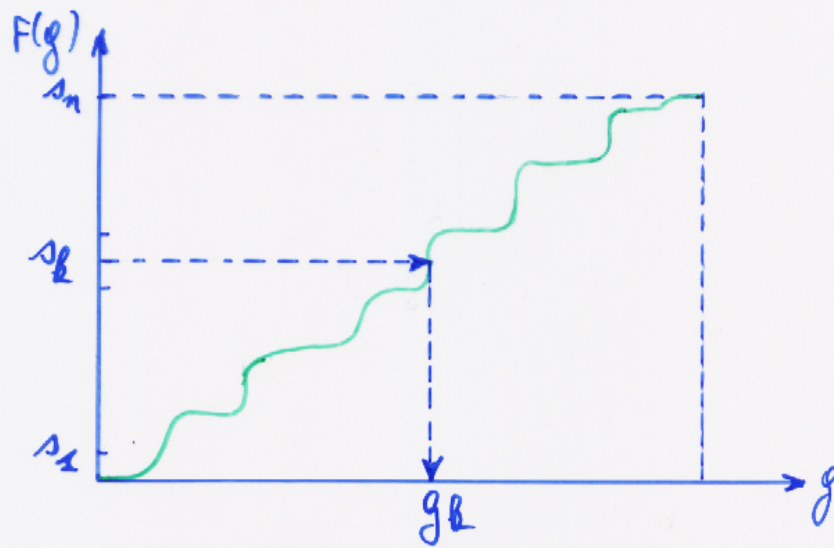
and because of the normalization condition

$$\sum_{i=1}^N f_i = 1$$

Eq. [2] defines a kind of "staircase" function.

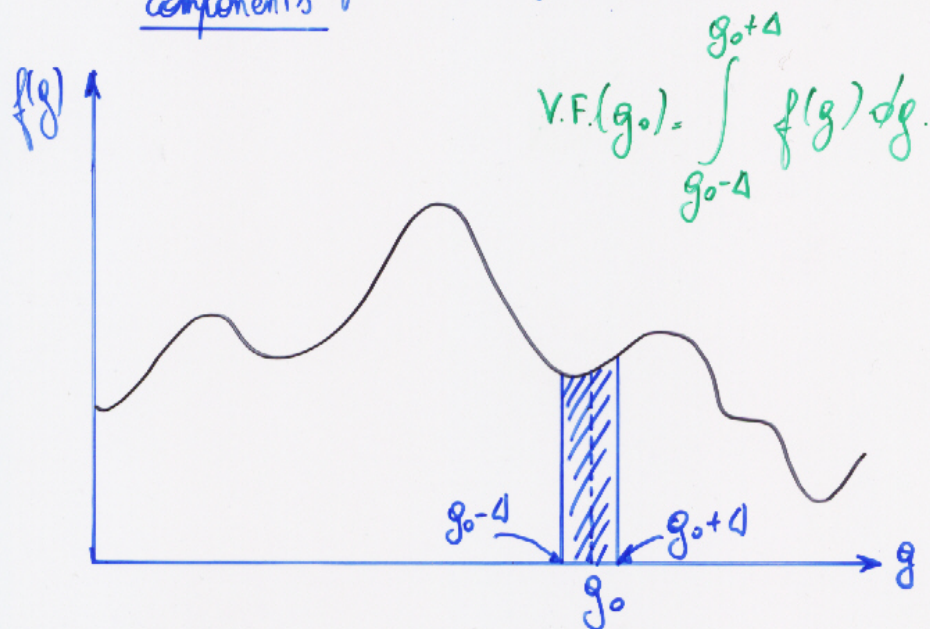
A discrete set of orientations is produced in the following way:

- 1) a set of n number ("selectors") is generated in a range between 0 and 1



- 2) One discrete orientation will be generated for each selector by using the inverse function of the staircase function.

The volume fraction of individual texture components



How to chose Δ ?

Method, proposed by Bunge :

- Consider a Gaussian spread function around g_0 with max S and spread ϕ :

$$\omega_{g_0}(g) = S \exp \left[-(\phi/\phi_0)^2 \right]$$

with ϕ_0 = the angular width

(typical $\phi_0 = 16.5^\circ$ or 11°).

- The volume fraction $VF(g_0)$ of the comp^t g_0 is estimated by :

$$V.F(g_0) = \frac{1}{S} \phi \omega_{g_0}(g) f(g) dg.$$

(cfr. Bunge, "Texture Analysis in Materials Science", p. 171).

Disadvantage of this method

- 1] Vol. fractions are not normalized to 1 (there is always the contribution from the tail of the Gaussian spread function)
- 2] Strong dependence from the choice of ϕ_0 .

Modelling of texture dependent physical properties!

Each orientation dependent physical property of the single crystal



Texture dependent physical property of the polycrystal!

Suppose: E = an orientation dependent physical property.

$$E = E(g).$$

- e.g. E = elastic modulus
= magnetization energy of a cubic crystal
= thermal expansion of non-cubic crystals
=

Most often: E is a tensorial property \bar{E}

\bar{E} will exhibit the symmetry of both crystal and sample

$$\Rightarrow E(g) = \sum_{l=0}^L \sum_{M,N} \frac{M,N}{M,N} e_l^{MN} \bar{e}_l^{MN}(g)$$

$E(g)$ can be expanded in a series of symmetrical spherical harmonics.

with $c_l^{\mu\nu}$ given by:

$$c_l^{\mu\nu} = (2l+1) \int E(g) \hat{T}_l^{\mu\nu}(g) dg.$$

If mutual grain interaction can be ignored
→ the polycrystalline average of $E(g)$
of a material with ODF $f(g)$ is given
by:

$$\bar{E} = \int f(g) E(g) dg.$$

$$\bar{E} = \frac{1}{(2l+1)} \sum_{l=0}^L \frac{1}{(2l+1)} \sum_{\mu, \nu}^{M, N} c_l^{\mu\nu} c_l^{\mu\nu}$$

Example: the elastic anisotropy.

Problem: How to obtain the effective elastic moduli of a polycrystalline material from the single crystal ones, taking the texture into account

For a single crystal: Hooke's law is formulated as follows:

$$\epsilon_{ij} = s_{ijkl} \sigma_{kl}$$

$$\text{or } \sigma_{ij} = c_{ijkl} \epsilon_{kl}$$

s_{ijkl} = elastic compliance tensor
 c_{ijkl} = " stiffness tensor.

$$[s_{ijkl}] = [c_{ijkl}]^{-1}$$

These tensors depend on the crystallite orientation g

→ s_{ijkl} and c_{ijkl} are expressed in the sample ref. system x_1, x_2, x_3 .

Literature data on stiffness or compliance are normally expressed in the cryst. ref. system x_1^c, x_2^c, x_3^c → s_{ijkl}^c, c_{ijkl}^c

According to the tensor transformation laws:

$$s_{ijkl} = a_{im} a_{jn} a_{lo} a_{lp} s_{mnop}^c$$

According to Bunge the product $a_{im} a_{jn} a_{lo} a_{lp}$ can be expressed by means of a series expansion of harmonic functions in which $L_{max} = 4$.

→ s_{ijkl} (and c_{ijkl}) can be written as a series expansion of order 4.

The polycrystalline averages of stiffness and compliance tensors are given by:

$$\bar{s}_{ijkl} = \int s_{ijkl}(g) f(g) dg =$$