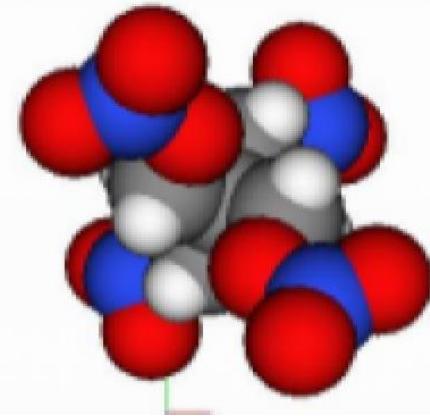


THE RIETVELD REFINEMENT METHOD IN GSAS-II

GSAS-2



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Acknowledgements: DOE/SC

HISTORY – H.M. RIETVELD

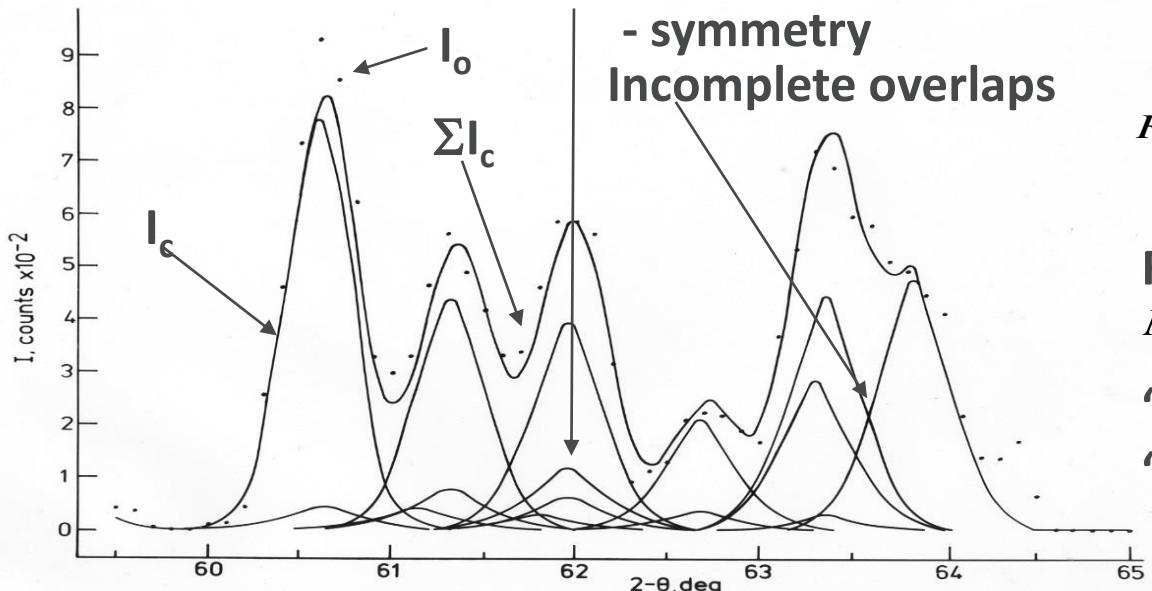


Hugo Rietveld; neutron powder diffractometer,
Petten, Netherlands

Papers: H.M. Rietveld, *Acta Cryst.* 22, 151-
2(1967)

H.M. Rietveld, *J. App. Cryst.*, 2, 65-71 (1969)
Multi-parameter, nonlinear LS curve fitting

Exact overlaps



Residuals:

$$R_{wp} = \sqrt{\frac{\sum w(I_o - I_c)^2}{\sum wI_o^2}}$$

Rietveld Minimize

$$M_R = \sum w(I_o - I_c)^2$$

“chi-squared” or
“goodness-of-fit”

$$\chi^2 = M_R/(n-p)$$

LINEAR LEAST SQUARES THEORY

Given a set of observations I_{obs}

and a function $I_{calc} = f(p_1, p_2, p_3, \dots, p_n)$

then the best estimate of the values p_i is found by
minimizing

$$M = \sum w(I_o - I_c)^2$$

This is done by setting the derivative to zero

$$\sum w(I_o - I_c) \frac{\partial I_c}{\partial p_j} = 0$$

Results in n “normal” equations (one for each variable) -
solve for p_i

NON-LINEAR LEAST SQUARES THEORY

Problem - $I(p_i)$ is nonlinear & transcendental (sin, cos, etc.)
so can't solve directly

Expand $I(p_i)$ as Taylor series & toss high order terms

$$I_c(p_i) = I_c(a_i) + \sum_i \frac{\partial I_c}{\partial p_i} \Delta p_i$$

a_i - initial values of p_i
 $\Delta p_i = p_i - a_i$ (shift)

Normal equations - one for each Δp_i

$$\sum w \left[\Delta I - \sum_i \frac{\partial I_c}{\partial p_i} \Delta p_i \right] \frac{\partial I_c}{\partial p_j} = 0 \quad \Delta I = I_o - I_c(a_i)$$

Outer sum over observations

Solve for Δp_i - shifts of parameters, NOT values

LEAST SQUARES THEORY - CONTINUED

Rearrange

$$\sum w \frac{\partial I_c}{\partial p_1} \left(\sum_{i=1}^n \frac{\partial I_c}{\partial p_i} \Delta p_i \right) = \sum w \Delta I \frac{\partial I_c}{\partial p_1}$$

⋮
⋮
⋮

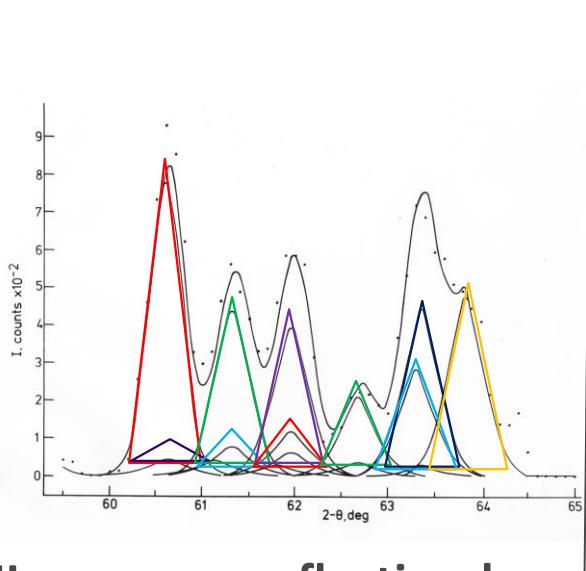
$$\sum w \frac{\partial I_c}{\partial p_n} \left(\sum_{i=1}^n \frac{\partial I_c}{\partial p_i} \Delta p_i \right) = \sum w \Delta I \frac{\partial I_c}{\partial p_n}$$

Matrix form: $\mathbf{Ax}=\mathbf{v}$

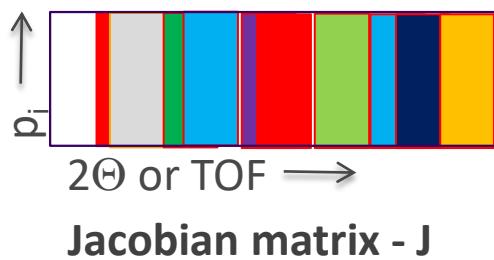
$$a_{i,j} = \sum w \frac{\partial I_c}{\partial p_i} \frac{\partial I_c}{\partial p_j} \quad x_j = \Delta p_j \quad v_i = \sum w(\Delta I) \frac{\partial I_c}{\partial p_i}$$

Solve: $\mathbf{x} = \mathbf{A}^{-1}\mathbf{v} = \mathbf{B}\mathbf{v}$; $\mathbf{B} = \mathbf{A}^{-1}$ This gives set of Δp_i to apply to “old” set of a_i ; repeat until Δp_i small.

GSAS-II ALGORITHM

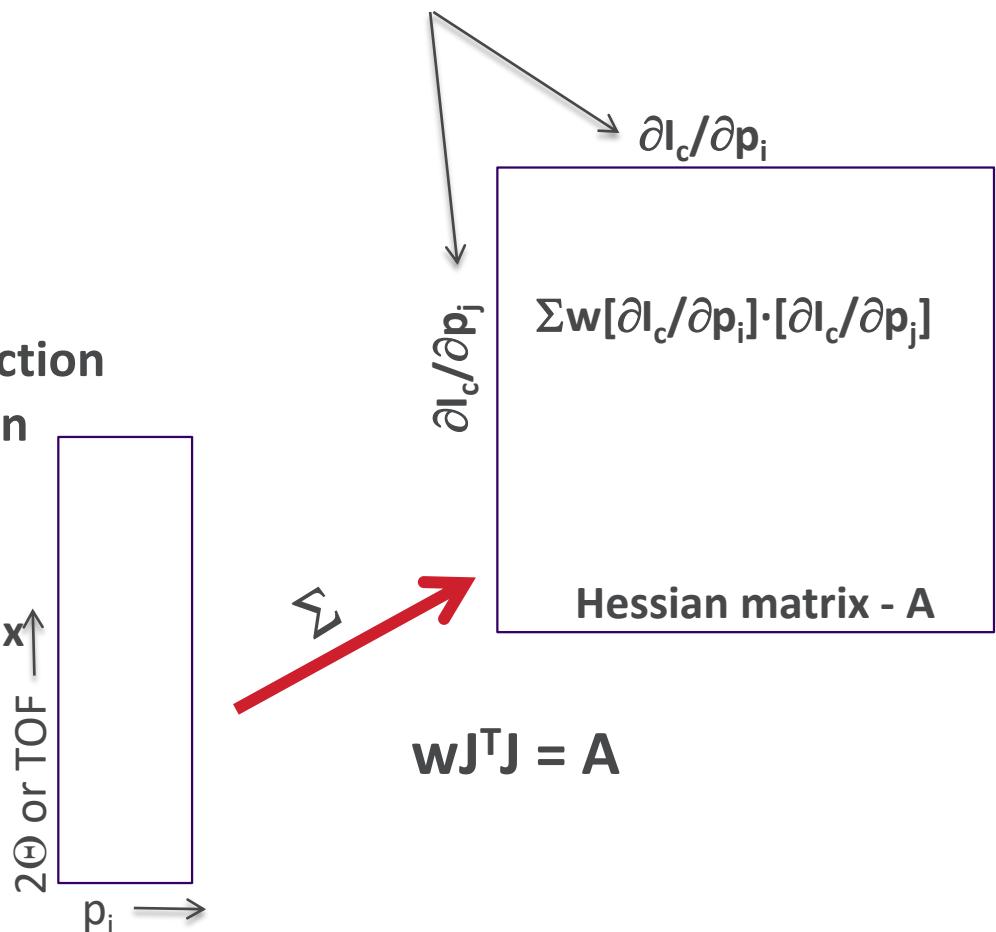


GSAS-II – process reflection-by-reflection
to make \mathbf{I}_c (vector) & $\partial\mathbf{I}_c/\partial\mathbf{p}_i$ (Jacobian
matrix)



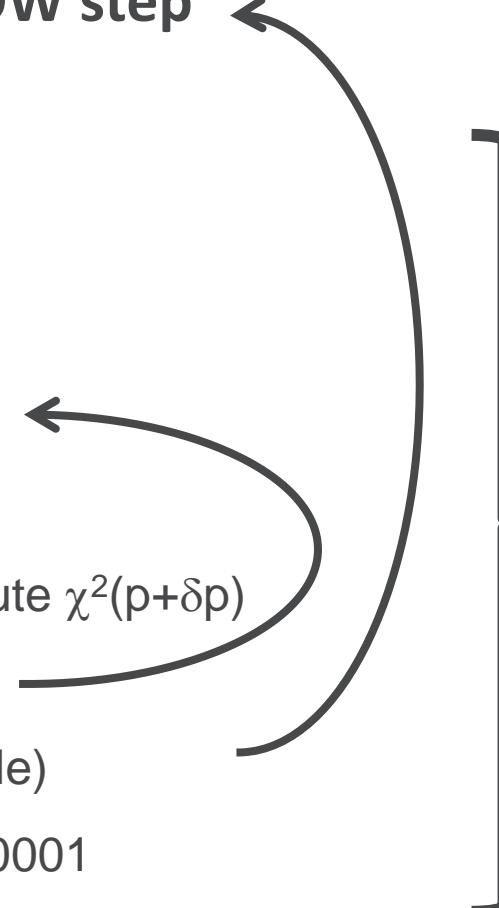
NB: GSAS-II – needs large memory!

GSAS – process point-by-point to make
 \mathbf{I}_c (value) & $\partial\mathbf{I}_c/\partial\mathbf{p}_i$ (vector)



REFINEMENT VIA MODIFIED LEVENBERG/MARQUARDT-SVD ALGORITHM

Steps:

1. Compute $A_{ij} = \sum w \frac{\partial I_c}{\partial p_i} \frac{\partial I_c}{\partial p_j}$ **SLOW step**
 2. Normalize $A'_{ij} = A_{ij} / \sqrt{A_{ii} A_{jj}}$
 3. compute $\chi^2(p)$
 4. Select $\lambda (=0.001, \text{"damping factor"})$
 5. Modify $A''_{ii} = A'_{ii}(1 + \lambda)$
 6. Make SVD inversion of A''
 7. Solve for δp (unnormalized!) & compute $\chi^2(p+\delta p)$
 8. If $\chi^2(p+\delta p) > \chi^2(p)$ then $\lambda * 10$ go to 5
 9. Else apply δp to p & go to 1 (new cycle)
 10. Quit when $\chi^2(p) - \chi^2(p+\delta p) / \chi^2(p) < 0.0001$
- 

NB: all in ~40 lines of python; all double precision

NB²: this thing is exceedingly robust – no user damping factors needed

SVD – SINGULAR VALUE DECOMPOSITION

Singularities & near singularities – see Mathematical Recipes 2.9

LS matrix: solve for x $Ax=b$ by $x=A^{-1}b$; x are the parameter shifts

SVD: replace $A = UwV$ where U & V are such that $U^{-1} = U^T$ & $V^{-1} = V^T$

& w – diagonal matrix; all same size as A

Then: $A^{-1} = V(1/w_{ii})U^T$

The trick: what to do if $w_{ii} \sim 0$? (singularity) \rightarrow make $1/w_{ii} = 0$! (instead of ∞)

Then: $x = V(1/w_{ii})U^Tb$ does away with ill-conditioned terms

Have to choose tolerance on $w_{ii} \sim 0$ (typically 10^{-6} but 10^{-3} for proteins works well)

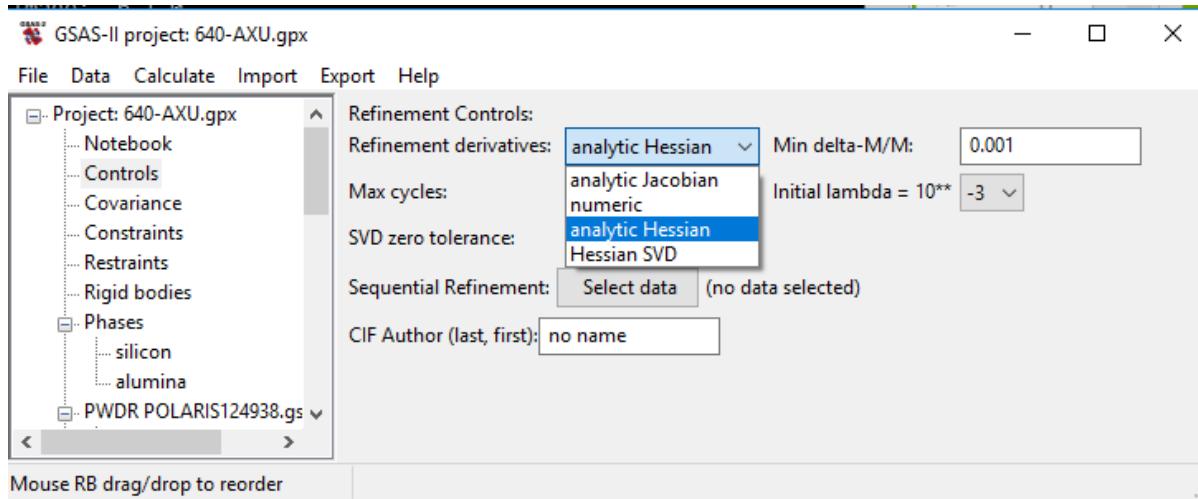
SVD is in python library as `numpy.linalg.svd`

& uses LAPACK _gesdd routine (fortran – code in MR 2.9)

NB: all double precision in python; downside is w_{ii} not 1:1 to parameters so id of failures difficult.

LEAST SQUARES ALGORITHMS IN GSAS-II

Useful choices – found in Controls



Analytic Hessian – **default** Levenberg-Marquardt SVD from Hessian & computed derivatives
Downside: hard singularities hard to find → “linear algebra errors” cause failures

Analytic Jacobian – uses Jacobian matrix (not Hessian) no SVD; identifies singularities &
Removes them from LS refinement; always runs to convergence

Hessian SVD – no Levenberg-Marquardt (might be better for single crystal data)
Same downside as Analytic Hessian

Numeric – no derivatives & slow – mostly for testing purposes.

LEAST SQUARES THEORY - CONTINUED

Error estimates (mostly from W.C. Hamilton)

Given observations $n > m$ parameters

with distributions that have finite 2nd moments

(no need to be “normal” although usually are for powders)

Then LS gives parameter estimates (shifts in our case)

with the minimum variance in any linear combination

The error estimates (“esd’s”) are

$$\sigma_i = \sqrt{\frac{b_{ii}}{\chi^2}} \quad \chi^2 = \frac{\sum w(I_o - I_c)^2}{n-m}$$

b_{ii} - diagonal elements of the inverted A matrix

Note: There is little justification for additional scaling of
the σ_i **NB: systematic errors will bias results beyond σ_i .**

$$\text{RIETVELD MODEL: } I_C = I_I \{ \sum K_p F_p^2 M_p L_p P(\Delta_p) + I_B \}$$

I_i - incident intensity - variable for fixed 2Θ (e.g. neutron TOF)

k_p - scale factor for particular phase

F_p^2 - structure factor for particular reflection

m_p - reflection multiplicity

L_p - correction factors on intensity - texture, etc.

$P(\Delta_p)$ - peak shape function - size & microstrain, etc.

Sum over all reflections under a profile point (multiple phases)

I_b – background function

More complex model than for single crystal diffraction

PROFILE FUNCTIONS $P(\Delta_p)$ – BASICS

$$\Delta_p = T_{\text{reflection}} - T_{\text{profile}} \quad (T = 2\Theta \text{ or TOF})$$

Gaussian profile - generally instrumental origin

$$G(\Delta T, \Gamma) = \sqrt{\frac{4 \ln 2}{\pi \Gamma^2}} \exp\left[\frac{-4 \ln 2 (\Delta T)^2}{\Gamma^2}\right]$$

Lorentzian profile - largely sample effect

$$L(\Delta T, \gamma) = \frac{2}{\pi \gamma} \frac{1}{1 + \left(\frac{2\Delta T}{\gamma}\right)^2}$$

Voigt – convolution = $G \otimes L$

Pseudo-Voigt – linear combination = $\eta L + (1-\eta)G$

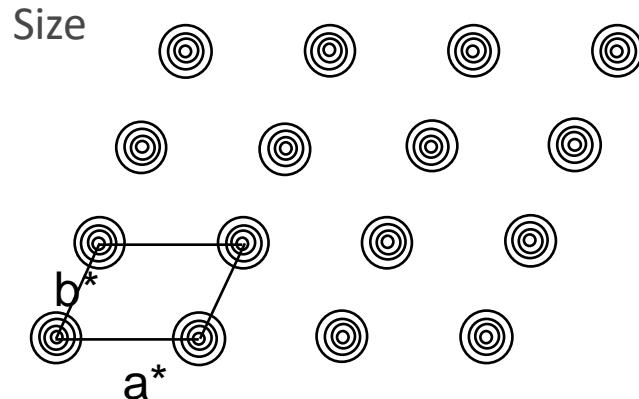
η via Thompson, Cox & Hastings – pseudoVoigt = Voigt

CW Asymmetry from axial divergence – Finger, Cox & Jephcoat

NB: in gsas & GSAS-II, T is 2Θ in centideg or TOF in μs

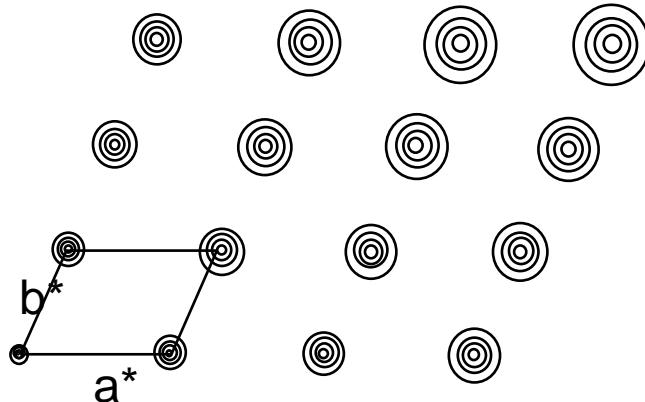
SAMPLE BROADENING

Isotropic Crystallite size & μ strain broadening



Small ($<1\mu\text{m}$) crystals \rightarrow not δ -functions
Size distribution \rightarrow
superposition of sharp to broad spots
 \rightarrow Shape \sim Lorentzian
Width $\Delta d^* = \text{constant} = \Delta d/d^2 = \Delta\Theta \cot\Theta/d$
Bragg's Law: $\Delta 2\Theta = \lambda \Delta d/d^2 \cos\Theta$ ($= X/\cos\Theta$)
 \rightarrow Scherrer equation
 $k=1, p = \text{size}$
$$S = \frac{180k\lambda}{\pi p \cos \theta}$$

μ strain



Unit cell variation (defects??)
Lorentzian distribution \rightarrow shape
 $\Delta d/d = \text{constant} = \Delta d^*/d^* = \Delta\Theta \cot\Theta$
Or: $\Delta 2\Theta = 2\Delta d \tan\Theta/d$ ($= Y \tan\Theta$)
$$M = 180\mu \tan \Theta / \pi$$

 $\mu - \mu$ strain ($\times 10^6$) parameter

CW PROFILE COEFFICIENTS

Lorentzian vs Gaussian sample broadening?

- Size:

$$S = \frac{180k\lambda}{\pi p \cos \theta} \quad \mu\text{strain:} \quad M = 180\mu \tan \Theta / \pi$$

- Need: S_Γ (Gauss) & S_γ (Lorentzian) sample broadening (2 slides back)

$$\Gamma_g^2 = 8\ln 2(U\tan^2\Theta + V\tan\Theta + W + S_\Gamma)$$

$$\gamma = \frac{X}{\cos\Theta} + Y\tan\Theta + Z + S_\gamma$$

- Mixing coeff for each; m_s & m_μ (NB: called 'mx' in GSAS-II; range 0-1)

$$S_\gamma = m_s S + m_\mu M$$

$$S_\Gamma = [(1 - m_s)^2 S^2 + (1 - m_\mu)^2 M^2] / 8\ln 2$$

- Normally m_s & $m_\mu = 1$ (all Lorentzian sample broadening) so:

$$S_\gamma = S + M$$

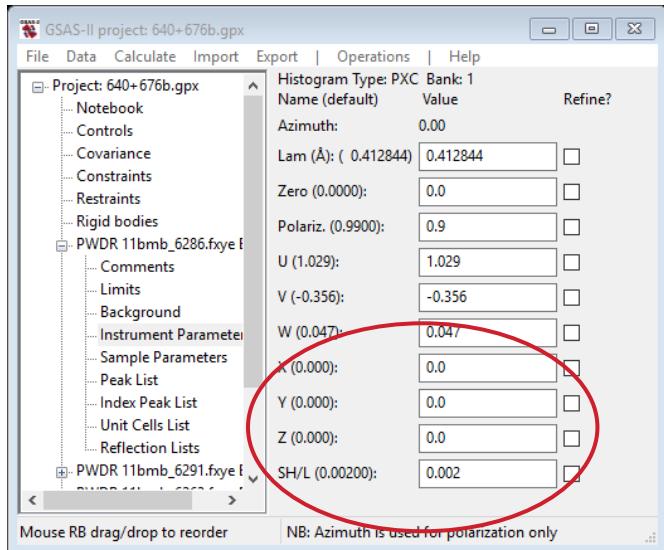
$$S_\Gamma = 0 \quad (\text{no Gaussian sample broadening})$$

- $X, Y, Z = 0$ (no Lorentzian instrument broadening)

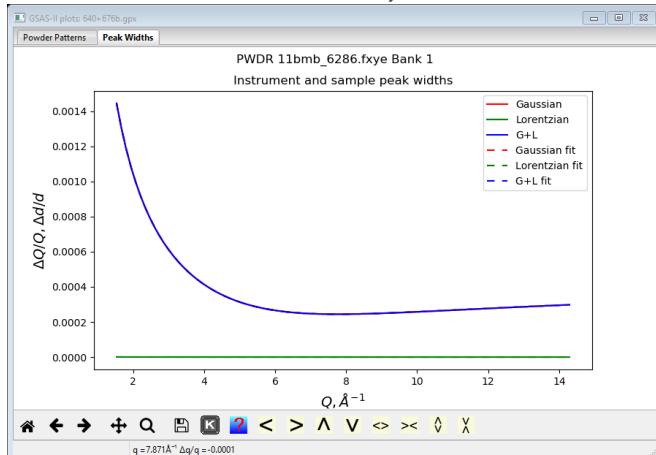
CW PROFILE PEAK BROADENING IN GSAS-II

The split of sample broadening from instrumental contribution

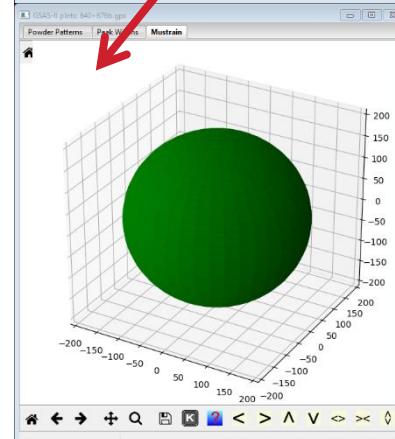
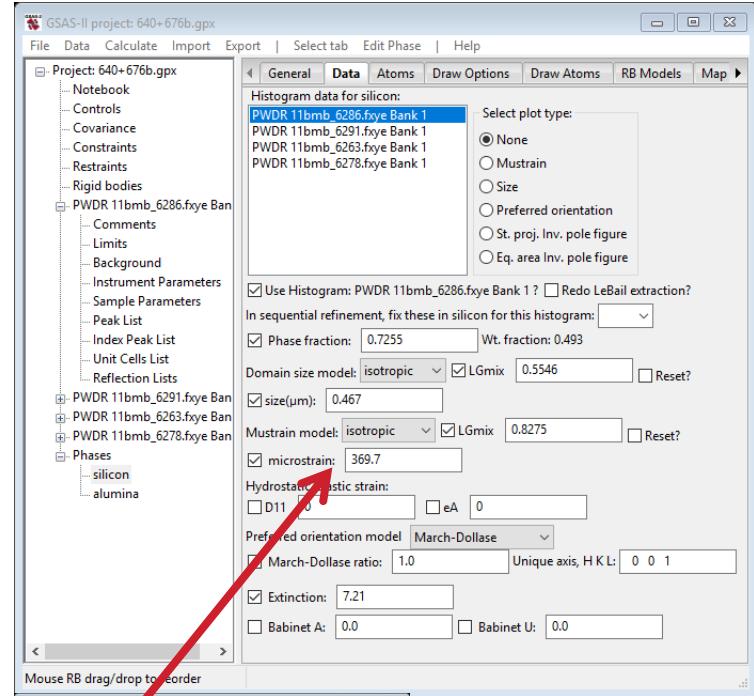
Instrument – fixed from calibration



NB: for APS 11BM X,Y & Z = 0



Sample – phase & histogram dependent
Refined & constrained as needed



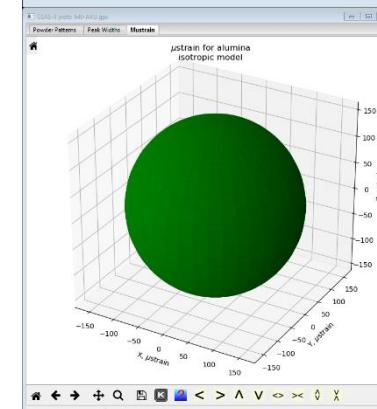
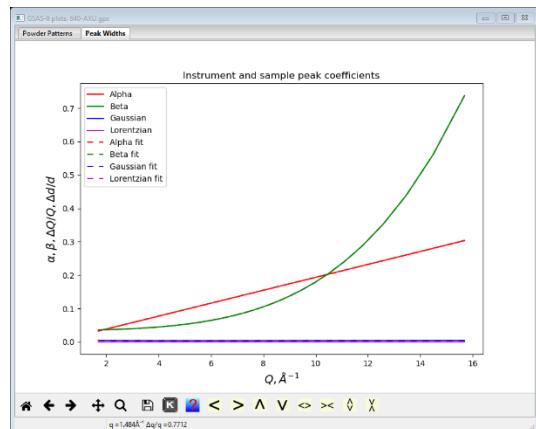
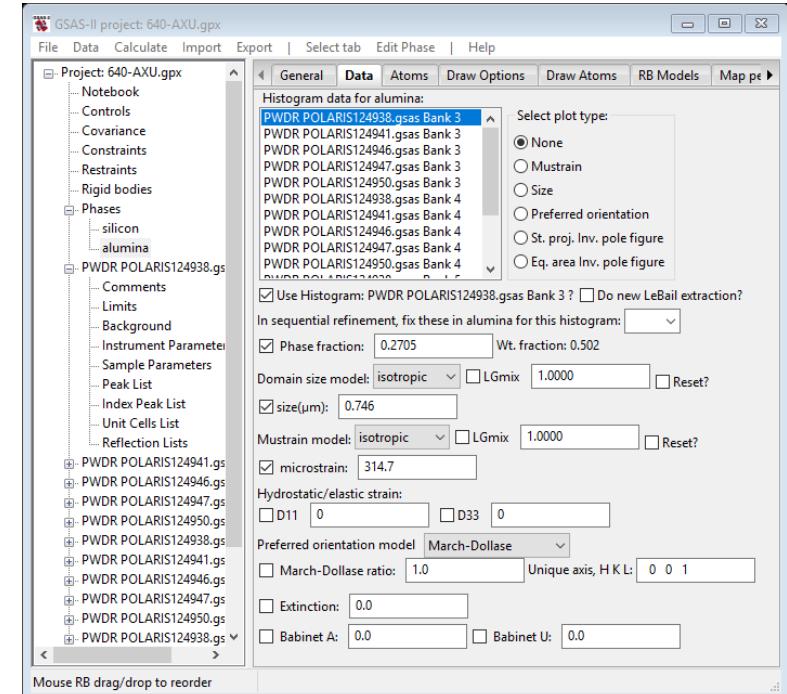
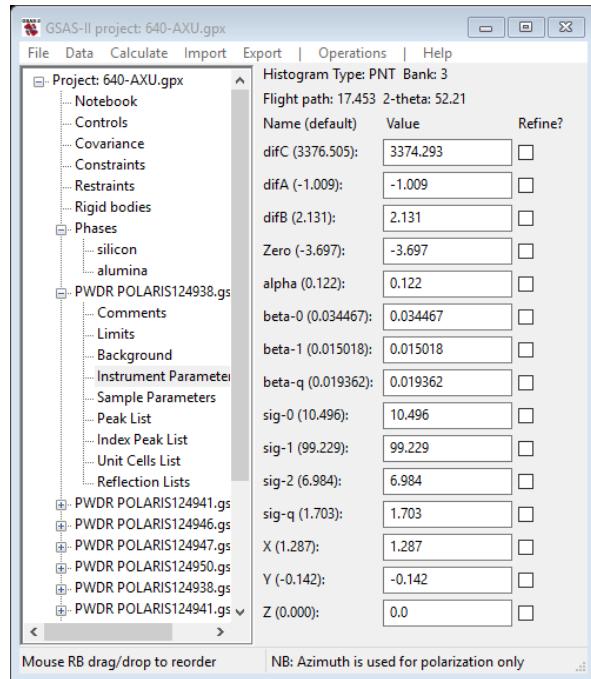
Sample:
New NIST SRMS
640f & 676b

TOF PROFILE PEAK BROADENING IN GSAS-II

The split of sample broadening from instrumental contribution

Instrument – fixed from calibration

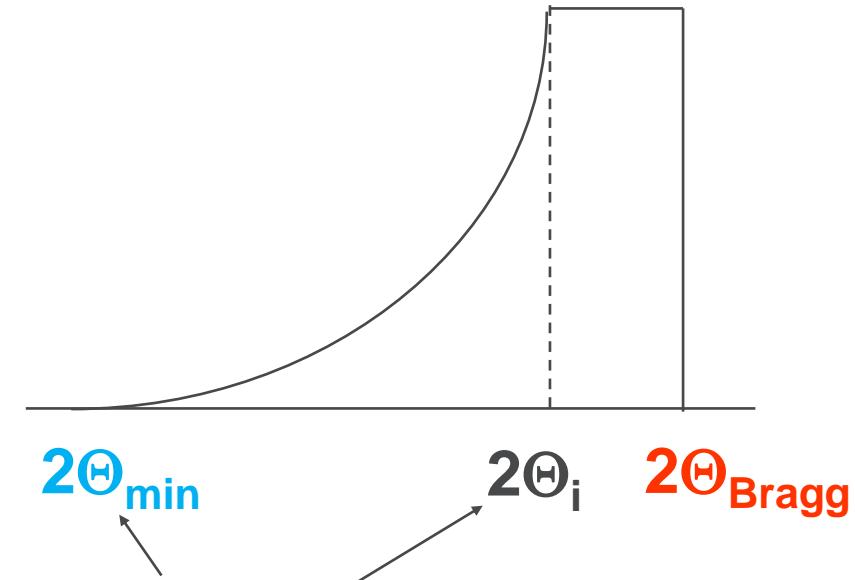
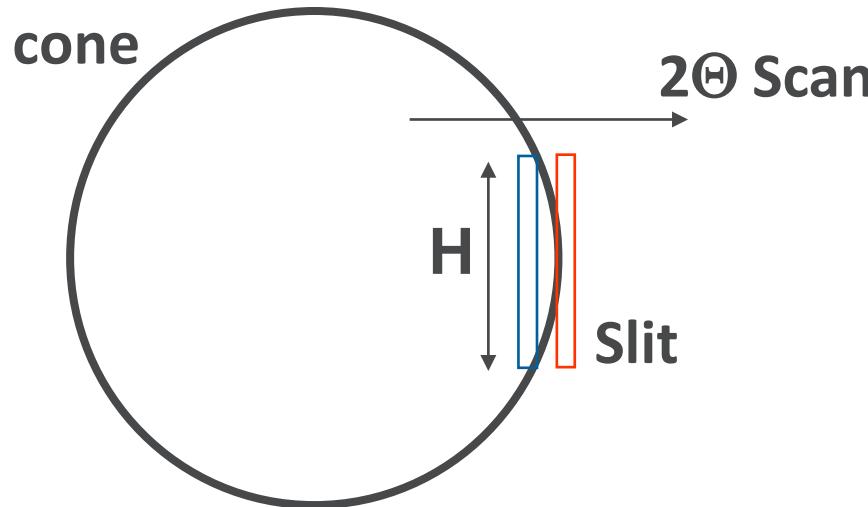
Sample – phase & histogram dependent
Independent of experiment (e.g. CW or TOF)



Sample:
New NIST SRMS
640f & 676b

AXIAL BROADENING FUNCTION – CONST. WAVELENGTH

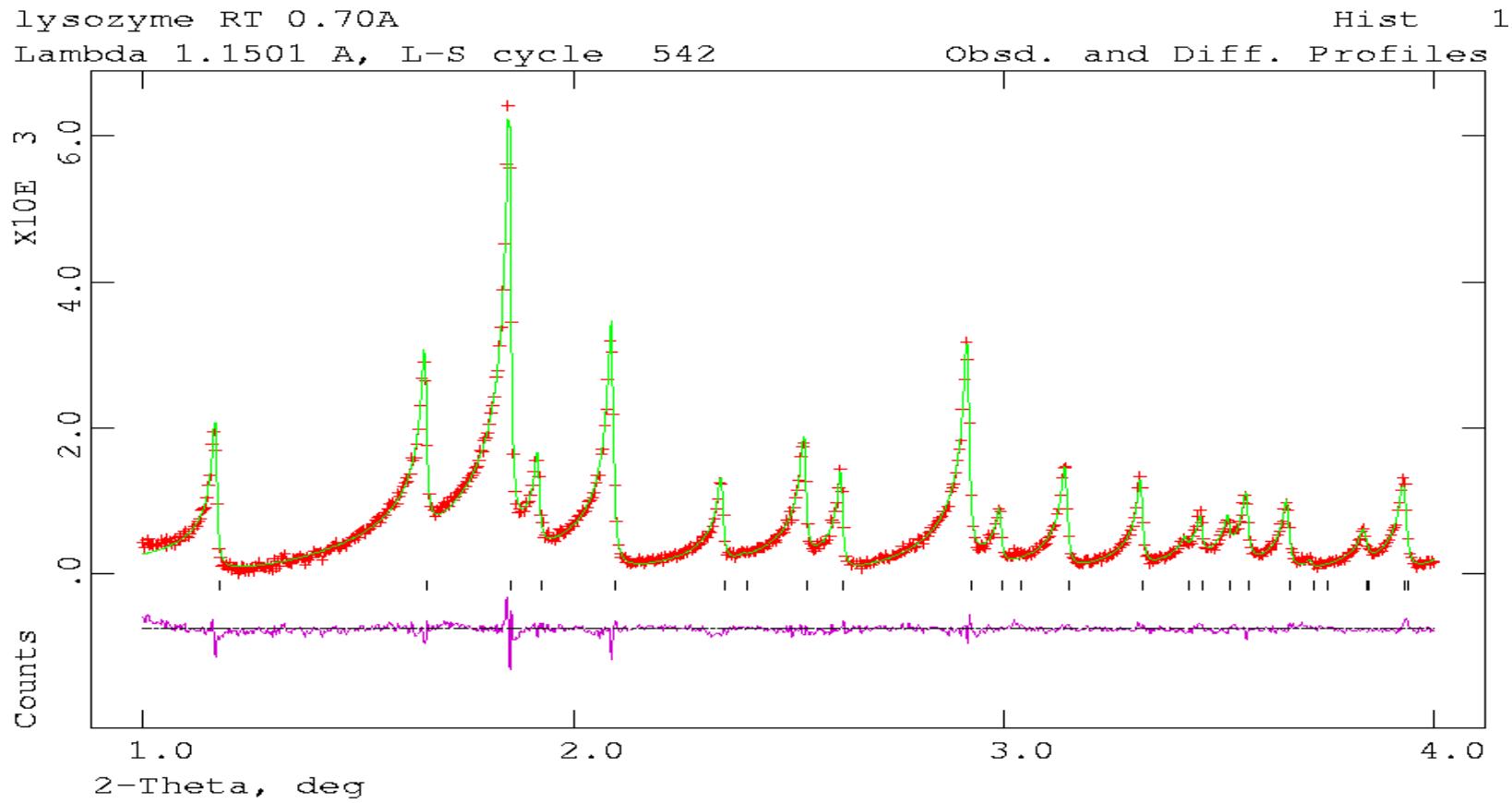
Finger, Cox & Jephcoat based on van Laar & Yelon
Debye-Scherrer



Depend on slit & sample “heights” wrt diffr. radius
H/L & S/L - parameters in function; combined as H+S/L in GSAS-II
(typically 0.005 - 0.020)

⊗ Pseudo-Voigt
= profile function

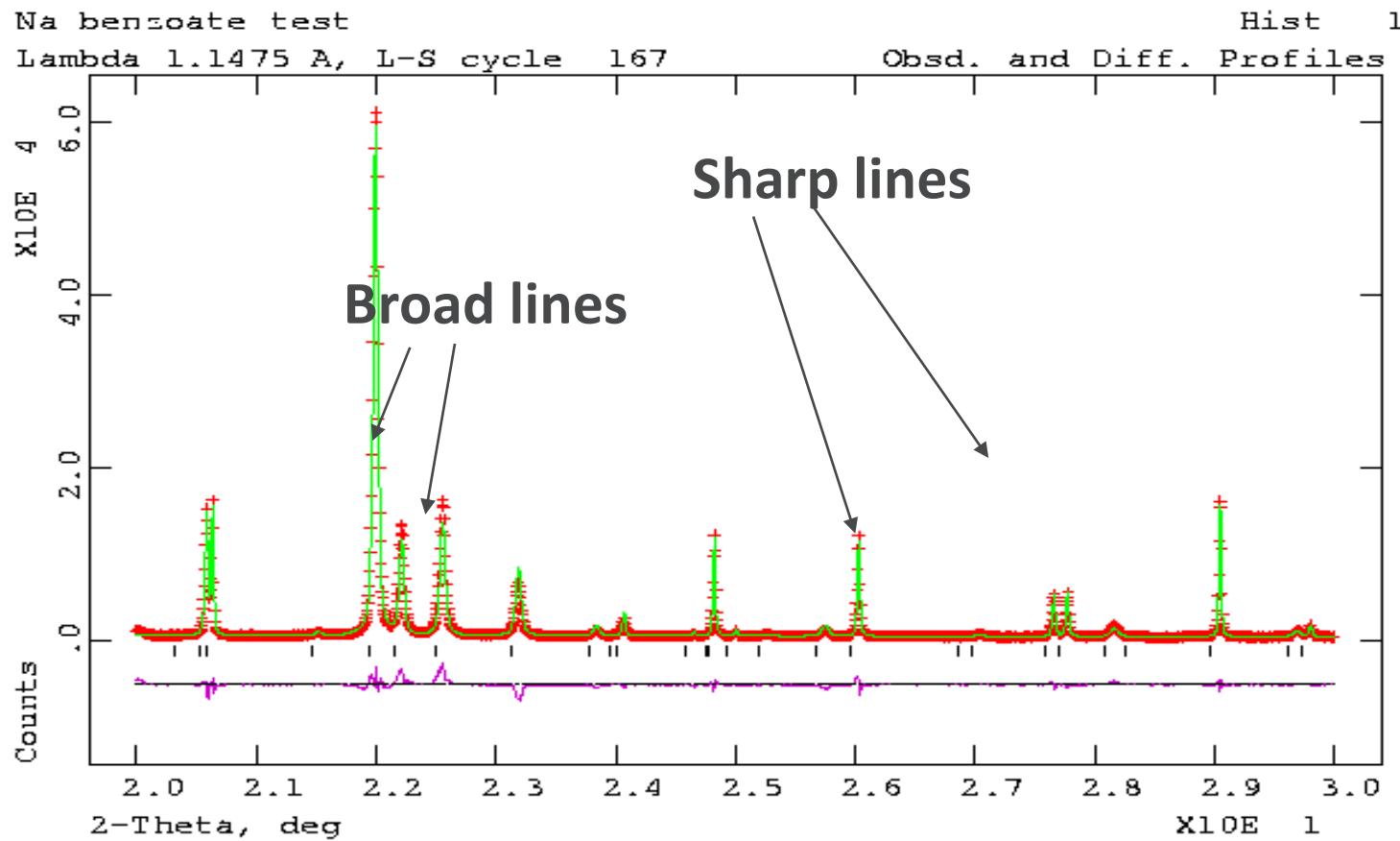
HOW GOOD IS THIS FUNCTION?



Protein Rietveld refinement - Very low angle fit
1.0-4.0° peaks - strong asymmetry
“perfect” fit to shape

PROFILE FUNCTION – COMPLEXITIES

AN EXAMPLE – UNUSUAL LINE BROADENING



Seeming inconsistency in line broadening - hkl dependent

MICROSTRAIN BROADENING – PHYSICAL MODEL

Model – elastic deformation of crystallites

Stephens, P.W. (1999). *J. Appl. Cryst.* 32, 281-289.

Also see Popa, N. (1998). *J. Appl. Cryst.* 31, 176-180.

d-spacing expression

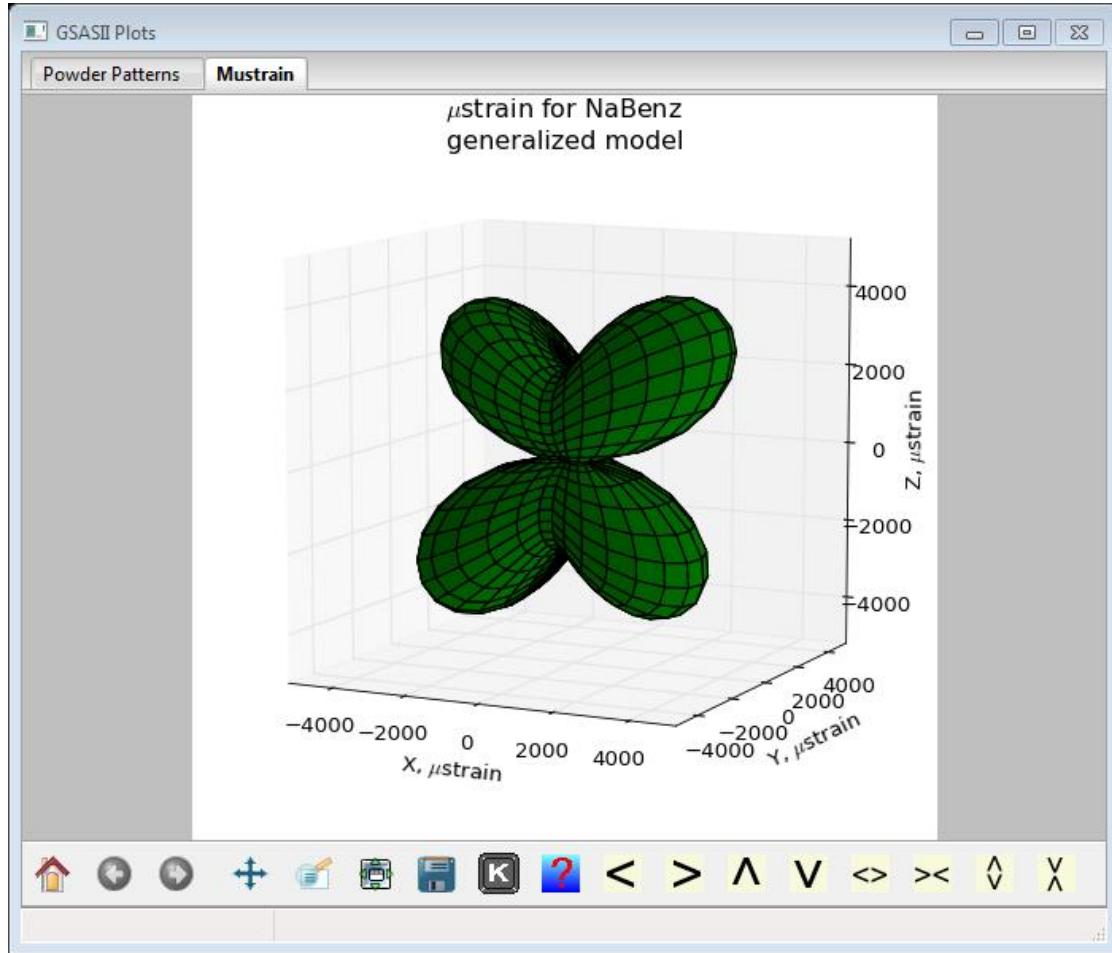
$$\frac{1}{d_{hkl}^2} = M_{hkl} = \alpha_1 h^2 + \alpha_2 k^2 + \alpha_3 l^2 + \alpha_4 kl + \alpha_5 hl + \alpha_6 hk$$

Broadening – variance in M_{hkl} ; refine C_{ij}

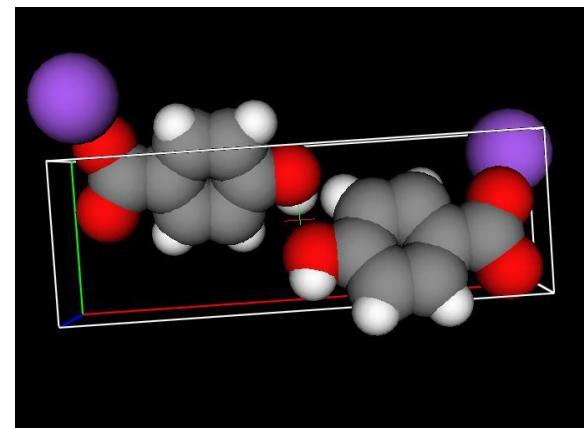
$$\sigma^2(M_{hkl}) = \sum_{i,j} C_{ij} \frac{\partial M}{\partial \alpha_i} \frac{\partial M}{\partial \alpha_j}$$

NA PARAHYDROXYBENZOATE

Unusual microstrain effects - peak broadening

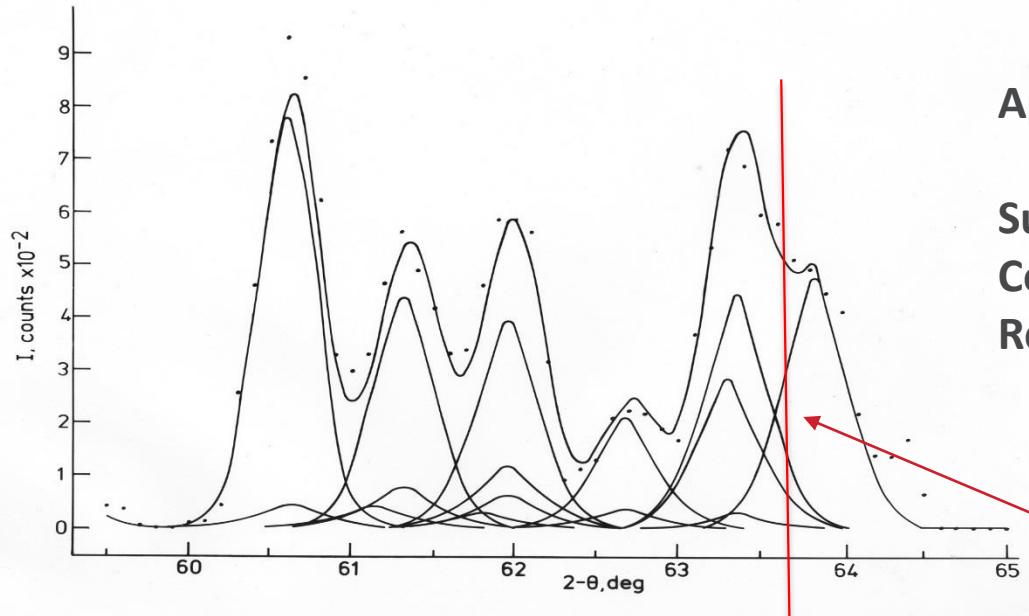


Directional dependence -
Lattice defects?
Inclusion allowed
OH atom
placement from ΔF
map



INTENSITY EXTRACTION

Structure factors from powder patterns? → structure solution



Apportion I_o by ratios of $I_c(H)$
for contributing reflections →
Sum over all under peak profile
Correct for multiplicity & L_p , etc.
Result is $F^2(H)$

Here 4 reflections contribute

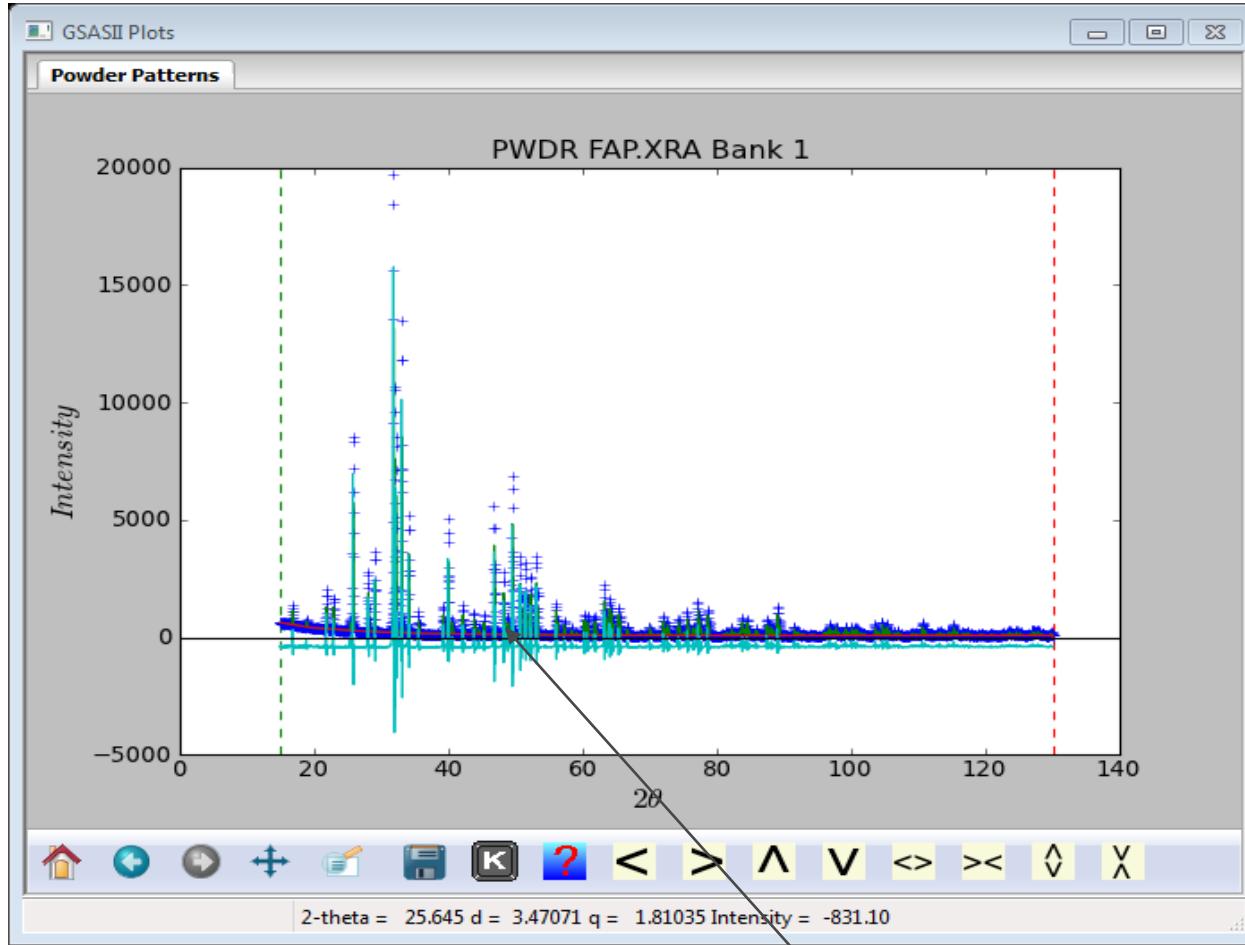
LeBail algorithm – extracted $F^2_o \rightarrow$ new F^2_c then next cycle;
refine only background, peak shapes & positions – few parameters
No constraints needed for overlaps – Simple

Pawley refinement – F^2_o are parameters
+ background, peak shapes & positions – many parameters
Constraints & restraints required for overlaps - Complex

RIETVELD REFINEMENT – A SIMPLE EXAMPLE

AN EXAMPLE: FLUROAPATITE

Add atoms & do default initial refinement
– scale & background



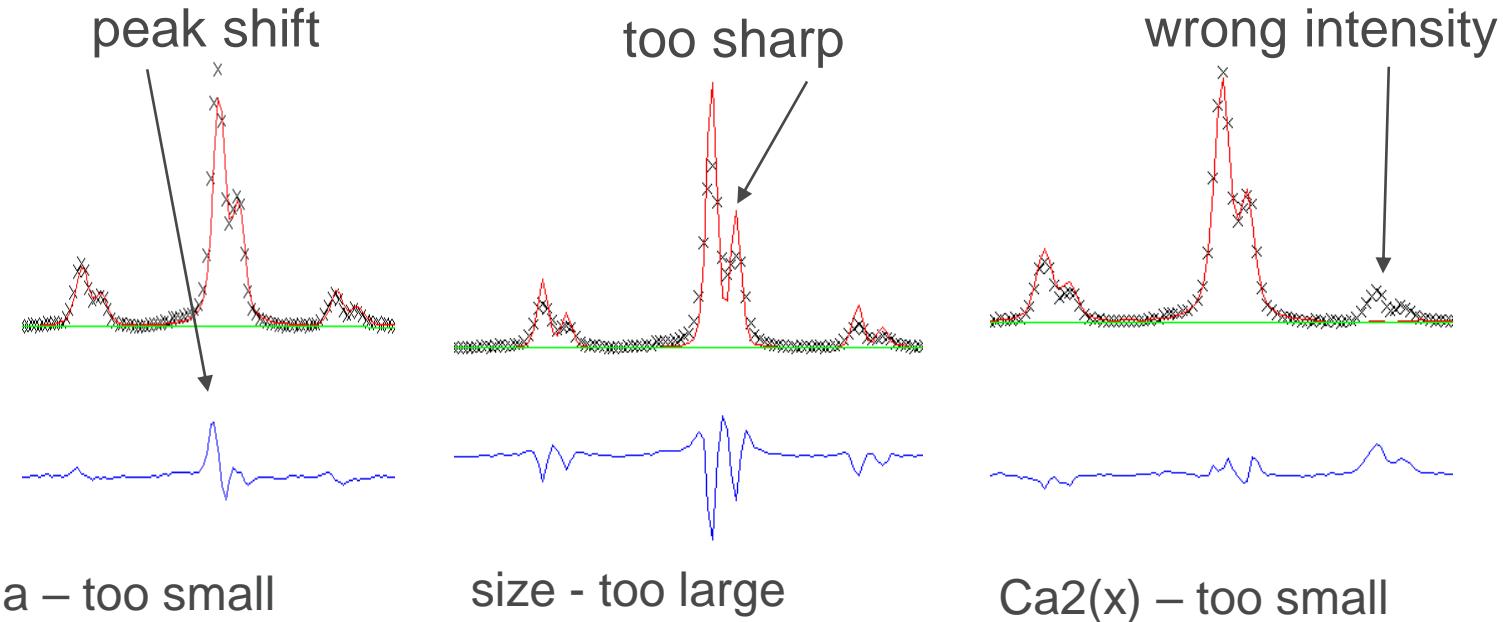
- Notice shape of difference curve – position/shape/intensity errors

ERRORS & PARAMETERS?

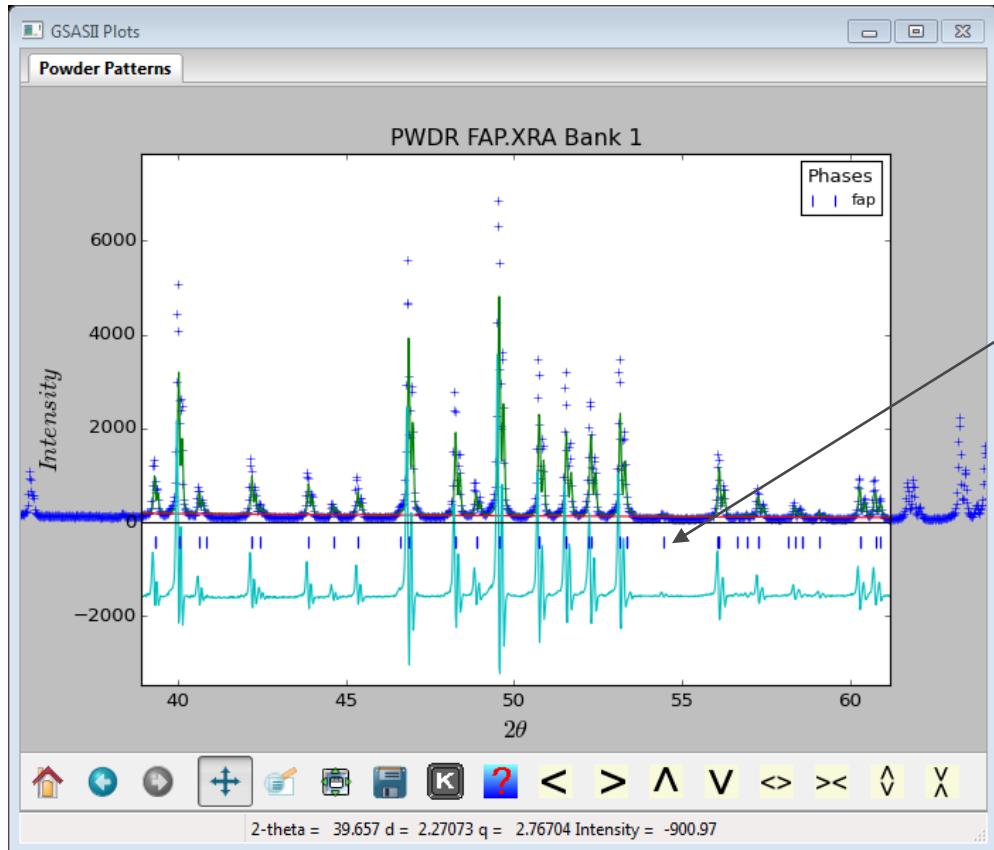
- position – lattice parameters, zero point (not common)
 - other systematic effects – sample shift/offset
- shape – profile coefficients – sample size/ μ strain
 - (U, V, W, X, Y, etc. in GSAS-II are instrument parms.)
- intensity – crystal structure (atom positions & thermal parameters)
 - other systematic effects (absorption/extinction/preferred orientation)

NB – get linear combination of all the above

NB² – trend with 2Θ (or TOF) important



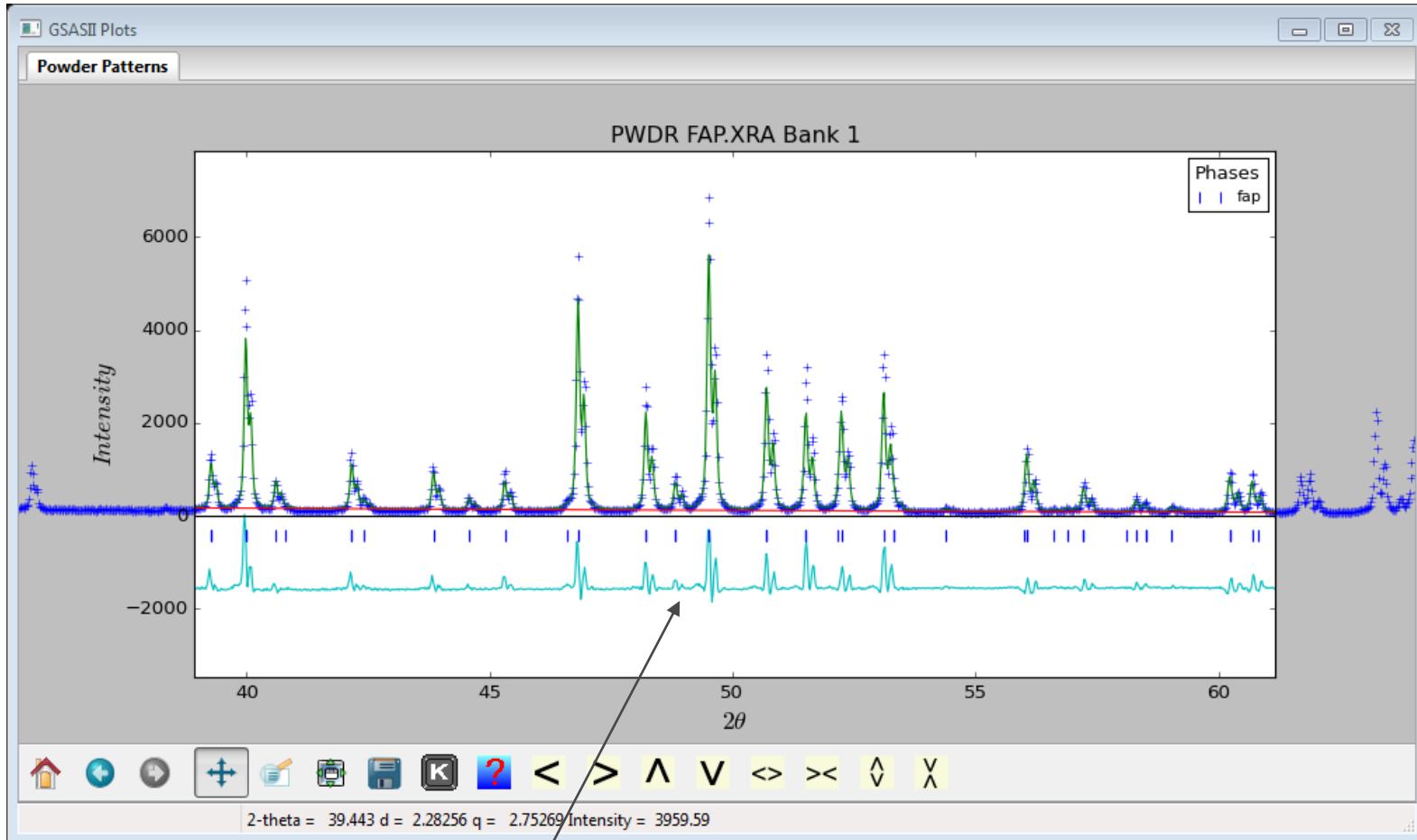
DIFFERENCE CURVE – WHAT TO DO NEXT?



Characteristic “up-down-up”
→profile error
NB – can be “down-up-
down” for too “fat” profile

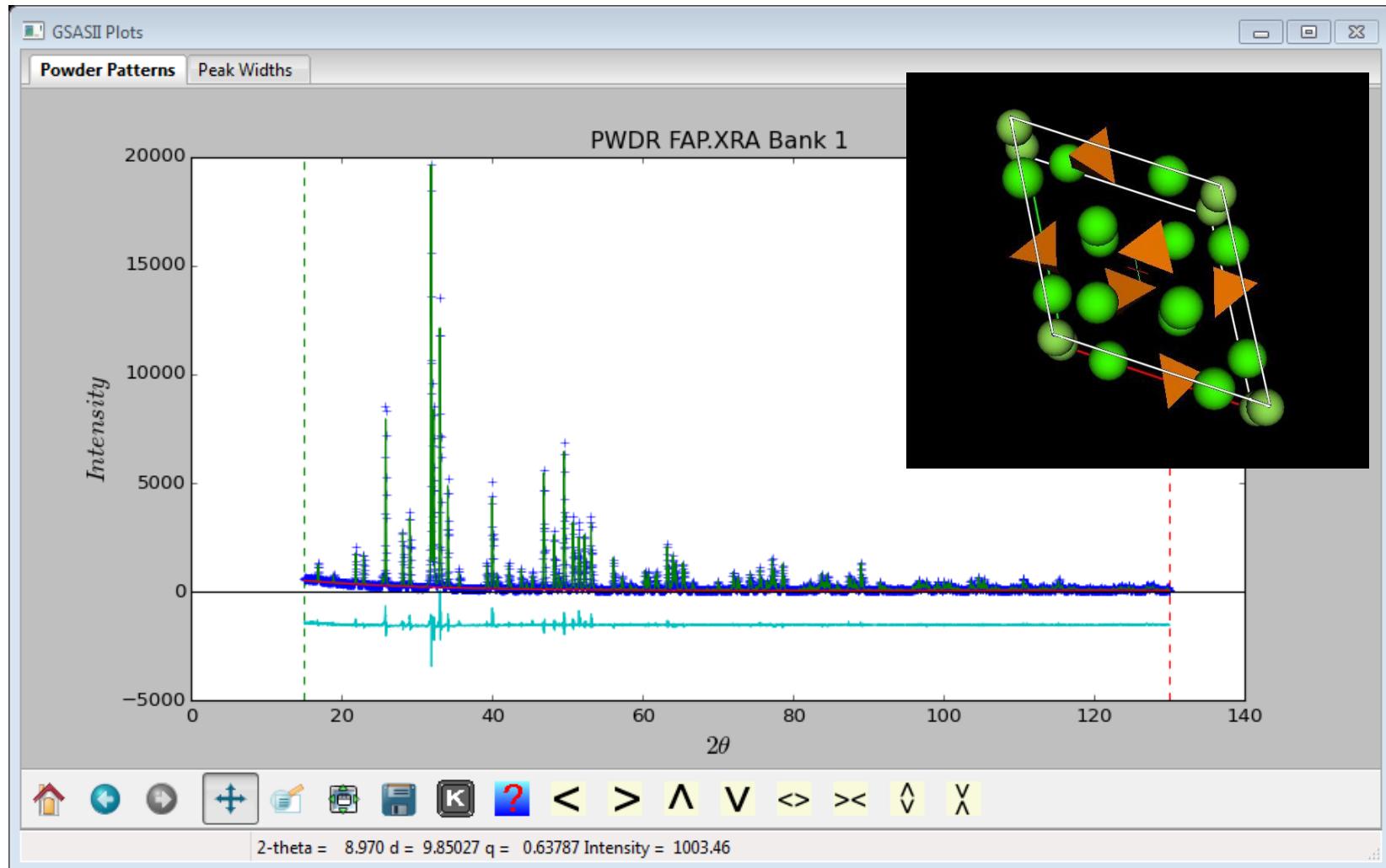
- Dominant error – peak positions? peak shapes - too sharp?
- Refine sample μ strain parameter next & include lattice parameters
- **NB - EACH CASE IS DIFFERENT – no magic recipe**

RESULT – MUCH IMPROVED!



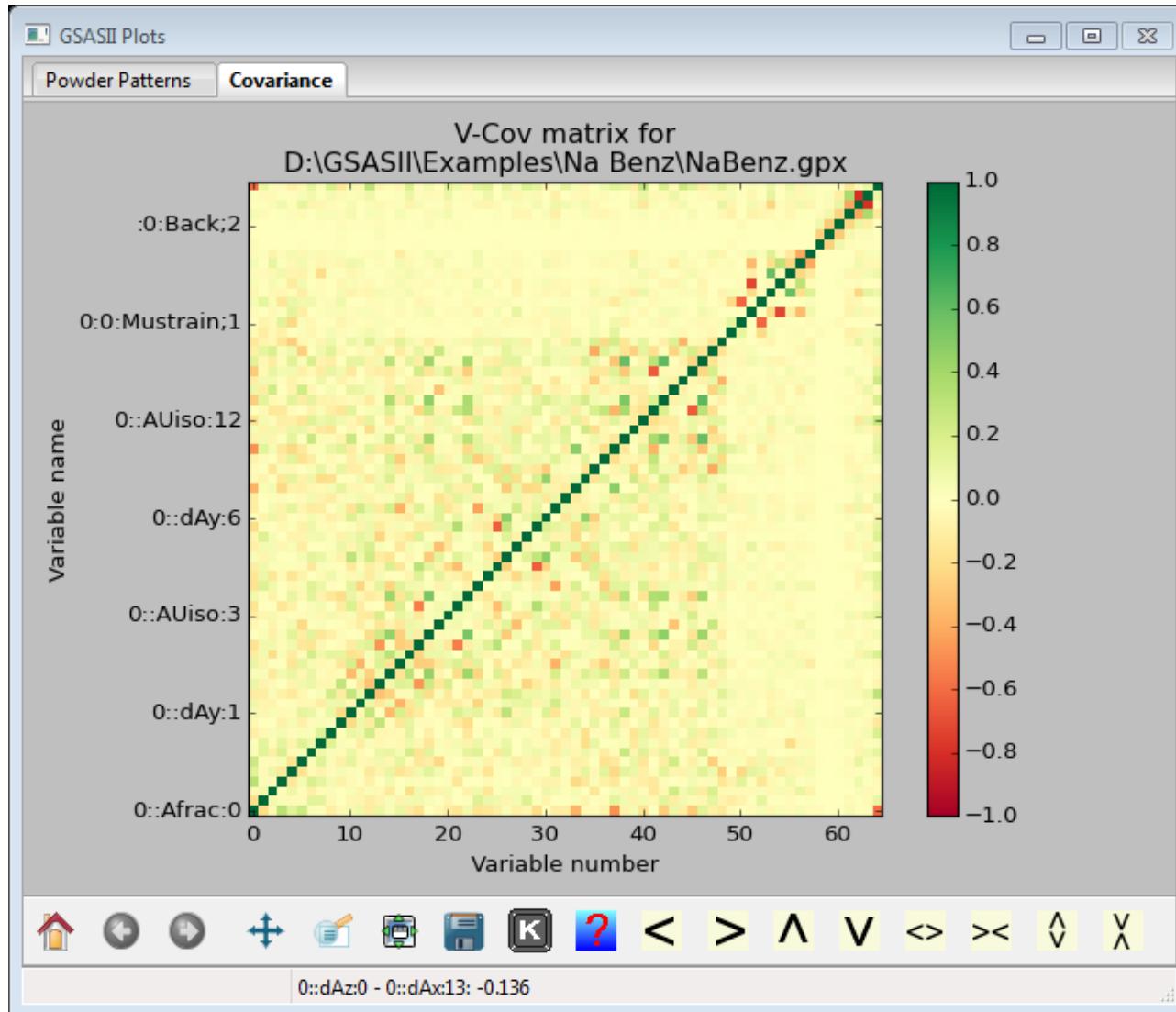
- maybe intensity differences remain
 - refine coordinates & thermal parms.

RESULT – ESSENTIALLY UNCHANGED



- Thus, major error in the initial model – peak shapes & sample displacement/lattice parameters

A USEFUL PLOT – COVARIANCE MATRIX



Green: $\text{cov} > 0$
Red: $\text{cov} < 0$
Yellow: $\text{cov} \sim 0$
Cursor reports:
Cov or value(esd)
on diagonal
Can be zoomed!

Beware white
bands & nan:
Singularities!

A FEW FINAL WORDS

“A Rietveld refinement is never perfected, merely abandoned” (P. Stephens, 2000)

“Rietveld refinement is one of those few fields of intellectual endeavor wherein the more one does it, the less one understands.” (Sue Kesson)

“A Rietveld refinement is done when you run out of parameters” (R. Von Dreele)

Books:

Modern Powder Diffraction, Eds. J. Post & D. Bish (1989)

The Rietveld Method, Ed. R.A. Young (1993)

Powder Diffraction: Theory & Practice, Eds. R. Dinnebier & S. Billinge (2008)

The background of the slide is a grayscale aerial photograph of the Argonne National Laboratory complex. The image shows a dense network of buildings, roads, and green spaces, with a large circular facility, likely the Experimental Breeder Reactor II (EBR-II), visible in the center-left. The surrounding area is a mix of developed land and natural terrain.

THANK YOU