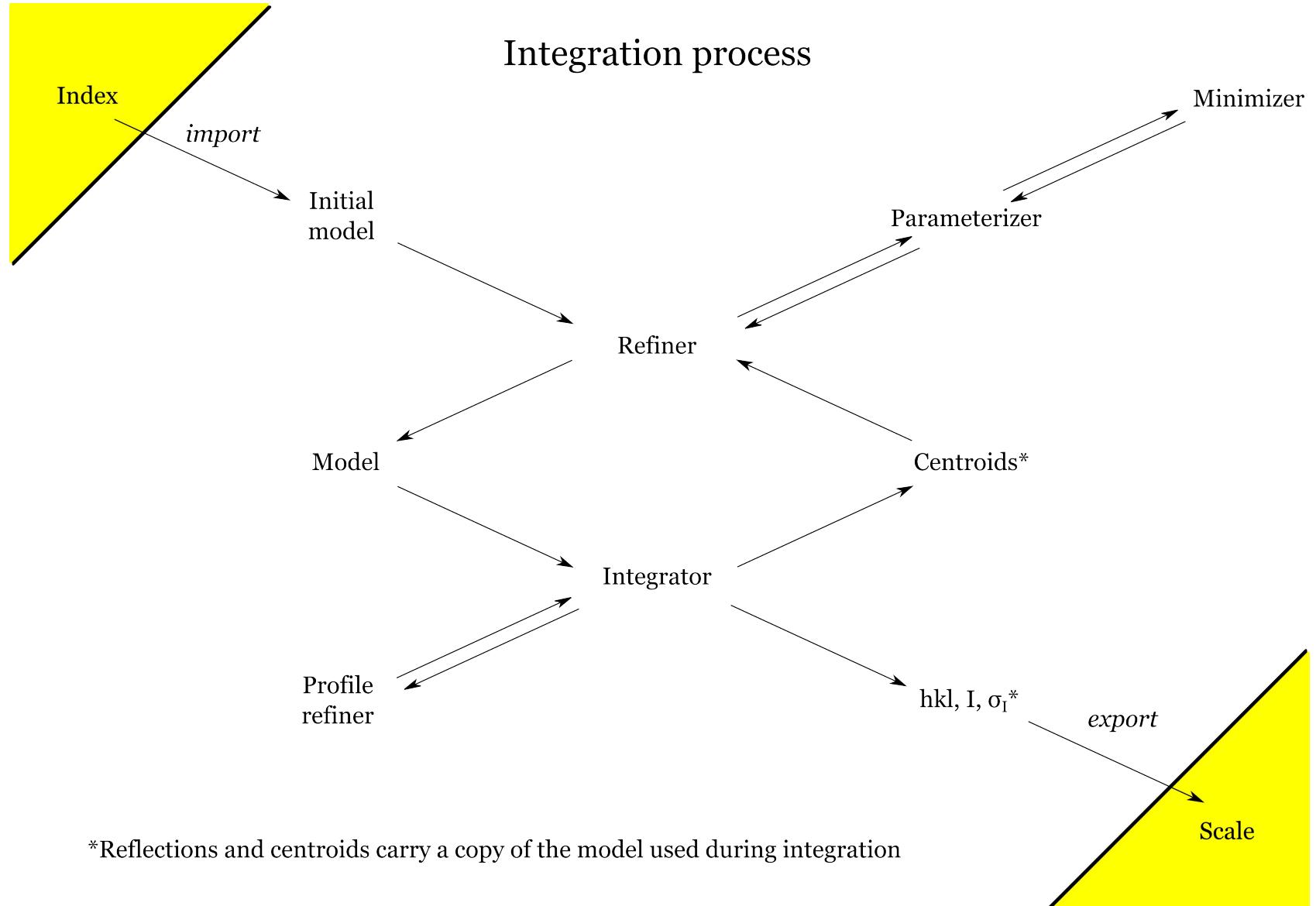


Diffraction geometry parameterisation and refinement

David Waterman

CCP4

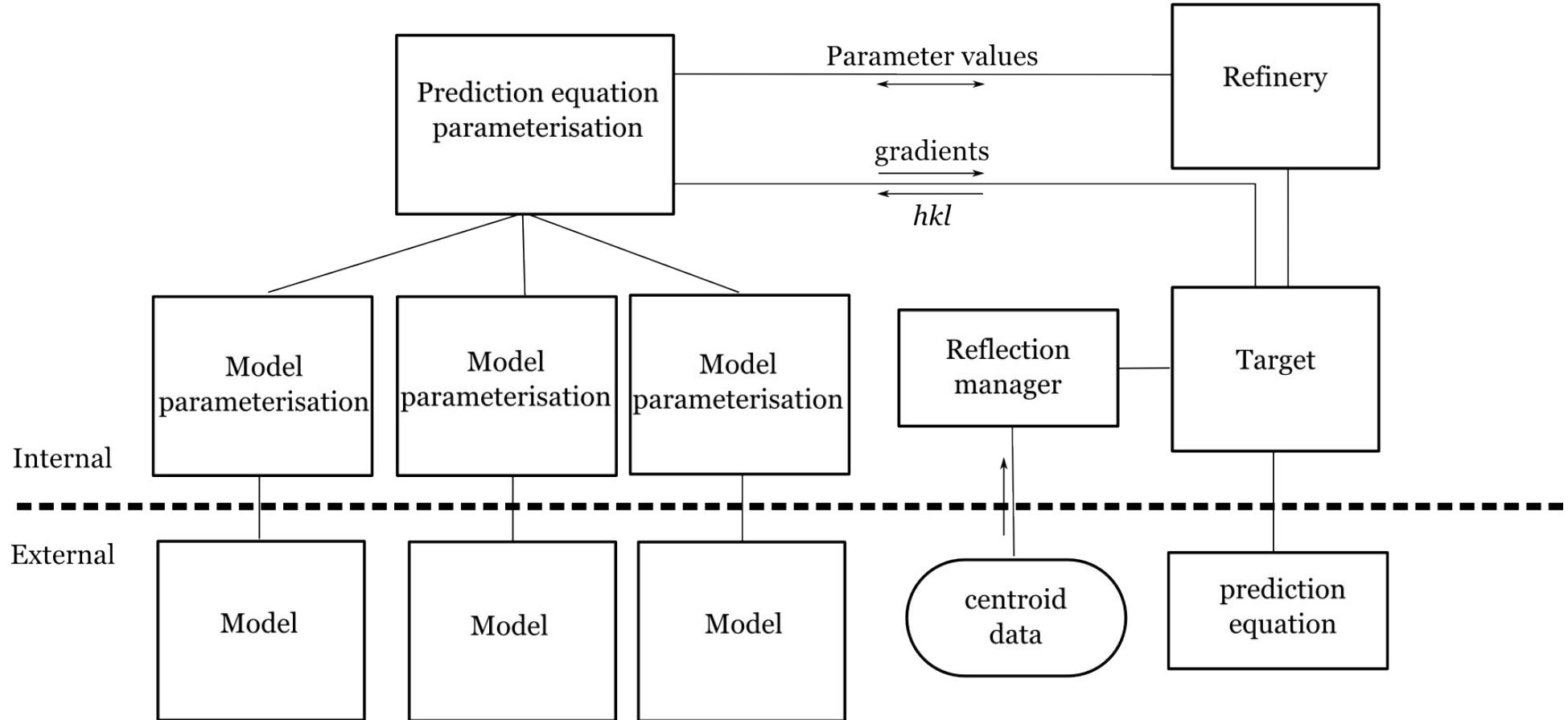
Cambridge 22 May 2013



dials Motivation for global refinement

- Input diffraction spot indices, their centroids and estimated uncertainties ($h, k, l; X, Y, \phi; \sigma_x, \sigma_y, \sigma_\phi$)
- Use all (useful) data available to refine a model to reduce rmsd of predicted centroids
- Global refinement helps to recover from poorly defined parameters in local ϕ window
- More physically meaningful: avoids mopping up of effects by correlated parameters and therefore obtains realistic parameter values
- Refine profile parameters separately
- Potential second round with improved centroid observations

Overview



Progress to date

- April 2012
 - Start work on detector model and parameterisation
- After Algorithm Camp II (May 2012)
 - Broad overview of module design formed with Target-Minimiser-Basis organisation and collection of parameter values and derivatives from components of a global model
- June 2012
 - Abstract model parameterisation class and a detector parameterisation instance
- August 2012
 - Tested detector parameterisation derivatives by finite differences
 - First refinery class
 - First refinement of detector parameters by L-BFGS on simulated data

- Sept 2012
 - Parameterisation of source orientation
- Oct 2012
 - Begin prediction equation parameterisation class (for derivatives of the basis function)
 - crystal model
 - goniometer model
- Nov 2012: LMB meeting
 - First orientation refinement results presented. Couldn't refine detector and source simultaneously!
- Dec 2012
 - Fixed derivatives of X, Y wrt parameters that also affect phi
 - All derivatives tested by FD
 - Add curvatures (LSQ approximation) to L-BFGS minimiser
 - Only expose free parameters to the minimiser

Progress to date

- Jan 2013
 - Rescaled angle parameters to mrad. Much improved results with L-BFGS
- Feb 2013: LBNL workshop
 - 3* speed-up on refactoring, JMP's help with moving code to C++, and avoiding an unnecessary copy
 - Don't normalise target by number of preds each cycle (req changes to prediction to return a result even when out of Panel bounds)
 - Interface to NKS crystal unit cell parameterisation. Test all gradients.
 - Supply inverse of curvatures (i.e. diagonal of Hessian) to minimiser(!)
 - Move refinement code to new DIALS project on sourceforge

Progress to date

- Mar 2013
 - Add LSTBX engine for non-linear least squares refinement by Gauss-Newton iterations (much better)
 - Convert to using new DIALS models (dxtbx) and DIALS reflection prediction throughout
- April 2013
 - DIALS centroid refinement sprint. First use against real data
- May 2013
 - Time dependent parameterisation of the crystal (work in progress)

Target function

- Simple least squares target. No restraints terms added (yet)

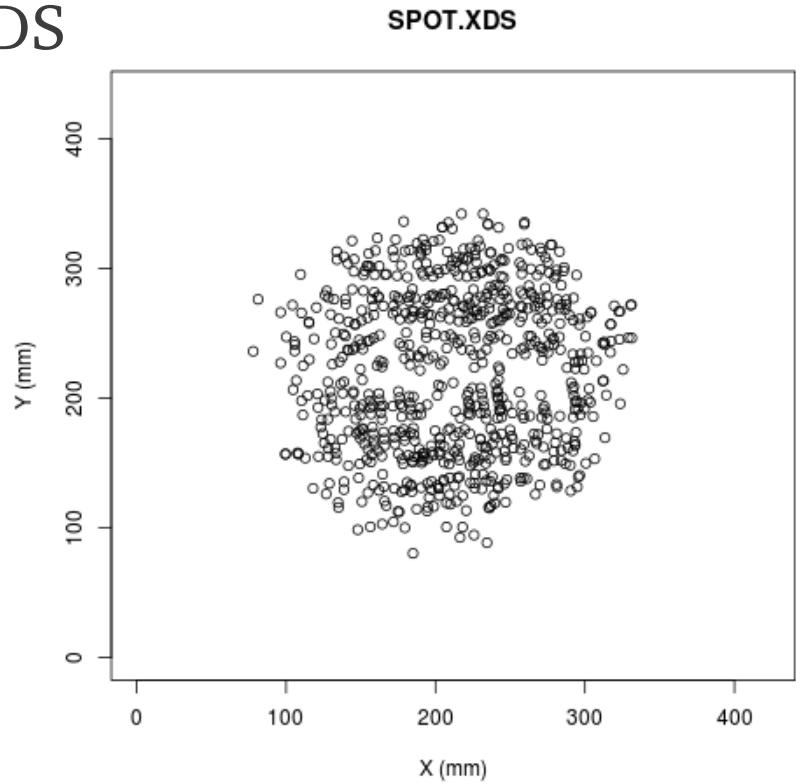
$$L = \frac{1}{2} \sum_h w_{x,h} (X_c - X_o)^2 + w_{y,h} (Y_c - Y_o)^2 + w_{\phi,h} (\phi_c - \phi_o)^2$$

$$\frac{dL}{dp} = \sum_h w_{x,h} (X_c - X_o) \frac{dX_c}{dp} + w_{y,h} (Y_c - Y_o) \frac{dY_c}{dp} + w_{\phi,h} (\phi_c - \phi_o) \frac{d\phi_c}{dp}$$

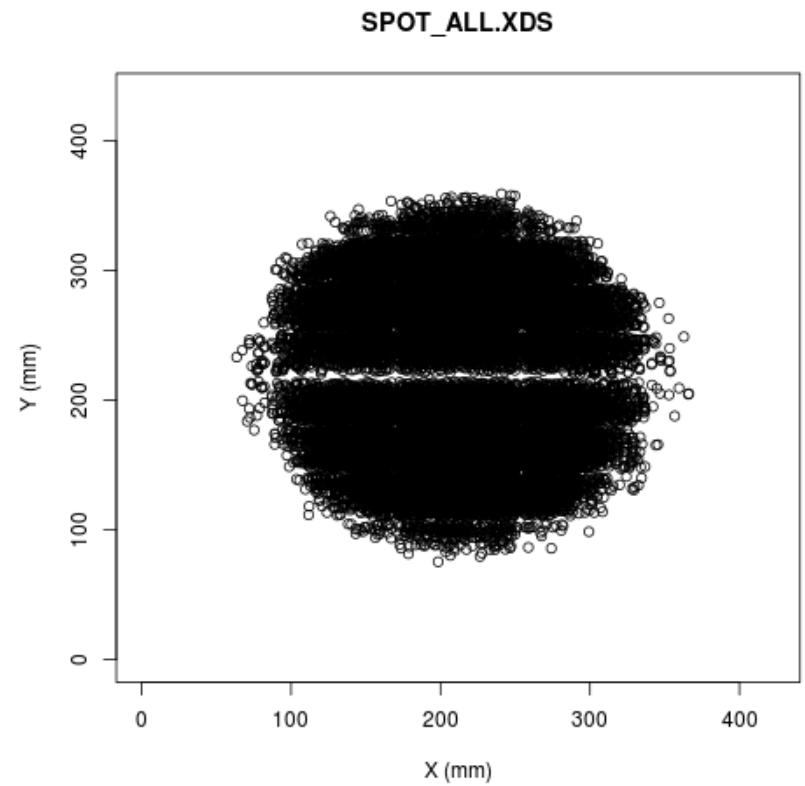
- First order approximation to the curvatures

$$\frac{d^2 L}{dp^2} \approx \sum_h w_{x,h} \left(\frac{dX_c}{dp} \right)^2 + w_{y,h} \left(\frac{dY_c}{dp} \right)^2 + w_{\phi,h} \left(\frac{d\phi_c}{dp} \right)^2$$

- SPOT.XDS: 742 strong reflections distributed over three 0.8° wedges around 0° , 45° and 90°
- Refinement starting with GXPARM.XDS
 - gradient converged in 6 steps
 - RMSD_X slightly worse, RMSD_Y and RMSD_phi slightly better
- Refinement starting with XPARM_REGULARIZED.XDS
 - gradient converged in 8 steps
 - slightly worse rmsd than above



- SPOT_ALL.XDS: 29023 reflections distributed over full 180° sweep
- Refinement starting with GXPARM.XDS
 - gradient converged in 3 steps
 - very slight improvement in rmsd
- Refinement starting with XPARM_REGULARIZED.XDS
 - gradient converged in 3 steps
 - obtains same end result



Centroid refinement results

Reading: "GXPARM.XDS"
 Reading: "INTEGRATE.HKL"

Experimental Models

Beam:

wavelength: 0.9795
 direction : {0.00785262,-2.51934e-14,-0.999969}

Detector:

Panel:

type:	SENSOR_UNKNOWN
fast axis:	{0.999956,0.00197257,0.00915725}
slow axis:	{0.001983,-0.999997,-0.00113048}
origin:	{-211.431,219.408,-192.801}
normal:	{0.009155,0.00114859,-0.999957}
pixel size:	{0.172,0.172}
image size:	{2463,2527}
trusted range:	{0,0}

Goniometer:

Rotation axis: {1,-1.80647e-15,-8.38392e-15}
 Fixed rotation: {1,0,0,0,1,0,0,0,1}

Scan:

image range: {1,900}
 oscillation: {0,0.2}
 exposure time: 0

Crystal:

Unit cell: (42.275, 42.275, 39.669, 90.000, 90.000, 90.000)
 U matrix: {{ 0.8336, 0.5360, 0.1335},
 {-0.1798, 0.0348, 0.9831},
 { 0.5223, -0.8435, 0.1254}}
 B matrix: {{ 0.0237, 0.0000, 0.0000},
 { 0.0000, 0.0237, 0.0000},
 { 0.0000, 0.0000, 0.0252}}
 A = UB: {{ 0.0197, 0.0127, 0.0034},
 {-0.0043, 0.0008, 0.0248},
 { 0.0124, -0.0200, 0.0032}}

Reading: "XPARM_REGULARIZED.XDS"
 Reading: "INTEGRATE.HKL"

Experimental Models

Beam:

wavelength: 0.9795
 direction : {-0,1.22465e-16,-1}

Detector:

Panel:

type:	SENSOR_UNKNOWN
fast axis:	{1,0,0}
slow axis:	{0,-1,-1.22465e-16}
origin:	{-212.754,219.609,-191.109}
normal:	{0,1.22465e-16,-1}
pixel size:	{0.172,0.172}
image size:	{2463,2527}
trusted range:	{0,0}

Goniometer:

Rotation axis: {1,0,0}
 Fixed rotation: {1,0,0,0,1,0,0,0,1}

Scan:

image range: {1,900}
 oscillation: {0,0.2}
 exposure time: 0

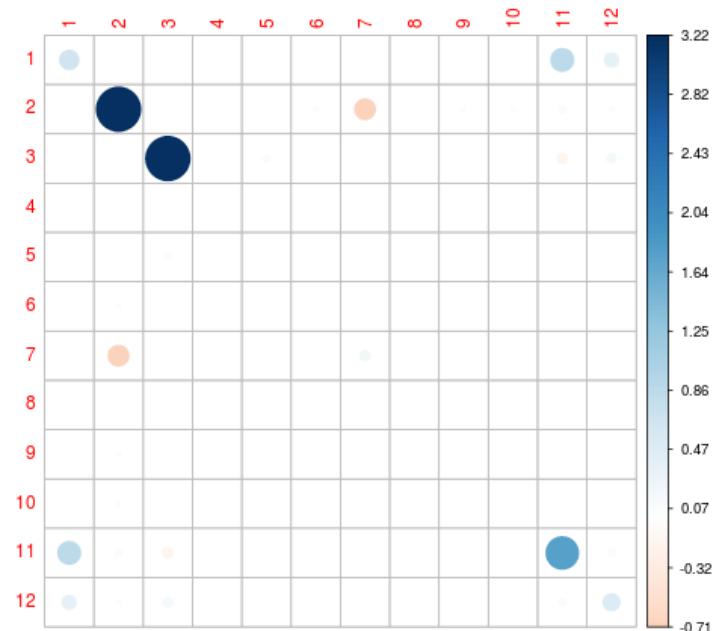
Crystal:

Unit cell: (42.275, 42.275, 39.669, 90.000, 90.000, 90.000)
 U matrix: {{ 0.8380, 0.5283, 0.1365},
 {-0.1808, 0.0328, 0.9830},
 { 0.5149, -0.8484, 0.1230}}
 B matrix: {{ 0.0237, 0.0000, 0.0000},
 { 0.0000, 0.0237, 0.0000},
 { 0.0000, 0.0000, 0.0252}}
 A = UB: {{ 0.0198, 0.0125, 0.0034},
 {-0.0043, 0.0008, 0.0248},
 { 0.0122, -0.0201, 0.0031}}

- Refinement benefits from inclusion of more data throughout the sweep
- Need time-dependent crystal model to reduce RMSDs further
- When wrong parameter fixed, crash with “Cholesky error” → non-positive-definite N
- Will be useful to study properties of the normal matrix

- Analysis of normal matrix. Not immediately insightful. 3 orders of magnitude range along diagonal
- The eigenvalues of N also range over a factor of 3000
- No reason to think this is “too much” though

```
Parameter order:name mapping
Parameter 001 : DetectorDist
Parameter 002 : DetectorShift1
Parameter 003 : DetectorShift2
Parameter 004 : DetectorTau1
Parameter 005 : DetectorTau2
Parameter 006 : DetectorTau3
Parameter 007 : SourceMu2
Parameter 008 : CrystalPhi1
Parameter 009 : CrystalPhi2
Parameter 010 : CrystalPhi3
Parameter 011 : Crystal_g_param_0
Parameter 012 : Crystal_g_param_1
```



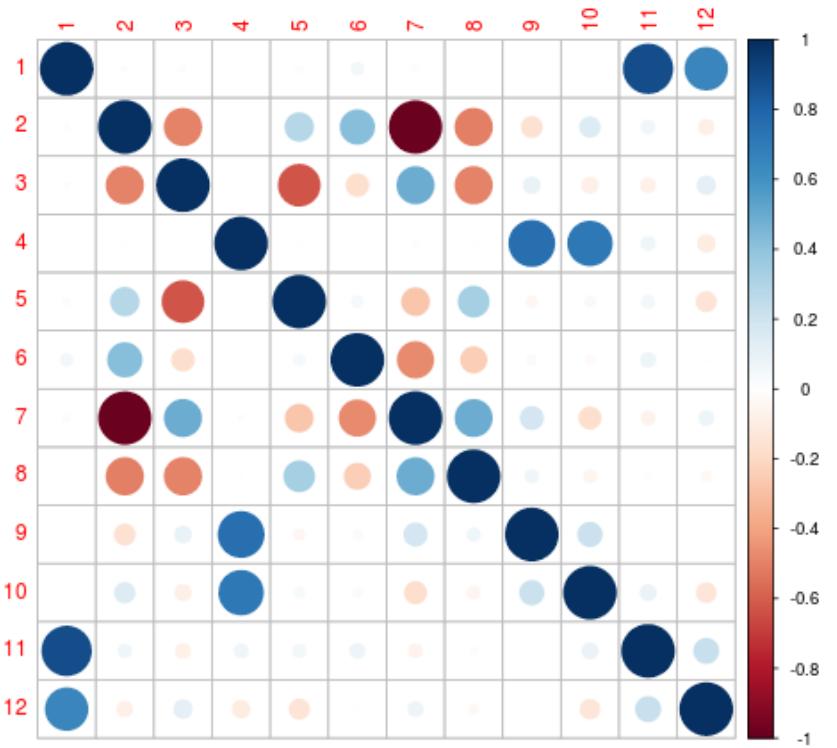
Parameter correlation

- Analysis of Jacobian $J(x)$, where
 $N = J(x)^T J(x)$
- Can J be near rank-deficient?
- Easier to calculate correlation between columns of J
- Is J ill conditioned? I don't know

```
Parameter order:name mapping
Parameter 001 : DetectorDist
Parameter 002 : DetectorShift1
Parameter 003 : DetectorShift2
Parameter 004 : DetectorTau1
Parameter 005 : DetectorTau2
Parameter 006 : DetectorTau3
Parameter 007 : SourceMu2
Parameter 008 : CrystalPhi1
Parameter 009 : CrystalPhi2
Parameter 010 : CrystalPhi3
Parameter 011 : Crystal_g_param_0
Parameter 012 : Crystal_g_param_1
```

$$J(x) = \begin{pmatrix} \frac{\partial r_1}{\partial x_1} & \frac{\partial r_1}{\partial x_2} & \dots & \frac{\partial r_1}{\partial x_n} \\ \frac{\partial r_2}{\partial x_1} & \ddots & & \\ \vdots & & \ddots & \\ \frac{\partial r_m}{\partial x_1} & \dots & \dots & \frac{\partial r_m}{\partial x_n} \end{pmatrix}$$

number of parameters, n →
↓ number of residuals, m

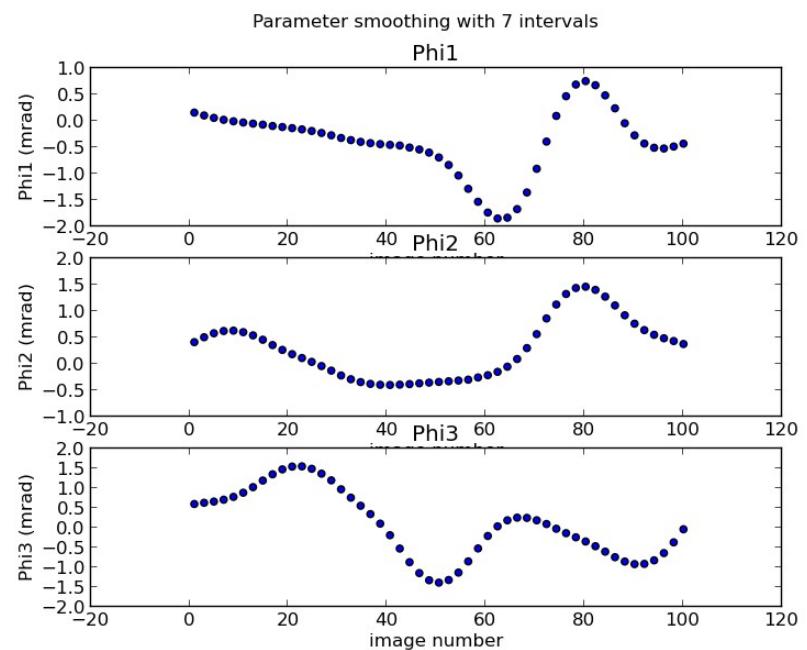


- LSTBX solves the normal equations using the Cholesky decomposition
- This is fast, but poorly behaved when J is ill-conditioned (accuracy suffers, or algorithm can even fail due to roundoff errors)
- QR is more robust, SVD even more so, at the expense of more CPU cycles
- SVD has the advantage of providing useful sensitivity information and option to filter out the smallest singular values to obtain an approximate solution less sensitive to perturbations

- This appears closely related to the Reeke/Bricogne “eigenvalue filtering” scheme
- What is the “best” way to solve the normal equations, perhaps admitting the possibility of filtering for correlated parameters?
- Can we get error estimates on the parameter values even in the case of filtering?
- Plan to modify LSTBX to implement a procedure for solving the normal equations that is appropriate for our circumstances
- Also try Levenberg-Marquardt iterations rather than Gauss-Newton (already available in LSTBX) for better behaviour when $J(x)$ is nearly rank-deficient

dials Time-dependent parameterisation

- Implemented Gaussian smoother from Aimless
- Derivatives $\partial\mathbf{U}(t)/\partial p$ and $\partial\mathbf{B}(t)/\partial p$ tested by FD for crystal orientation and unit cell parameters
- There are three adjustable variables for the smoother
 - number of samples
 - sigma
 - number of points to average
- How do we know what values are appropriate?



dials Time-dependent parameterisation

Proposed scheme for refinement:

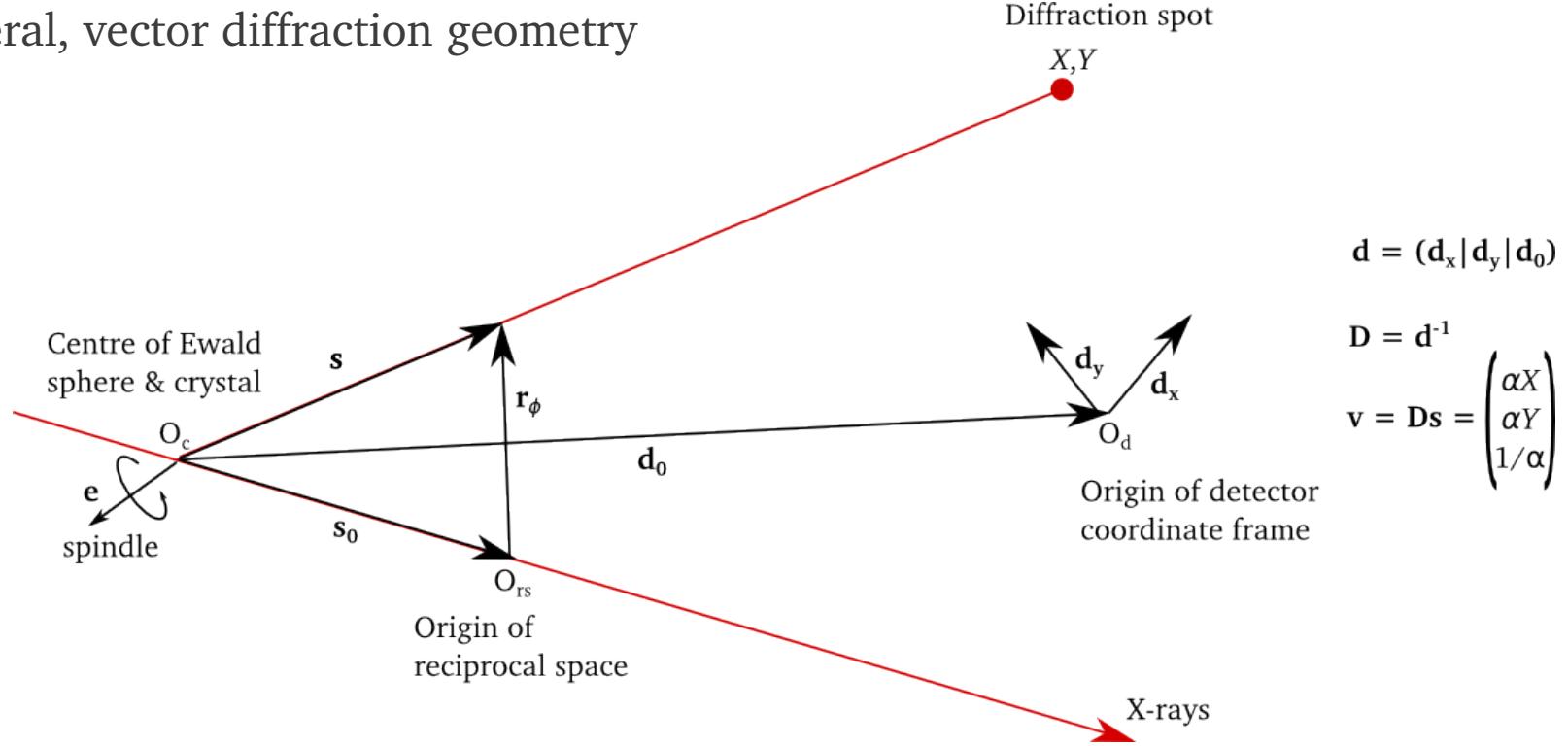
- A fully time invariant macrocycle to convergence to improve the detector and source models and define \mathbf{U}_0 and \mathbf{B}_0
- A macrocycle using time-dependent crystal parameterisations and static detector and source parameterisations
 - Parameters of the time dependent (Gaussian smoothed) models are restrained (tied) to the values that define \mathbf{U}_0 and \mathbf{B}_0
- Integration forms models for profiles, potentially improving the centroid positions
- Repeat

dials

21 May 2013



- General, vector diffraction geometry



- The detector abstract frame is a hardware-independent adapter
- Positional corrections can be accounted for in the mm-to-px mapping function

For refinement we want at least the first derivatives of predicted centroids

$$\frac{\partial \phi}{\partial p} = -\frac{\frac{\partial \mathbf{r}_\phi}{\partial p} \cdot \mathbf{s} + \mathbf{r}_\phi \cdot \frac{\partial \mathbf{s}_0}{\partial p}}{(\mathbf{e} \times \mathbf{r}_\phi) \cdot \mathbf{s}_0}$$

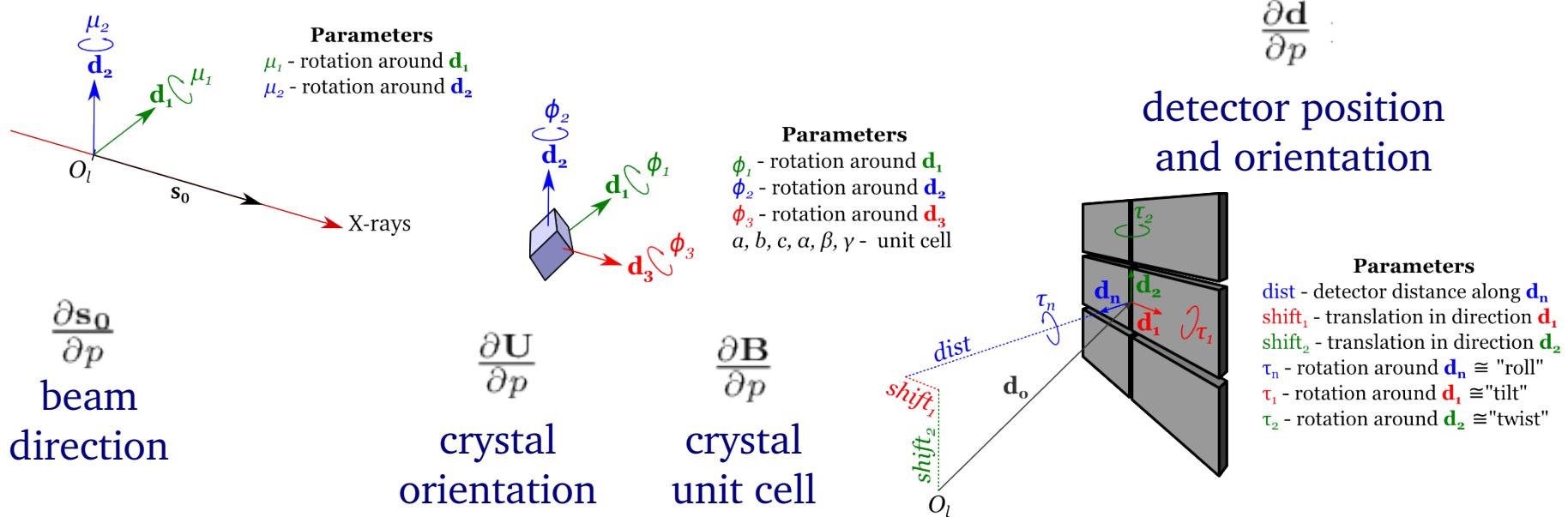
$$\frac{d\mathbf{v}}{dp} = -\mathbf{D} \frac{\partial \mathbf{d}}{\partial p} \mathbf{v} + \mathbf{D} \left[\frac{\partial \mathbf{r}_\phi}{\partial p} + (\mathbf{e} \times \mathbf{r}_\phi) \frac{\partial \phi}{\partial p} + \frac{\partial \mathbf{s}_0}{\partial p} \right]$$

Neatly, these are factored into independent models

$$\begin{array}{ccc} \frac{\partial \mathbf{d}}{\partial p} & \frac{\partial \mathbf{r}_\phi}{\partial p} & \frac{\partial \mathbf{s}_0}{\partial p} \\ \text{detector} & \text{crystal} & \text{beam} \\ & & \text{direction} \end{array}$$

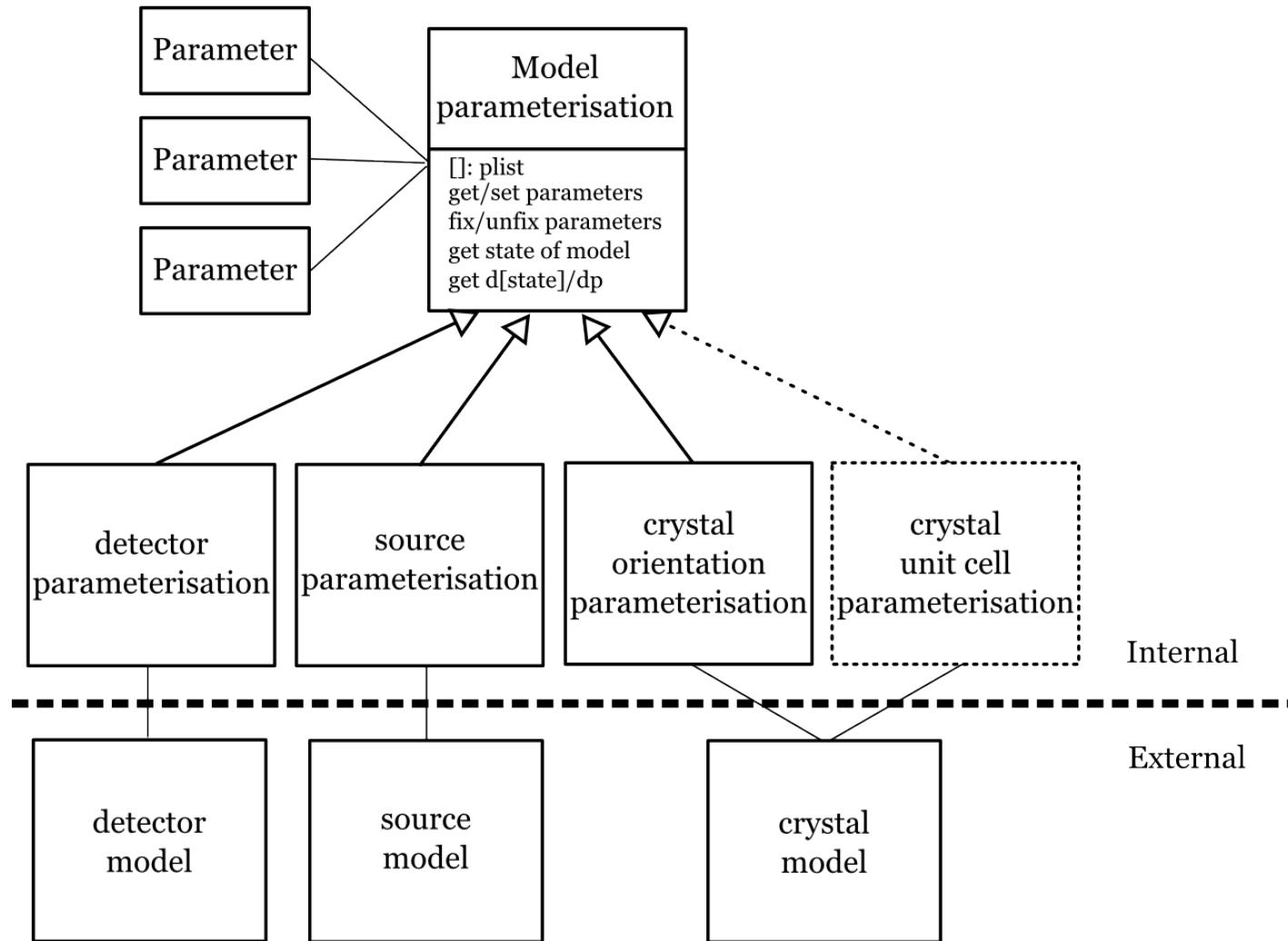
Parameterisation

Each model parameterisation provides $\partial[\text{state}]/\partial p$



- Separate 'parameterisation of prediction equation' object takes $\partial[\text{state}]/\partial p$ for each model and converts to derivatives of X, Y, ϕ for each reflection
- Individual model parameterisations are encapsulated

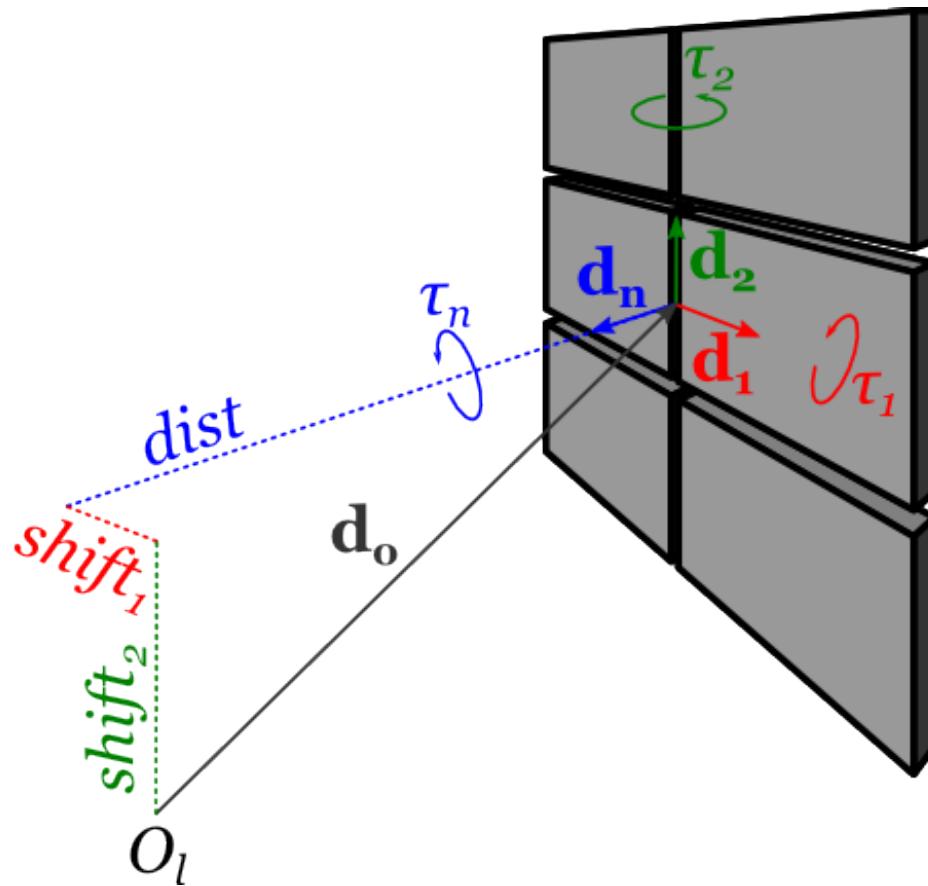
Parameterisation of experimental models



The abstract interface specifies that:

- Model parameterisations are initialised with an initial state of the model
- New states are composed by the action of functions of the parameters on the initial state
- A state and its derivatives are either a vector or a matrix
- The parameters are either distances or angles with associated unit directions

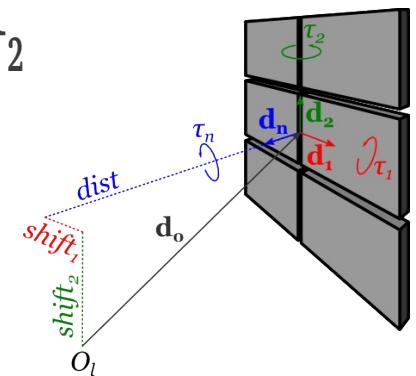
- A concrete example: detector parameterisation



Parameters

dist - detector distance along \mathbf{d}_n
shift₁ - translation in direction \mathbf{d}_1
shift₂ - translation in direction \mathbf{d}_2
 τ_n - rotation around \mathbf{d}_n \cong "roll"
 τ_1 - rotation around \mathbf{d}_1 \cong "tilt"
 τ_2 - rotation around \mathbf{d}_2 \cong "twist"

- Initial sensor matrix provides \mathbf{d}_0 , \mathbf{d}_1 , \mathbf{d}_2 , \mathbf{d}_n
- Translation parameters are immediately *dist* along \mathbf{d}_n and *shift*, $shift_1$, $shift_2$ along \mathbf{d}_1 , \mathbf{d}_2
- Initial rotation angles all 0.0, around axes \mathbf{d}_1 , \mathbf{d}_2 , \mathbf{d}_n



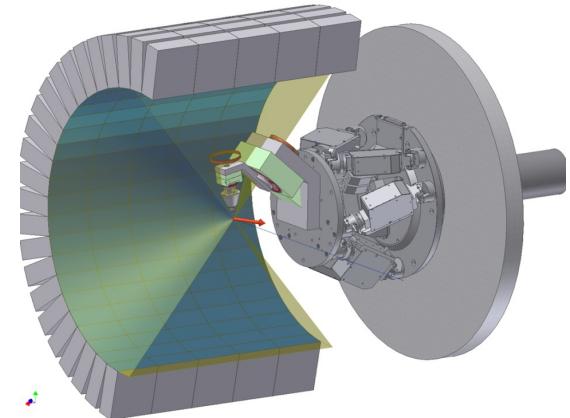
Parameterisation

- Accommodates refinement of multi-tile detectors as one rigid unit
- Each sensor panel k has its own matrix $\mathbf{d}^k = (\mathbf{d}_x^k | \mathbf{d}_y^k | \mathbf{d}_0^k)$
- These vectors are linear combinations of \mathbf{d}_0 and the local coordinate system $\mathbf{d}_1, \mathbf{d}_2, \mathbf{d}_n$ that moves with the detector:

$$\mathbf{d}_x^k = \alpha_1^k \mathbf{d}_1 + \alpha_2^k \mathbf{d}_2 + \alpha_3^k \mathbf{d}_n$$

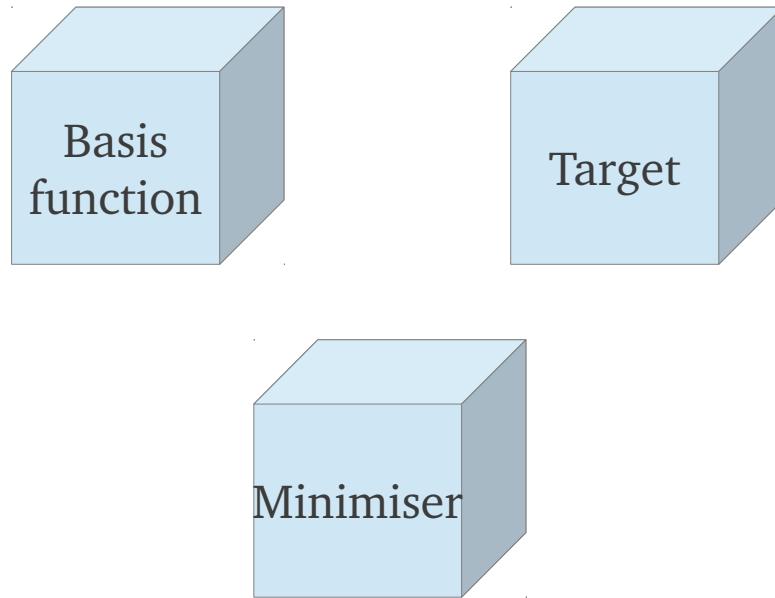
$$\mathbf{d}_y^k = \beta_1^k \mathbf{d}_1 + \beta_2^k \mathbf{d}_2 + \beta_3^k \mathbf{d}_n$$

$$\mathbf{d}_0^k = \mathbf{d}_0 + \gamma_1^k \mathbf{d}_1 + \gamma_2^k \mathbf{d}_2 + \gamma_3^k \mathbf{d}_n$$



- Thus the derivatives $\partial \mathbf{d}^k / \partial p$ for each sensor are easily calculated by linear combinations of $\partial \mathbf{d}_0 / \partial p, \partial \mathbf{d}_1 / \partial p, \partial \mathbf{d}_2 / \partial p$ and $\partial \mathbf{d}_n / \partial p$

- Further encapsulation within refinement module



- Make these independent (where possible)