POLDER, HDF5, BETA STRANDS, MD WATER

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New programs

phenix.polder

phenix.polder calculates OMIT maps by preventing the bulk solvent mask from penetrating the region of the OMIT atom selection. This tool is useful in cases where the density of the selected atoms is weak and possibly obscured by bulk solvent. Polder maps are less biased than procedures where the atoms are simply removed from the model or where the selected atoms are included in the solvent mask calculation and their occupancy is set to zero. As a larger volume is excluded from the bulk solvent, misinterpretation of bulk solvent density as OMIT density can be avoided. Polder maps are suitable for parts of the structure with weak density (such as ligands, flexible residues, alternative conformations, loop regions). phenix.polder is available from the command line and in the Phenix GUI.

Phenix News

Announcements

Amber

A new version of Amber (ambermd.org) was released 30th April 2016. This version, known as Amber16 (and AmberTools16), has been integrated into Phenix from dev-2499. Please use the documentation from the current installation to install (see FAQ page 20).

New features

Parameters of NCS search procedure

Since the last Phenix release (1.10-1) the parameters for non-crystallographic symmetry (NCS) search procedure have been changed. In the current nightly builds the...
parameters in the ncs_search scope are (with default values)

enabled = False
eclude_selection = "not (protein or nucleotide) or element H or element D"
chain_similarity_threshold = 0.85
chain_max_rmsd = 2.0
residue_match_radius = 4.0

The first parameter, enabled, turns on NCS search while exclude_selection is the selection to choose the part of a model from exclusion – NCS search will disregard the selection in the initial model in the search procedure. By default NCS is searched only in protein or nucleotide chains excluding all ligands, waters and hydrogens. A list chain pairs is generated and checked for similarity using the following criteria:

• chain_similarity_threshold: alignment of residues in prospective chains
• chain_max_rmsd: RMSD between atoms of superposed chains
• residue_match_radius: maximum allowed distance difference between pairs of matching atoms of two residues

If all the criteria are satisfied a chain pair is considered NCS-related.

More relaxed parameters are used for validation of user-supplied NCS groups. If one wishes to supply poorly related NCS groups, it is necessary to relax these parameters even more to ensure the user-supplied definitions pass the validation.

The same procedure and parameters are used to match chains of reference model with chains of refined model.

Additional or updated information can be found in documentation about phenix.refine and phenix.simple_ncs_from_pdb. See FAQ later in this newsletter about how to access documentation for new features.

cis/trans peptide bond geometry control

All geometry-based programs in Phenix can now have the peptide bond specified as cis or trans regardless of the input geometry. The phil parameter scope is useful when are peptide has adopted an incorrect local minimum and needs coercion to move to the correct confirmation.

Peptide plane control

Refinement of low-resolution structures has many challenges arising from the lack of data. One consequence is the increased number of ω angle outliers. Phenix now has a parameter that will impose a plane restraint on each peptide bond including the Cα from each residue and the intervening C, N and O atoms. This should not be used in the mid- to high-resolution ranges as there is significant deviation in the ω angle from planar as investigated by Karplus et al.

Crystallographic meetings and workshops

The 30th European Crystallographic Meeting, August 28-September 1, 2016
Location: Basel, Switzerland.

Expert advice

Fitting Tip #12 - Twist Tells: better β strands at >3.5Å in x-ray or cryoEM

David Richardson and Jane Richardson, Duke University

β sheets are just as important as α helices, and much more elegant. However, somewhere around 3.5 to 4Å resolution (a range common in the currently most exciting cryoEM and crystal structures) β strands start to merge together patchily in the density and then become a continuous slab. In the de novo trace of a new protein even when the helices are very clear, determining orientation and connectivity of β strands is still very difficult or even impossible (Baker 2012). For this purpose, a recent tool called StrandTwister (Si & He 2014) makes use of a powerful source of information apparently new to current automated methodology: the handedness and magnitude of β sheet twist.
A new tool for cryoEM

The StrandTwister software starts from a slab of density identified as probable β sheet by one of many existing methods (Kong 2003; Baker 2007; Si 2012), but which does not show clearly separated strands. It first fits a polynomial surface to the voxels in the density slab. Then it fits several potential models, with varying β-strand number and orientation, to the polynomial surface. Most importantly, it scores each of the models by the strength of its average right-handed strand twist. The algorithm produces quite accurate tracings within the top-scored few, both within simulated density (figure 1) and within initial experimental maps in the 3 to 7Å resolution range (figures 5-7 in Si & He 2014), and for either parallel or antiparallel β. However, the choice among similar-scoring models, and the connectivity of the strands with the rest of the structure, needs to be made in the context of the full molecular tracing. Executable software for StrandTwister and related information are available at (see footnote 1).

Old advice for crystal structures

StrandTwister is aimed at application to cryoEM structures but it could also be used for low-resolution x-ray. Indeed, a simpler version of the twist criterion was proposed by us in the early stages of protein crystallography, when all initial chain tracing was done manually. Figure 2 (from Richardson 1985) shows the choice of vertical vs horizontal β strands, based on which diagonal pair of density-slab corners twist up vs down. This builds on the long-known rule that β strands always show a right-handed twist as measured by peptide orientation along the strand (Chothia 1973), of between about 0° and 30° per residue (Richardson 1981). If twist were measured perpendicular to the strands, it would give a left-handed value, which is why a given slab of twisted electron density is compatible with only a narrow range of strand orientations. This simple fact can help guide either manual or automated model-building for either x-ray or cryoEM, or can be brought in by using StrandTwister. For each near-optimum strand orientation, the strand number can be estimated from the density-slab width perpendicular to the strands (figure 2).

The early papers also provide further rules that could be of use in current low-resolution modeling, such as which irregularities are frequent and which are vanishingly rare (figure 3). StrandTwister does not yet treat β barrels, but sorting out their strands could

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1 Stran twist website – http://www.cs.odu.edu/~jhe/software/strandtwister

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**Figure 1:** StrandTwister β-strand detection from simulated density maps at 10 Å. The best of the top ten sets of detected β traces (red lines) are superimposed with the backbone of the β strands (dark gray) and the density map (pale gray). A large β sheet in 1UD9_B is shown in top view (left) and in side view, which has maximum right-handed twist (right). (This is a simpler version of Figure 4C from Si & He 2014)

and helices make about a 40° angle with the barrel axis.

Antiparallel β barrels are more difficult, because their shapes vary and their overall cross-section changes only slowly with strand number. However, if strand number can be guessed from predictions or from the overall trace, they can then be fairly well placed using empirically observed correlation of shape, twist and strand number. These relationships can be visualized in terms of the strand-crossing angle between front and rear sides of the barrel, as seen in figure 4. Cylindrical barrels, with some β-sheet H-bonding all the way around, average a strand-crossing angle of about 90° for 6 strands (figure 4a), and don’t get much below ~70° even for 10 or 11 strands (e.g., in the “can” of green fluorescent protein). If H-bonding is continuous around one end but open on the other, which splits open one end of the cylinder of density as seen for superoxide dismutase in figure 4 of Richardson 1985, the strand-crossing angle is usually near 40° (figure 4b). If β-sheet H-bonding is broken on both sides, to form a β sandwich, then the two sheets can be analyzed separately, but it may help to know that the strand-crossing angle between them is seldom much less than 30° (figure 4c).

The bottom line
Don’t give up if your density map has seemingly featureless slabs of density for its β
Figure 4: Strand-crossing angles between front and back sides of antiparallel β barrels. (a) The 6-strand, fully H-bonded barrel in trypsin has a strand-cross angle of 90°. (b) The 8-strand barrel in Cu,Zn superoxide dismutase, with one opening between strands, has a strand-cross angle of 60°. (c) The 10-strand β sandwich in tomato bushy stunt virus capsid has a strand-cross angle of 30°.

sheets. Leveraging information from the twist of that density, in addition to its size and shape, may well determine the number and orientation of the β strands quite closely.

References
Richardson JS, Richardson DC (1985) "Interpretation of electron density maps", *Meth Enzymol* 115: 189–206
FAQ

Why is the documentation on the website different from what I’m seeing in my installation?

Major numbered versions of Phenix are released periodically to provide a stable highly tested installations. There are also nightly builds of Phenix (and some external programs like Amber, AFITT and Rosetta) so test the software suite on several operating systems. This provides the ability to release very up-to-date versions to disseminate new programs and features as well as bug fixes.

Documentation generation is part of this nightly testing so that current information can be packaged with each version’s installer and accessed via the GUI or typing

    phenix.doc

on the commandline.

The documentation on the Phenix website is associated with the major release. If you have a nightly build or wish access to a feature added since the last major release, you will need to access the documentation from a nightly build. This requires installing the latest nightly build and reading the documentation on your computer.
Using Molecular Dynamics Simulations to Enrich the Water Structure in Biomolecular Crystals

Irem Altan¹, Pavel V. Afonine² and Patrick Charbonneau¹

¹Department of Chemistry, Duke University, 124 Science Drive, Box 90346, Durham, NC 27708, USA.
²Molecular Biophysics and Integrated Bioimaging Division, Lawrence Berkeley National Laboratory, Berkeley CA 94720, USA

Introduction
Describing the solvation of biomolecules is challenging due to their complex surface geometry with a mosaic of hydrophobic and hydrophilic regions, as well as the intrinsically probabilistic nature of solvation itself. The problem also carries over to crystals of biomacromolecules, which contain between 20 and 90% solvent by volume with an average of about 50% (Weichenberger et al., 2015). The crystal solvent is traditionally considered in two categories: ordered and disordered (or bulk) solvent (figure 1). Solvent at the biomolecule surface constitutes the ordered part and can be probed by diffraction. This portion is modeled using an atomic model to describe distinctly resolved features in the residual Fourier F₀-Fc map. Solvent distant from the protein surface is disordered and cannot be visualized as individual molecules. The disordered part is modeled as a region of flat electron density. Clearly, these two descriptions are overly simplistic and address only two opposite extremes of a spectrum. The reality, of course, is more complex: bulk-solvent density may vary locally (Burling et al., 1996, Lounnas et al., 1994; Sonntag et al., 2011) and the signal arising from semi-ordered solvent may be at or below the noise level in F₀-Fc map. It is therefore not surprising that even high quality X-ray structures of biomolecules can have R-factors as high as 20%; a significant component of this discrepancy may be due to the imperfections in the solvent model (Holton et al., 2014). Further improvement of crystal structure solvent description may be beneficial for:

1. providing higher-quality structural models,
2. assessing models routinely used for simulating water and
3. revealing information that is not directly present in the X-ray diffraction patterns, such as protonation states.

Molecular dynamics (MD) simulations’ natural capability to treat arbitrary surfaces and the probabilistic nature of solvation has inspired us to develop a methodology to explore and eventually exploit these strengths to complement existing solvent descriptions. Here we document the work in progress in its rather initial stage.

MD simulations
The electron density obtained from diffraction data is an average over both unit cells and time, with each configuration contributing to the diffraction proportionally to its occurrence. For an ergodic fluid, the time

Figure 1: Protein unit cell showing the protein and the water model. Bulk solvent region is shown in blue while red spheres denote crystal waters.
distribution of water molecules within a given unit cell is equal to their instantaneous distribution in all unit cells. Hence, peaks to which crystal waters are typically assigned correspond to a high occupancy probability. Because that occupancy can be far below unity, approaches that can sample it more systematically can be potentially far-reaching. MD simulations, for instance, provide a naturally probabilistic description of solvation. Its biggest weakness is that the probability distribution it actually samples may differ from that observed in experiments, because it relies on imperfect (classical) force fields. In order to examine how well it fares, we started simulating water and protein within the crystal unit cell. Averaging over only the water density in the MD snapshots (keeping the biomolecule fixed) yields an estimate of the solvent distribution in the unit cell.

We use a test protein, (PDB ID: 1YTT), for which experimental phases are available, allowing us to probe the solvent density without bias from the protein model (Burling et al., 1996). The simulated system uses a protein in its unit cell with water molecules from different water models used to describe water in MD simulations, in order to assess their ability to reconstruct the solvent density. We use Gromacs (Berendsen et al., 1995) to run MD simulations. For each MD snapshot, we generate model-calculated electron densities using Phenix (Adams et al., 2010).

We then average over these snapshots to generate the electron density map. For the experimental map, we generate a \( \{ F_{\text{obs}}, \varphi_{\text{obs}} \} \) absolute scaled Fourier map. Once the electron density maps from MD simulations and experimental data are calculated, we compare the radial distribution (figure 2) of water around surface atoms. Calculating correlation coefficients to quantify the agreement between MD and experimental densities, we find that the experimental and MD-derived radial distribution functions correlate well, with Pearson correlation coefficients above 0.9. The various tested water models perform essentially identically.

**Inferring protonation states**

Unless the diffraction data has sub-atomic resolution (~1.0-0.7Å or better), it is not possible to directly detect hydrogen atoms from X-ray crystallography. Nevertheless, whether a proton is present or not affects the nearby solvent density. Running simulations with different protonation states and comparing the resulting water density with experiments help infer protonation states. For 1YTT, for instance, in a histidine residue (HIS116, chain A), protonating N\( \delta \)1 instead of N\( \epsilon \)2 results in clear peaks in the MD density that overlap with two crystal waters (figure 3). This effect is particularly remarkable because different prediction tools disagree: while Molprobity correctly assigns a proton to N\( \delta \)1 only, Gromacs protonates N\( \epsilon \)2. Note, however, that this strategy is only

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**Figure 2:** Radial distribution functions, calculated by averaging electron density in shells extending from the protein surface, for surface nitrogens, oxygens, and carbons. (Shown for experimental densities (black) and various water models).
**Figure 3:** For a particular histidine (residue number 116, chain A), the surrounding water density suggests that Nε2 is not protonated. Gromacs, by analyzing the hydrogen bond network, protonates the wrong nitrogen, while Molprobity (Chen et al., 2010), which takes clashes into account, assigns it correctly. The maps shown here are calculated from MD snapshots in units of $e^2/\text{Å}^3$ (shown contoured at 0.6 $e^2/\text{Å}^3$).

applicable to atoms with sufficient solvent exposure.

**Conclusion**
While MD water models appear to perform well when averaged radially, they may not be sufficient for accurately reconstructing the real space solvent density on their own. Interestingly, we find that shortcomings in sampling methods and protein force fields are as much of a problem to accurately reconstructing water density as the MD water models themselves. We are nevertheless actively working on strategies for assigning protonation states and for developing hybrid schemes that utilize both experimental data and information from MD simulations that is complementary to the experimental data to describe solvation.

**References**
Introduction
Over the past six months, a new generation of detectors has arrived at several macromolecular crystallography beamlines. EIGER sets new standards in terms of readout time, achievable frame rate and pixel size. The largest model, EIGER X 16M, has been installed at beamline X06SA of Swiss Light Source (SLS), beamline FMX of National Synchrotron Light Source II (NSLS-II) and at GM/CA of Advanced Photon Source (APS). EIGER X 9M detectors are in operation at beamlines PROXIMA-2 of Synchrotron SOLEIL and BL32XU of SPring-8 and at LS-CAT of APS. EIGER X 4M detectors operate at beamline BL-1A of Photon Factory and at beamlines ID13 and ID30-3 of the European Synchrotron Radiation Facility (ESRF). More are scheduled to be delivered throughout 2016.

Because of high pixel counts (16 million active pixels for the EIGER X 16M) and high maximum frame rates (750 Hz for the four-million-pixel EIGER X 4M or corresponding regions of interest in the EIGER X 9M and the EIGER X 16M), EIGER detectors produce raw data at rates unprecedented for macromolecular crystallography. They also generate output in HDF5, a big-data format that is new to the crystallographic community. Together, this has caused considerable alarm, especially in environments, e.g., individual academic or industrial labs, that are currently working close to their limits of throughput and processing. In terms of data volume, advances in compression should alleviate most fears. Nevertheless, updates to computer hardware and processing scripts are recommended for an uncomplicated user experience.

In this contribution, which is an update to a report in Jan 2014 (Bernstein et al. 2014), we will demonstrate that working with EIGER data is not considerably more challenging than working with PILATUS data. In contrast to all previous discussion of this topic, this is the first time that practical user experiences and collected HDF5 data are available. We will briefly describe the rationale behind choosing HDF5, explain the implementation of the format, and give advice on how to work with and process data collected with EIGER. These points are all directly addressed to crystallographers that face the prospect of collecting data with EIGER or are already doing so.

High-data rate crystallography
With the advent of X-ray free-electron lasers, macromolecular crystallography has entered a new era, joining fields like particle physics, climate modeling and astronomy in a world of big data. Datasets can comprise hundreds of thousands of images and amount to hundreds of gigabytes. The ideas pioneered at XFELs are now percolating down to synchrotrons, where attempts to assemble complete datasets from highly partial diffraction data obtained from a large number of randomly oriented small crystals have become known as synchrotron serial crystallography (SSX, e.g., see Coquelle et al. 2015).

Traditional crystallography is also progressing. Thanks to noise-free photon detection by PILATUS and EIGER, the permissible x-ray dose can be spread over arbitrarily many images, with commensurate increases in multiplicity and useful anomalous signal (Weinert et al. 2015). Furthermore, work at SLS has shown that EIGER produces better data when oscillation images are finer phi-sliced than is recommended for PILATUS (Mueller et al. 2012). Instead of half the crystal mosaicity (as reported by XDS), much smaller oscillation increments are recommended when working with EIGER (Casañas et al. 2016). An inexorable corollary of best data is thus bigger datasets.

To deal with big data, it makes sense to use tools that have proven powerful in other fields that face similar problems. The hierarchical data format (HDF, see link in section "Links to software and documentation") has been designed and is widely used for the storage of large datasets like those obtained by NASA’s Earth Observing System, high-resolution MRI brain scans, financial modeling and many more. It is the file format of choice for all
Figure 1: Illustration of NeXus tree inside HDF5 file. The structure of an HDF5 file of a 3600-image dataset is shown. Inside the master file (right), there is a broad division into information about the detector (/entry/instrument/detector), information about the sample (/entry/sample) and the measured data (/entry/data). Note that the /entry/data group does not contain the actual diffraction data but links to any number of separate data files (left).

HDF5 and NeXus
Early in 2013 when EIGER was in development, DECTRIS invited developers of processing programs, software engineers responsible for beamline infrastructure and experts on data formats to a workshop at its premises in Baden, Switzerland. Over the course of two days, the company’s scientists and software developers solicited advice from a representative selection of stakeholders on the best format for EIGER, the new generation of HPC detector that was then in development. Presentations given at the workshop are available for download (see link list). There was broad consensus that DECTRIS should adopt HDF5 as the file format and represent data in NeXus.

NeXus is a set of design principles for organizing neutron, x-ray and muon data inside HDF5 (or xml). This includes not only the experimental measurements but also relevant metadata to describe the experiment and guarantee processability without additional information. The metadata are also important for data archiving and – especially – the usefulness of retrieved archived data. To ensure compatibility between HDF5 files generated at different experimental facilities, NeXus defines a dictionary of specific field names and lays out rules for organizing them. Thus, NeXus can be considered a standard for encoding experimental metadata in HDF5. NeXus fields lie on branches of a hierarchical structure, as illustrated in figure 1.

By design, all data and metadata could be saved in one HDF5 file. This would result in the simplest
Table 1: Advantages and disadvantages of storing crystallographic data in NeXus/HDF5.

<table>
<thead>
<tr>
<th>Advantages</th>
<th>Disadvantages</th>
</tr>
</thead>
<tbody>
<tr>
<td>• Exhaustive metadata</td>
<td>• Unfamiliar to users</td>
</tr>
<tr>
<td>• Better compression than CBF</td>
<td>• Poor support by processing programs</td>
</tr>
<tr>
<td>• Dataset-centric approach</td>
<td>• Parallel processing not supported by HDF5 library</td>
</tr>
</tbody>
</table>

file structure but poses problems because of the size of crystallographic datasets. Metadata are thus saved in one file (called the master file) and image data in one or more data files depending on the size of the images. In terms of system and network loads, saving somewhere between 50 and 100 images per data file seems a reasonable approach.

The master file and the data files share a file name root (e.g. set1_master.h5, set1_data_000001.h5, set1_data_000002.h5, etc.) and should always remain together. The master file contains links to the data files such that for programs displaying images (e.g. Albula) or processing datasets (e.g. DIALS, see website list), only the master file needs to be given as an argument.

HDF5 fully comprises the Image-supporting Crystallographic Information File (imgCIF) and the Crystallographic Binary Format (CBF) and its derivative, miniCBF (popularized by PILATUS). Since version 0.9.2.12, CBFlib (see link list) can store arbitrary CBF files in HDF5 and recover them. It supports use of all CBFlib compressions in HDF5 files and can convert sets of miniCBF files to a single NeXus HDF5 file. Thus consistency is ensured between image formats that the community is familiar with and the new addition, both of which serve to hold crystallographic data.

Table 1 summarizes the advantages and disadvantages of using NeXus/HDF5 in macromolecular crystallography. The exhaustive definition of metadata items by the NeXus standard allows for comprehensive documentation of the experiment. As crystallographers are rarely interested in individual images, it makes sense to convert to a dataset-centric view and use the datasets as basic experimental units. The much-reduced number of files helps with organization and puts less strain on hardware during file system operations, especially at large facilities with shared computer resources.

Since version 1.5 of the firmware, EIGER compresses image data using the bitshuffle-LZ4 algorithm (see link list) by default. This algorithm is more efficient than CBF (table 2) and makes EIGER datasets smaller than PILATUS datasets on a pixel-by-pixel basis. The speed of compression is compatible with the rate of data generation and adds no appreciable delay. As the compression algorithm is available as a filter plugin in HDF5, its use is fully transparent to the end user.

Processing of HDF5 data

Processing of diffraction data saved in HDF5 format is possible with commonly used programs, but adaptations are required in some cases. Table 3 gives an overview of the processing choices. DIALS (see link list), essentially the first new software package for analyzing crystallographic oscillation data in two decades, is currently the only program that can read HDF5 natively. It extracts metadata from master.h5 and processes data without requiring additional parameters or, if called by the xia2 pipelines, user input. DIALS was used to process the data that led to the first publication with EIGER X 16M data (Yamano et al. 2016).

XDS (see link list) accepts HDF5 files as input (NAME_TEMPLATE_OF_DATA_FRAMES=my_name _??????_h5) but temporarily extracts images to CBF format for processing. For this to work, the H5ToXds utility (Linux only, see link list) must be in the path. As always, the user has to provide a file containing processing parameters (XDS.INP), which can be created either manually upon inspection of the master file with an HDF viewer (e.g. HDFview, see link list) or by a script that extracts the values from the master file. An example of such a script is given in the EIGER section of the XDS wiki (see link list).

Mosflm (see link list) and HKL (see link list) obtain processing parameters from image headers and thus require CBF images with intact mini-CBF headers. The open source tool eiger2cbf (Linux
**Table 2: Compression efficiency of bitshuffle-LZ4.** EIGER X 16M datasets collected at SLS under different conditions were compressed with CBF and bs-LZ4. Values for uncompressed images in 16- and 32-bit mode are shown for comparison. Note that images acquired at frame rates below 50 Hz are saved in 32-bit mode. Otherwise they are saved in 16-bit mode.

<table>
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**Table 3: Strategies for processing HDF5 data.** Different degrees of manipulation are required to process crystallographic data in HDF5 format, from none at all (when using DIALS) to creating complete mini-CBF files (when using Mosflm or HKL).

<table>
<thead>
<tr>
<th>Strategy</th>
<th>Program</th>
<th>Processing pipeline</th>
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<tbody>
<tr>
<td>Direct processing of HDF5 data</td>
<td>DIALS</td>
<td>xia2</td>
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<tr>
<td>Use of an integrated extraction utility</td>
<td>XDS</td>
<td>autoPROC, xia2</td>
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<tr>
<td>Conversion to CBF prior to processing</td>
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<tr>
<td></td>
<td>HKL</td>
<td>HKL2000</td>
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and OSX, see link list) has been developed for this conversion. Since master.h5 contains many more metadata items than mini-CBF files (as written by PILATUS detectors), some information is inevitably lost during the conversion. For example, rotation axes that turn opposite to convention (e.g. at SPring-8 BL32XU) must be recognized through lookup tables in the processing software instead of being defined in master.h5 (figure 2). Users are thus advised to use the converted CBF images for processing only and keep the original HDF5 files for archiving.

Because of the flexibility of the NeXus tree inside HDF5, it is possible to save the processing history together with all processing parameters and output inside master.h5 and thus jointly with the data. The NeXus base class NXprocess should be used for that. One could have any number of NXprocess groups for processing runs with different software suites and different processing parameters, all neatly grouped and properly documented. Automation of this process is unfortunately still some way off.
Figure 2: Detail of NeXus tree. The attributes of the /entry/sample/transformations/omega dataset show the rotation axis opposite to convention (-1, 0, 0).

Despite a raw size three times as large, compressed EIGER 16M images are of similar size as, or even smaller than, PILATUS 6M images. However, as the recommended data collection strategy will call for the acquisition of more images with smaller individual oscillation increments (Casañas et al. 2016), EIGER datasets will tend to be larger than what crystallographers are used to. In addition, the higher achievable frame rate of EIGER and accelerations of automatic crystal mounting and centering will let users collect more datasets in a given time. This side effect of better and faster hardware is not unique to crystallography. In fact, crystallographers are well advised to talk to their colleagues in electron cryomicroscopy who have been grappling with similar issues since the introduction of high-frame rate direct electron detectors.

High data-rate MX meeting
In May 2016, Herbert J. Bernstein, Nicholas K. Sauter and Robert M. Sweet organized a meeting at NSLS-II where developers of crystallographic processing software, beamline software engineers and representatives of DECTRIS discussed bottlenecks in EIGER data collection, transfer and processing at synchrotron beamlines and possible ways of overcoming them. The results of this meeting will be dissipated on a dedicated website
Computer requirements for transfer and processing of EIGER data were the subject of lively debate at the meeting, and it was concluded that rigorous benchmarking is necessary before firm recommendations can be made. Nevertheless, three general suggestions are justified for individual academic labs. For processing, one should invest in a powerful multi-socket machine to leverage the parallel processing features that various software packages offer. In most cases, this will mean working under Linux. For processing with XDS, enough RAM should be available to extract temporary CBFs there, which substantially increases processing speed over extracting to disk. Lastly, enough redundant storage space should be available for the archiving of collected data.

Archiving and adding metadata to master.h5
One of the great advantages of saving diffraction data in HDF5 lies in the ease of archiving. Even for large datasets, only a small number of files have to be transferred and stored, and annotation is inherent in the master file. Because of the separation of datasets into master file and any number of linked data files, it is crucial not to rename the data files. Otherwise broken links will dissociate the metadata from the diffraction data. Processing will not be possible anymore, and reassociating the data with the correct master file is nearly impossible. As the stem the file names of corresponding master and data files are identical, it is recommended to move datasets using shell wildcard characters, e.g.

```
mv series_1_*_.h5 /path/2/my/archive
```

The master file contains metadata describing the detector, the data and, to some limited extent, the experiment. In its description of the metadata, it follows the NeXus standard, specifically the NXnxm application definition formulated for macromolecular crystallography (see link list). EIGER writes metadata to the master file to identify and describe the detector and record experimental details that are known to the detector. Values are set during data collection and include quantities like X-ray wavelength, exposure time and the x and y coordinates of the direct beam on the detector.

For proper archiving, a more comprehensive record of experimental and peripheral information than what can be provided by the detector is necessary. The IUCr Working Group on Diffraction Data Deposition (WGDDD) is working on defining a minimum standard for this. We urge developers of beamline software to expand the master file written by the detector with as much relevant information as possible and always follow the standards laid out by NeXus. Beamline users are encouraged to demand full NeXus-compliant metadata with the data they collect.

Conclusion
With EIGER, a new manner of working with crystallographic data is coming to synchrotrons around the world. Thanks to its unprecedented speed, it is helping serial crystallography transition from free-electron lasers to synchrotrons and making grid scans feasible with ever-smaller beams. The inevitable side effect of these new approaches is unprecedented amounts of data. The file format HDF5 was chosen by the MX community to cope with these data. Instead of saving individual images, few HDF5 files are written to store entire datasets. This accelerates file system operations and helps with organization. The NeXus definitions allow (and even demand) the inclusion of metadata that will prove invaluable for data archiving.

Despite the change in format, the workflow familiar to crystallographers remains largely unchanged. Data can be processed easily, even though at present most programs require extraction of images to CBF, which adds some time to the processing and requires additional temporary space. Once programs have been updated to be fully conscious of HDF5, NeXus NXmx-compliant metadata will enable straightforward automatic processing.

Acknowledgements
We thank the NeXus International Advisory Committee for fruitful discussions on NXmx compliance of EIGER HDF5 data.
Links to software and documentation

- HDF website: [https://www.hdfgroup.org](https://www.hdfgroup.org)
- CBFlib: [https://sourceforge.net/projects/cbflib/](https://sourceforge.net/projects/cbflib/)
- EIGER workshop 2013: [https://www.dectris.com/presentations.html](https://www.dectris.com/presentations.html)
- bs-LZ4 filter: [https://github.com/kiyo-masui/bitshuffle](https://github.com/kiyo-masui/bitshuffle)
- Mosflm: [http://www.mrc-lmb.cam.ac.uk/harry/imosflm](http://www.mrc-lmb.cam.ac.uk/harry/imosflm)
- eiger2cbf conversion utility: [https://github.com/biochem-fan/eiger2cbf](https://github.com/biochem-fan/eiger2cbf)
- HSToXds extraction utility: [https://www.dectris.com/HSToXds.html](https://www.dectris.com/HSToXds.html)
- NXmx application definition: [http://download.nexusformat.org/sphinx/classes/applications/NXmx.html](http://download.nexusformat.org/sphinx/classes/applications/NXmx.html)
- HDRMX: [http://hdrmx.medsbio.org](http://hdrmx.medsbio.org)

References


Processing XFEL data with cctbx.xfel and DIALS

Aaron S. Brewster\textsuperscript{a}, David G. Waterman\textsuperscript{b}, James M. Parkhurst\textsuperscript{c}, Richard J. Gildea\textsuperscript{c}, Tara M. Michels-Clark\textsuperscript{a}, Iris D. Young\textsuperscript{a}, Herbert J. Bernstein\textsuperscript{d}, Graeme Winter\textsuperscript{e}, Gwyndaf Evans\textsuperscript{e} and Nicholas K. Sauter\textsuperscript{a}

\textsuperscript{a}Lawrence Berkeley National Laboratory, Berkeley, CA 94720
\textsuperscript{b}CCP4, Research Complex at Harwell, STFC Rutherford Appleton Laboratory, OX11 0FA (UK)
\textsuperscript{c}Diamond Light Source, Harwell Science and Innovation Campus, OX11 0DE (UK)
\textsuperscript{d}School of Chemistry and Materials Science, Rochester Institute of Technology, Rochester, NY 14623

Correspondence email: asbrewster@lbl.gov

1. Introduction

Protein crystal diffraction data from X-ray free-electron lasers (XFELs) pose difficult challenges to conventional data reduction software. In a typical XFEL experiment, short pulses of photons tens of femtoseconds long containing \(10^{12}\) photons/pulse interact serially with thousands of individual crystals, each producing a single pattern before the intense light destroys the crystal. These diffraction patterns represent ‘still’ shots and must be treated differently than rotation datasets collected using a goniometer. For example, with only one shot, refining the crystal orientation around the x- and y-axes of rotation orthogonal to the beam becomes difficult, as the rotation around these axes does not affect the locations of reflections, but only which reflections are in the diffracting condition. Furthermore, without a way of measuring the rocking curve directly by transitioning reflections through the Ewald sphere during a crystal rotation, it is difficult to produce estimates of mosaicity and thus predict which weak reflections will be in the diffracting condition. For these and other reasons described elsewhere, we have implemented the package cctbx.xfel, based on cctbx (Grosse-Kunstleve et al. 2002), which includes specializations of known indexing and refinement algorithms specific for the stills case (Sauter et al. 2013, Hattne et al. 2014).

Further challenges unrelated to the physics of crystallography are encountered when processing XFEL data. First, the short pulse length makes photon-counting detectors unsuited for recording XFEL data. Integrating detectors such as those used with charged-coupled devices (CCDs) are more appropriate, but in order to handle the high pulse rate typical of XFELs (120 Hz at the Linac Coherent Light Source (LCLS)), new detectors were created, such as the Cornell-SLAC Pixel Array Detector (CSPAD) (Hart et al. 2012). The CSPAD is composed of 32 separate sensors, the positions of which are not precisely known.

Finally, diffraction data collected at 120 Hz must be reduced using large computing clusters with memory, file I/O and network capacity capable of keeping pace with the experiment in a timely manner, so that a data reduction team can provide live feedback to beam line operators and scientists, allowing them to change sample measurement conditions as needed.

This article illustrates features and processing patterns of a typical XFEL experiment and provides commands for running cctbx.xfel in conjunction with the new software for reducing data from difficult systems, DIALS (Diffraction Integration for Advanced Light Sources) (Waterman et al. 2013). DIALS is built on the cctbx toolkit and takes advantage of many of its features, including image file reading from a variety of formats, crystal symmetry libraries and minimization engines. Given a collection of diffraction images, DIALS produces a series of models of the experiment, describing the detector, beam and crystal, and, as appropriate, goniometer and scan (Parkhurst et al. 2014). It also implements new indexing, refinement and integration algorithms, continuing in the
‘toolbox’ tradition of open source and object-oriented programming set in place by *cctbx*.

### 1.1. System overview

The ability to do serial X-ray crystallography has relied on the convergence of several critical technologies, such as the X-ray source and sample injection systems. Reduction of XFEL diffraction data requires further technologies including a) a detector capable of recording XFEL pulses at speeds matching the source, b) parallel computing infrastructure, c) new and adapted file formats for data storage and d) data reduction software. Here, we provide a brief overview of these technologies.

### 1.2. The detector

XFEL pulses challenge detector technology. The CSPAD detector is capable of integrating high signals from the ultra-bright XFEL source, while operating at 120 Hz. Two CSPADs are installed at the CXI end-station at LCLS; another is installed at the XPP end-station. The detector comprises 32 sensors, each consisting of 2 Application-Specific Integrated Circuits (ASICs) 194x185 pixels in dimension. The sensors are arranged in 4 quadrants of 8 sensors each. The CSPADs at CXI have their quadrants each on a diagonal rail that allows tuning the size of the central aperture through which the transmitted beam must pass, while the quadrants of the CSPAD at XPP are in a fixed arrangement. Each detector is regularly upgraded and improved and sensor positions vary as a consequence, although these are measured using an optical microscope. Indexing, predicting spot locations using a crystal orientation matrix and integrating reflection intensities require precise knowledge of the locations of these sensors in three-dimensional space (Hattne et al. 2014). For this reason, a portion of this article describes the calibration and refinement of the sensor or tile metrology using a reference dataset.

Each pixel in the detector can be configured to a low or high gain setting. The low gain setting has a full well capacity of 3000 photons, while the high gain setting saturates about seven times as quickly but has a higher sensitivity (Hart et al. 2012). This feature allows the user to specify, for example, a circular gain mask that sets low resolution pixels to low gain mode to avoid saturation of intense low resolution reflections while keeping high resolution pixels in high gain mode to capture weak diffraction data near the detector limits.

Other detectors in use at XFEL sources for protein crystals include the octal sensor detector at SACLA (Kameshima et al. 2014), a Rayonix MX 170 HS detector at XPP (Chollet et al. 2015) and a MAR 325 CCD detector, also at XPP at LCLS, occasionally brought in from SSRL for use with fixed target experiments (Cohen et al. 2014). Each of these detectors has its own set of tradeoffs and *cctbx.xfel* has been used to process data from all of them.

### 1.3. Parallel computing

Recording at 120 Hz yields 72000 2.2 megapixel CSPAD images in a typical 10 minute LCLS run. Processing this volume of data without a parallel computing environment quickly becomes impractical. The clustering environment at LCLS is ideal. Hundreds of nodes can be harnessed with 12 - 16 computer cores each to greatly accelerate indexing and integration. The program `cxi.mpi_submit`, a component of *cctbx.xfel*, provides an interface for submitting processing jobs to the LCLS cluster.

Additionally, we have collaborated with NERSC (National Energy Research Scientific Computing center) to transfer the data streams from the CSPAD detectors to the NERSC clustering systems for processing (Kern et al. 2014). NESRC is utilized for some of the largest data reduction problems world wide, including climate and astrophysics simulations and is ideal for efficient, parallel reduction of data from serial crystallographic experiments.
1.4. Data reduction software

Diffraction data recorded on CSPADs at LCLS is streamed by dedicated Data Acquisition Systems (DAQs) to container files in XTC format. The programmatic interface to interact with these files at LCLS is psana (Damiani et al. 2016) cctbx.xfel was originally designed to use the older pyana interface (Sauter et al. 2013, Hattne et al. 2014) and has transitioned to psana while maintaining backward compatibility.

psana uses a calibration store to read frames and apply pixel corrections such as dark current subtraction and common mode correction and it is designed with computational parallelization in mind. As each image is independent, multiple computer cores can process images in parallel. cctbx.xfel interfaces with psana to read XTC streams, parse the LCLS metrology file that describes the layout of the CSPAD and load pixel data for each image. Next, cctbx.xfel uses DIALS or LABELIT to create detector and beam models, find spots, index the reflections, refine the experimental model and integrate the reflection intensities. The user specifies processing parameters in cctbx-style parameter files (PHIL files, see below) and passes the parameter file to cctbx.xfel, which calls psana and submits the job to the queueing system. Specific details are described below and in online tutorials at http://cci.lbl.gov/xfel.

cctbx.xfel is the XFEL data reduction package developed by the Computational Crystallographic Initiative at LBNL and is installed for all users at LCLS.

1.4.1. cctbx.xfel indexing, refinement and integration back ends (LABELIT and DIALS)

The cctbx.xfel backend was originally implemented using 1D Fourier indexing algorithms (Steller et al. 1997), as made available in LABELIT (Sauter et al. 2004). This LABELIT backend was expanded in the cctbx framework to include stills-specific algorithms, such as additional targets for refining crystal orientation and refining mosaic estimates needed to determine which reflections are in the diffracting condition (Sauter et al. 2013, Hattne et al. 2014, Sauter et al. 2014, Sauter 2015). These procedures have been implemented and expanded in DIALS, taking advantage of the diffraction models and refinement engine made available in that platform (Waterman et al. 2016).

For example, the DIALS indexer provides three algorithms for determining initial sets of basis vectors during indexing: fft3d, fft1d and real space grid search. fft1d is an implementation of the 1D FFT algorithms described above, with the addition that the user can markedly increase the success rate of indexing by providing a target unit cell and space group based on prior knowledge (Hattne et al. 2014). fft3d is an implementation of 3D FFT methods (Bricogne 1986, Campbell 1998) and isn’t relevant for stills. The real space grid search approach was described recently as a simplification of Fourier methods when unit cell dimensions are already available (Gildea et al. 2014). In the case of unknown unit cell parameters, real space grid search is not available and fft1d remains the best choice for stills. However, after an initial indexing test, the unit cells determined by fft1d from many still diffraction patterns can be hierarchically clustered (Zeldin et al. 2015) and a consensus unit cell for the sample can be determined. There are choices of lattice distance functions to be used in clustering. The most effective is the \( G^6 \) space distance function (Andrews & Bernstein 2014, McGill et al. 2014). Then, the consensus cell can be used as a target for indexing using real space grid search. Notably, in a recent experiment with lysozyme crystals, we found that real space grid search with a well-determined set of unit cell parameters can find up to 21% more lattices than fft1d. However, a different experiment found that fft1d gave dramatically more results than real_space_grid_search. We encourage users to try both options to determine which yields...
better results according to a chosen figure of merit and we invite users to contact the authors to share their experiences.

DIALS also provides a mechanism for parameterizing experimental models that lends itself naturally to building complex refinement target functions (Waterman et al. 2016). The complete experiment is described through a series of models: the crystal (unit cell and orientation), the beam, the detector, the goniometer axis and orientation and the scan oscillation range and increment. While the goniometer and scan models are not applicable for stills, the crystal, beam and detector models can be refined against measured data according to stills-specific targets (Sauter et al. 2014). Importantly, individual parameters such as the wavelength or the detector distance, tilt or orientation can easily be fixed, i.e. locked into place, depending on the use case.

The DIALS backend for cctbx.xfel includes a derivation of the DIALS indexer optimized for stills and includes all of the stills-specific algorithms mentioned above, taking advantage of the open-source and object-oriented nature of the cctbx framework for which DIALS, LABELIT and cctbx.xfel are all derived.

1.4.2. XTC and CSPAD CBF formats

The LCLS data acquisition systems stream terabytes of data to container files in XTC format. XTC is a linear, sequential-access file format, where individual events can be recorded rapidly to the file system as they are collected. ‘Derived’ metadata such as detector position and percent beam attenuation are not provided directly; instead, motor positions and the status of LCLS instrument parameters are recorded. It is up to end user software to transform this information into the parameters needed to describe the crystallographic experiment. psana abstracts many of these transformations. For instance, cctbx.xfel interfaces with psana to couple the raw pixel data from the 64 ASICs in each CSPAD detector event in the XTC stream with transformed metadata to create files in Crystallographic Binary Format (CBF) that contain the pixel data and also completely describe the experiment in their headers, using standards established by the International Union of Crystallography (IUCr) (Bernstein & Hammersley 2006). This format has been described in detail previously (Brewster et al. 2014). Briefly, the geometry of the CSPAD detector is recorded as a series of basis transformations that move an observer from the sample interaction point (the crystal) to the detector, then to each of 4 quadrants, then to each of 32 sensors, then to each of 64 ASICs. All transformations are relative to a parent frame, which allows the positions of groups of objects (such as sensors in a quadrant) to be refined as a unit by only refining the vectors defining the parent object’s frame of reference.

1.4.3. PHIL format

Python Hierarchical Interface Language (PHIL) is the syntax for specifying parameters in cctbx (Grosse-Kunstleve et al. 2005, Grosse-Kunstleve et al. 2006, Bourhis et al. 2007). Phenix users will know it as .eff format (effective parameter file). This short example is used to configure the DIALS spotfinder (details explained below):

```python
spotfinder {
    filter.min_spot_size=2
    threshold.xds.gain=25
    threshold.xds.global_threshold=100
}
```
PHIL format uses curly braces to establish parameter scopes among programs and uses name-value pairs to specify parameters. Here, spotfinding parameters such as minimum spot size, gain estimates and global background thresholds are provided inside the spotfinder scope.

1.4.4. Intermediate DIALS formats
DIALS utilizes two intermediate file formats to store experimental models and reflection information prior to merging and scaling data and writing MTZ format files for subsequent structure solution and refinement.

1.4.4.1. JSON format
DIALS represents crystallographic experiments as a series of physical models, including detector, beam, goniometer and scan, through the Diffraction Experiment Toolbox library (dxtbx) included in cctbx (Parkhurst et al. 2014). For stills, acquired by serial femtosecond crystallography, no goniometer or scan objects are used. For each component in the hierarchical detector, the model includes the positional vectors $d_0$, $d_f$ and $d_s$. $d_0$ points from the parent component’s origin to the child’s origin, while $d_f$ and $d_s$ define orthogonal fast and slow vectors that, when combined with the normal vector $(d_f \times d_s)$, specify a basis frame for the component. The ‘leaves’ of the detector model, e.g. the ASICS for the CSPAD, also contain information to convert from pixel to millimeter coordinates, such as pixel size, taking into account a parallax correction by including ASIC thickness and material composition. The beam model includes the beam vector ($s_0$) and the wavelength of incident photons. All of these metadata are serialized using JavaScript Object Notation (JSON) in .json files.

The JSON files are organized thusly (here, each indentation level represents a JSON entry):

```
ExperimentList
Experiment1
  BeamID
  DetectorID
  CrystalID
Experiment2
  ...
Beams
  Beam1
    <beam properties i.e. wavelength, direction>
  Beam2
    ...
Detectors
  Detector1
    <detector properties i.e. $d_0$, $d_f$, $d_s$ vectors>
  Detector2
    ...
Crystals
  Crystal1
    <crystal properties i.e. unit cell, orientation>
  Crystal2
    ...
```
At the top, an experiment list defines a set of experiments each containing a beam, detector and crystal model, identified using numerical IDs. Then, the set of detector, crystal and beam models referenced by the experiments are listed. An individual experiment always contains exactly one detector, beam and crystal model. A single detector model can be shared by every experiment, for example if there were multiple crystals in a single shot. Alternatively, each experiment could reference a different detector model, perhaps taking account shot-to-shot variability in detector position due to a fluctuating injection system or from other jitter. Likewise, two experiments could each share a crystal model, perhaps having different beam models in the case of a two-color experiment, or perhaps having different detector models from two different detectors. This organization provides a flexible means of organizing a variety of possible experiment types and has already been used to jointly refine multiple lattices simultaneously from multiple crystals exposed during a single rotation series (Gildea et al. 2014, Waterman et al. 2016).

1.4.4.2. Reflection table pickles
Spots found by spotfinding, indexed by indexing, or integrated during integration are recorded in reflection tables, where each spot is one row and the columns contain data such as Miller index, observed or predicted location relative to the panel origin in mm, summed intensity and variance, etc. These tables are serialized into python pickle files.

Both JSON experiment files and reflection table pickle files can be inspected with the command:

```
dials.show filename
```

Note that commands and parameters presented here will use the above formatting. Program names are shown in **bold**. Generally, additional documentation is available for each program listed with -h, --help, or -c (for configuration). Use -e to specify ‘expert level’ if desired (10 shows all parameters, or use 0, 1 or 2 for fewer parameters). Use -a 2 to show documentation for each parameter.

2. **cctbx.xfel** operational overview at LCLS

It has been our experience that analyzing data collected using serial crystallography (SX) typically requires three distinct processing stages labeled here **calibration**, **discovery** and **batch** (figure 1). **Calibration** refers to refining the geometry of the experiment, but also includes some pre-processing steps, such as creating dark and light averages, bad pixel masks and gain masks. Using these inputs, initial parameters are derived that describe the experiment, such as detector distance, quadrant and sensor layout, any beam correction parameters needed and so forth. During **discovery**, the user examines individual diffraction patterns and searches for appropriate parameters for data reduction, including hitfinding parameters if used, spotfinding parameters, target unit cell dimensions and crystal symmetry and an optimal merging strategy. Finally, when optimal software configuration is established, the user enters **batch** processing mode, endeavoring to maximize the parallel computing options offered and, during live experiments, attempting to provide constructive feedback to beam line operators in as close to real-time as possible. After the experiment, the user will often need to reprocess the runs collected in batch mode. During batch processing, the user will continue to refine processing parameters as the results are evaluated, perhaps even revising initial metrology estimates. Thus the three stages are somewhat fluid as feedback from later stages may call for repeating earlier stages.
LCLS organizes its data according to end-station name, experiment name and run number, under the global directory /reg/d/psdm. Thus, if a user was assigned experiment ID cxi84914, their data would be at /reg/d/psdm/cxi/cxi84914. The XTC streams will be in an xtc directory at that location, the calibration store, including for example geometry files and pedestals, will be in a directory named calib and the user will be able to store processing results temporarily in a scratch directory at that location, or more permanently in a designated results directory named res. The magnitude of the data being collected — over 100 terabytes from a five-day experiment is not unusual — leads to a paradigm where the user never takes their raw data home. Instead, they reduce their data to integrated, merged and scaled intensities in MTZ format using the LCLS clustering system, then transfer that much smaller file to their home computer for downstream analysis.

3. Averaging CSPAD data

An important part of discovery and calibration involves averaging images together to get a sense of the behavior of the data. cxi.mpi_average produces three images from a run: an average image, a standard deviation image and a composite maximum image. In these images, every pixel is the average, standard deviation, or maximum of all pixels in that register over an entire run. Averages from a dark run, where the detector
is not exposed to X-rays, or from a light run, where the detector is exposed to X-rays, possibly in the presence of sample, are useful for several reasons. Light and dark images can be used to determine which pixels are not trusted. Inactive and non-bonded pixels, as well as hypersensitive pixels and any pixels shadowed by a beam stop should be masked out. The light images and the composite maximum in particular, also serve as a virtual, dark-subtracted powder pattern that can be used to inspect the data for diffraction. The light maximum should also flag obvious errors in metrology, because rings will be non-continuous if the quadrants are misaligned. Intensity values from the corners of the light average or maximum are also useful to estimate background when determining initial estimates of thresholds for spot-finding and integration (see below).

3.1. Example averaging commands

Use this command to average the CSPAD data:

```
cxi.mpi_average -x cxid9114 -r 95 -a CxiDs2.0:Cspad.0 -d 572 -v -g 6.85
```

The program `bsub` is used to submit jobs to the LCLS queuing system. The parameters are:

- `-n`: number of processors
- `-q`: queue name
- `-o`: log file name

`mpirun` is used to enable inter-process communication so the averaging program can dispatch images to different computing cores and gather the results when complete. For more information, please see documentation from LCLS:

https://confluence.slac.stanford.edu/display/PCDS/Submitting+Batch+Jobs

3.2. Using averages to create an untrusted pixel mask

`cctbx.xfel` uses three images created during averaging to create a mask for the CSPAD detector. From the average image of a dark run, pixels with intensities ≤ 0 are considered dead, while intensities > 2000 are flagged as hypersensitive. From the standard deviation of a dark run, pixels with intensities ≤ 0 are considered dead and intensities ≥ 10 are too uncertain and noisy. From the composite maximum from a lighted run (i.e. an experimental run), pixels with intensities < 300 are considered non-bonded or in shadow. The presence of diffraction is not needed for the third image, but also will not interfere. The default of 300 here is unique among the numbers listed in that it will likely vary with the sample's background while the other defaults do not usually need to be changed. To tune this value, the user can carefully examine the corners of the lighted maximum projection and choose a value lower than the ADU values displayed. Also note that there is no cutoff on the high end of the maximum.

```bash
for i in `seq 95 114`; do bsub -n 12 -q psanaq -o avg_r$i.log mpirun cxi.mpi_average -x cxid9114 -r $i -a CxiDs2.0:Cspad.0 -d 572 -v -g 6.85; done
```

The gain ratio is a constant multiplier for all low-gain pixels to be scaled to be at the same level as the high-gain pixels. When operating the CSPAD in mixed-gain mode, this ratio is needed to apply this correction. The value of 6.85 used here is not exact, but it is sufficient for this purpose.

The user may find it useful to average all the available data using the queuing system and to use multiple cores to reduce processing time. This can be done using a single command:
projection image specified in the mask, as that is defined as the saturation value for the detector.

Here is an example command. Note that we have chosen 50 instead of 300 for the parameter maxproj-min after examining the data:

```
cxi.make_dials_mask --maxproj--
  min=50 -o mask.pickle
  cxid9114_avg-r0089.cbf
  cxid9114_stddev-r0089.cbf
  cxid9114_max-r0096.cbf
```

4. Calibrating the CSPAD detector

As described above, the CSPAD consists of 32 2x1 sensors whose position in space must be refined. Before the user arrives at the XPP or CXI end-station, the beam line operator will have used an optical microscope to measure within each quadrant where the 2x1 sensors are located relative to one another and will have deployed these measurements as a starting metrology. While each of the 8 sensors within a quadrant will be well positioned relative to each other, the overall position of each quadrant relevant to the beam is usually not well determined. Initial quadrant positioning can be done using virtual powder rings from the average of many individual crystals, as described below. Average images with strong powder rings allow the user to determine the relative placement of each quadrant by aligning them such that the rings are contiguous. With these positions, initial indexing of a subset of data can produce measured and predicted Bragg reflection positions from which tiles are refined, minimizing the difference between observed and predicted spot locations from indexed diffraction data. The user re-indexes the data and re-refines the detector metrology until convergence is reached. It is recommended that high resolution, highly reproducible diffraction data from a reference set such as lysozyme or thermolysin be collected for calibration and that the detector is positioned such that diffraction reaches its corners.

4.1. Manual quadrant calibration

Typically the best powder rings come from the composite maximum (example: the file cxid9114_max-r0113.cbf will have been generated from the above averaging command). To manually align the quadrant positions, either use calibman (see LCLS documentation) or use cctbx.image_viewer with the composite maximum. Under actions, click on 'Show quadrant calibration' and then use the spinners to align the powder rings. One may find the ring tool or the unit cell tool, also under the Actions menu, to be useful visual aids during this process. When done, click 'Save current metrology' to save the changes to a .def file, which is a CBF header.

4.2. Automatic quadrant calibration using cctbx.xfel

If a quadrant is properly positioned relative to the beam center, the pixel values for a strong powder pattern will be highly correlated after rotating the quadrant 45 degrees around the beam center. cspad.quadrants_cbf performs a grid search of XY offsets for each quadrant,
searching for the position with the highest rotational autocorrelation. It then writes out a new CBF file with the adjusted header:

```
cspad.quadrants_cbf cxid9114_max-r0105.cbf
```

Specify the '-p' parameter to enable plots of the grid search results for each quadrant, in addition to reporting correlation coefficients (CCs) for each quadrant. The aligned image can be inspected with the image viewer:

```
cctbx.image_viewer cxid9114_max-r0105_cc.cbf
```

It is possible that a maximum of all the runs would have more contiguous and brighter rings, leading to higher CC values. This can be done quickly using the previously generated per-run composite maxima:

```
cxi.cspad_average *_max*.cbf -m all_max.cbf
```

Once a satisfactory maximum has been obtained, the quadrants tool can be called:

```
cspad.quadrants_cbf all_max.cbf
```

If the CC values are higher than when using individual run maxima, then this is a better approach for finding a good set of quadrant positions prior to initial indexing attempts.

### 4.3. Deploying new quadrant positions

Before the new quadrant positions can be used for indexing, the new layout needs to be converted to SLAC’s metrology file format. Use this command if the quadrants were aligned manually using `cctbx.image_viewer`:

```
cxi.cbfheader2slaccalib cbf_header=quadrants.def
```

Or this command if the quadrants were aligned using `cspad.quadrants_cbf`:

```
cxi.cbfheader2slaccalib cbf_header=all_max_cc.cbf
```

The resultant file, 0-end.data, needs to be copied to the calibration store for the experiment. Typically, this will be in a directory in this form:

```
/reg/d/psdm/cxi/cxid9114/calib/CsPad::CalibV1/CxiDs2.0:Cspad.0/geometry
```

There will already be a 0-end.data file in this folder. We recommend renaming this file to 0-end.data.v0, copying the new 0-end.data to this folder under the name 0-end.data.v1, then soft-linking 0-end.data.v1 to 0-end.data. This will maintain a version history, as metrology is refined for this experiment.

### 4.4. Metrology Versioning

The user may find it useful to keep track of the improvement in metrology estimates using a versioning system. In this article, we version the metrology using the conventions in table 1.
### Table 1: Metrology versioning

<table>
<thead>
<tr>
<th>Metrology version</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Version 0 (v0)</td>
<td>Initial metrology deployed by beam line operators. The tile positions are measured using an optical microscope, but, as the quadrants can move independently, they are not correctly aligned in relation to each other or to the beam center.</td>
</tr>
<tr>
<td>Version 1 (v1)</td>
<td>After collecting some data, virtual powder rings can be seen after averaging the events in a run. Quadrants are aligned by eye or automatically using <code>cspad.quadrants_cbf</code>.</td>
</tr>
<tr>
<td>Version 2 (v2)</td>
<td>After indexing the images using v1, the tile positions are refined to produce metrology v2.</td>
</tr>
<tr>
<td>Version 3 (v3)</td>
<td>After re-indexing the images using v2, the tile positions are re-refined to produce metrology v3.</td>
</tr>
<tr>
<td>...</td>
<td>And so forth until convergence.</td>
</tr>
</tbody>
</table>

### 4.5. Detector distance and detz\_offset

Each individual event in the XTC stream includes motor position settings for the detector (figure 2). The detector’s position is measured from the back of the rail on which the CSPAD detectors at CXI moves to the detector itself. Call this distance detz. Naturally, the desired distance needed for crystallographic analysis is from the detector’s current position to the sample interaction region. For this reason, during processing it is necessary to supply an offset (detz\_offset) from the sample interaction region to the back of the detector rail, a value that is constant over the course of a given experiment but changes between experiments when new injectors are substituted in and out of the sample chamber, a common occurrence.

![Figure 2: Schematic of the detz\_offset parameter.](image)

The CSPAD at the LCLS CXI endstation can be translated on a detector rail shown in black. The position of the CSPAD along this rail, detz, is determined from motor positions and recorded in the XTC stream for each event. As the sample injection system varies between users, the distance between the crystal and the back of the detector rail, the detz\_offset, needs to be measured or experimentally refined. Then, the detector distance d can be determined for each event from the difference between the detz\_offset and the detz parameter.
cctbx.xfel will compute the actual detector distance for each frame using the difference between detz_offset and detz. Detz_offset is available from the beam line operators but needs to be refined. See (Hattne et al. 2014), where detz_offset is refined by screening a range of values and finding the detz_offset that indexes the greatest number of images and see (Nass et al. 2016), where the detector distance is optimized by minimizing the standard deviation of the distribution of unit cell dimensions from P1 indexing trials at different distances. Importantly, eliminating multi-modal unit cell dimension distributions by using slightly different detector distances on a day-by-day or even run-by-run basis can increase the final quality of the data.

4.6. Initial indexing
During indexing, the user parameterizes cctbx.xfel with settings regarding spotfinding, indexing, refinement and integration. An initial set of parameters needs to be established before indexing can be reliable. These parameters will be recorded in a PHIL file.

4.7. Spotfinding
The most important parameter for the DIALS spotfinding algorithm is the gain, meaning the number of analog-to-digital units (ADU) per incident photons on a pixel, recorded on that pixel. The program cctbx.xfel.xtc_dump can be used to dump CSPAD CBFs from an XTC stream to determine the flat gain value from many images at once. Here we use run 113, as its average image revealed the presence of strong diffraction:

```
cctbx.xfel.xtc_dump
   dispatch.max_events=100
   input_experiment=cxid9114
   input.address=CxiDs2.0:Cspad.0
   input.run_num=113
   format.file_format=cbf
   format.cbf.Detz_offset=572
   input.override_energy=8950
```

Then the program dials.estimate_gain can be run on these files which will estimate a flat gain for each image gain based on the statistical distribution of reordered pixel values:

```
dials.estimate_gain
   cspad_image.cbf
```

The final set of parameters for spotfinding is:

```
spotfinder {
   filter.min_spot_size=2
   threshold.xds.gain=25
   threshold.xds.global_threshold=100
}
```

The min_spot_size parameter specifies the minimum number of high-intensity pixels that must be present to classify the pixels as belonging to a spot. A value of 2 is appropriate for the CSPAD, which can record very small reflections. The global_threshold parameter should not generally be needed for PAD data, but, here in mixed high/low gain mode, the flat gain estimate of 25 is not reliable for the entire detector. Because of this, we arbitrarily state that all pixels less than 100 ADU are background for spotfinding (but not for integrating).

The program dials.image_viewer includes a mechanism for displaying which pixels will be included as signal for a given set of spotfinding parameters. This is a highly useful tool for estimating these parameters.

4.8. Indexing
Initial parameters for the DIALS indexer include a target unit cell, an indexing method and a resolution cutoff:

```
indexing {
   known_symmetry {
      space_group = P43212
      unit_cell = 78.9 78.9 38.1 90 90 90
   }
   method=real_space_grid_search
   refinement_protocol.d_min_start=1.7
}
```
As lysozyme is well known, it is straightforward to assign known symmetry. In unknown cases, this can be left blank and the fft1d method can be used in place of real_space_grid_search. The resolution cutoff here is chosen to include the whole detector, but this may not be appropriate if the unit cell is not well known or if the resolution shown in the average composite is lower.

4.9. Refinement

The DIALS refiner is used to optimize the experimental parameters after indexing. Refinement is performed after each indexing solution is determined for each diffraction pattern. Here, we change some settings in the parameterization block for the refiner:

```plaintext
refinement {
  parameterisation {
    beam.fix=all
    detector.fix_list=Dist,Tau1
    auto_reduction {
      action=fix
      min_nref_per_parameter=1
    }
  }
  crystal {
    unit_cell {
      restraints {
        tie_to_target {
          values=78.9,78.9,38.1,90,90,90
          sigmas=1,1,1,0,0,0
        }
      }
    }
  }
}
```

As these data were collected using seeded pulses instead of self-amplified stimulated emission (SASE) XFEL pulses, should have a constant energy, so we fix the beam parameters in place such that they are not refined. We also fix the detector distance (dist) and the rotation of the detector around the z axis. For SASE data, the user can consider allowing the detector distance to refine for each image by setting detector.fix_list equal to Tau1 only. Regardless, either the beam model or the detector distance (if not both) should be fixed, as they are co-dependent.

When the refiner determines that there are too few observations to reliably refine a model, its behavior is determined by the auto_reduction parameters. Some models, such as the beam model, have few parameters, while others, such as the detector or crystal models, have many. If the number of reflections needed per parameter is set to zero, all parameters for all models will be refined regardless of how many reflections are available. Otherwise, if too few reflections are available to refine a given model, then one of three actions is taken: 1) Fail: refinement does not proceed and processing stops for this image. 2) Fix: the parameters for this model are fixed in place and not refined, but the reflections associated with it will still be used for other models if possible. 3) Remove: the model and all reflections associated with it will be removed from refinement; when refining a single still shot, this effectively means refinement will not occur for any of the models.

For stills, we find there are often not enough reflections to refine all the parameters (cell orientation, detector position, etc.) using the DIALS default minimum value of 5 reflections per parameter, because the default is optimized for a rotation experiment using a goniometer in
which there are many more reflections than there are in a single still shot. Here we set the minimum to 1, but 3 may also be a reasonable alternative, especially if the unit cell is larger. Poorly determined parameters will increase the root mean square deviation (RMSD) of the differences between observed and predicted reflections, which would affect the quality of the refined tile positions during metrology refinement later. However, during tile refinement, an image will be rejected if its overall positional RMSD is too high, so a minimum reflection count per parameter of 1 may be sufficient.

Finally, we restrain the unit cell parameters. The tie_to_target option uses a known set of cell dimensions and a set of weights for each dimension specified using the sigma parameter. Here, a sigma of 1 for the cell dimensions allows for some variation during refinement and is used for the edge lengths. For this orthorhombic space group we remove the restraint for the angles by setting those sigmas to zero.

4.10. Integration

The DIALS integrator uses the stills-specific mosaic parameters estimated during integrating to predict which reflections are in the diffracting condition (Sauter et al. 2014). Additional parameters are shown here:

```
integration {
    profile.fitting=False

    background {
        algorithm = simple
        simple {
            model.algorithm = linear2d
            outlier.algorithm = plane
        }
    }

    profile {
        gaussian_rs {
            min_spots.overall = 0
        }
    }
}
```

We disable profile fitting for stills. For the background, the DIALS default algorithm, glm, is optimized for counting detectors such as the Pilatus series from Dectris, which have a very low background. We have found this is not appropriate for the CSPAD, so we choose a “simple” linear 2d algorithm which estimates a gradient from nearby pixels for each reflection after fitting the background to a plane (Leslie 1999).

Lastly, even though we disable profile modeling, we instruct the profile modeler to accept the experiment even if the number of reflections is low (which is typically true for stills).

4.11. XTC and CSPAD specific parameters

cctbx.xfel.xtc_process is the program used to read the XTC streams, create CBF images in memory and then invoke the DIALS procedures. It needs to be parameterized in ways specific to the LCLS experiment:

```
input {
    address=CxiDs2.0:Cspad.0
}
format {
    file_format=cbf
    cbf {
        detz_offset=572.3938
        invalid_pixel_mask=mask.pickle
        gain_mask_value=6.85
        override_energy=8950
        common_mode.algorithm=custom
        common_mode.custom_parameterization=5,50
    }
    border_mask {
        border=1
    }
}
```

The address string identifies which CSPAD should be used to read data from in the XTC streams and can be obtained from the beam line operator. The format section specifies experiment-specific parameters to be written into the CBF file headers or to be used to correct the pixel data before adding it to the CBF main body. Here is where we specify the untrusted pixel mask created previously. The detz_offset parameter was chosen by indexing the data while letting the z axis refine, then creating a histogram of the distance values found for each image. Generally, however, the beamine operator’s estimate is a good initial
point to index enough images. A test of several different detz_offset values is another approach (Hattne et al. 2014).

For these seeded pulses the energy is known, so we override the energy value found for each shot in the XTC stream. For SASE data, the XTC stream contains an estimate of the overall energy for the pulse and should be used, meaning that the user should leave this field blank. Finally, LCLS provides a variety of common mode correction algorithms (see https://confluence.slac.stanford.edu/display/PSDM/Common+mode+correction+algorithm s). The common mode is a per shot, per sensor small offset on the order of 10 ADU that occurs due to changes in potentials during readout of the sensors. Determining it for protein crystallography, which generally has a high background due to solvent scattering, is difficult. If common_mode.algorithm is left unspecified in the PHIL file, no correction will be applied. The user may otherwise specify ‘default’ for this option, in which case the current default LCLS corrections will be applied. Currently this is algorithm 1, a pixel histogramming method applicable to weak signal and likely not applicable to protein crystallography with strong solvent scattering. The user can also specify ‘custom’ and pick an algorithm from the algorithms described at the above link. Algorithm 5, selected in this PHIL file, uses non-bonded pixels, i.e. pixels not bump bonded to the electronics in a given ASIC and specifies a maximum correction applied to each pixel of 50 ADU. Currently, we can offer no advice as to which is best as it is a matter of active research. Generally, we have been processing with no correction, leaving the common_mode parameters blank.

Finally, we specify a border mask of 1 pixel for each of the 64 tiles, because the wider edge pixels of the CSPAD are not on the same ADU scale as the rest of the pixels in the ASIC.

4.12. Final PHIL file for initial indexing

Putting it all together, the initial PHIL file for indexing will look like this:

```plaintext
input {
  address=CxiDs2.0:Cspad.0
}

format {
  file_format=cbf
cbf {
    detz_offset=572.3938
    invalid_pixel_mask=mask.pickle
    gain_mask_value=6.85
    override_energy=8950
    common_mode.algorithm=custom
    common_mode.custom_parameterization=5,50
  }
}

border_mask {
  border=1
}

spotfinder {
  filter.min_spot_size=2
  threshold.xds.gain=25
  threshold.xds.global_threshold=100
}
```
indexing {
    known_symmetry {
        space_group = P43212
        unit_cell = 78.9 78.9 38.1 90 90 90
    }
    method=real_space_grid_search
    refinement_protocol.d_min_start=1.7
}

refinement {
    parameterisation {
        beam.fix=all
        detector.fix_list=Dist,Tau1
        auto_reduction {
            action=fix
            min_nref_per_parameter=1
        }
    }
    crystal {
        unit_cell {
            restraints {
                tie_to_target {
                    values=78.9,78.9,38.1,90,90,90
                    sigmas=1,1,1,0,0,0
                    apply_to_all=True
                }
            }
        }
    }
}

integration {
    integrator=stills
    profile.fitting=False
    background {
        algorithm = simple
        simple {
            model.algorithm = linear2d
            outlier.algorithm = plane
        }
    }
    profile {
        gaussian_rs {
            min_spots.overall = 0
        }
    }
}
4.13. Indexing commands

Indexing in `cctbx.xfel` is typically done in a series of trials. Our first trial will be trial 0, using metrology v1 (initial metrology from beam line operators, with quadrants corrected using one of the above techniques). The program `cxi.mpi_submit` will submit the indexing command to the queuing system. LCLS’s LSF queue is supported and SGE and custom queuing commands are also available. Please contact the authors for advice on running the software on any queuing systems not yet supported.

We generally organize our work into numbered ‘trials’, where each trial represents an experimental set of parameters. The program `cxi.mpi_submit` creates a directory for the trial, copies all config and PHIL files as backups and submits the processing job to the LCLS cluster. For example, to do the initial indexing trial for this data, we use the command:

```
for i in `seq 95 114`; do
    cxi.mpi_submit
    input.experiment=cxid9114
    output.output_dir=/reg/d/psdm/cxi/cxid9114/ftc/brewster/dalis/mp.nproc=36 mp.queue=psanaq
    output.split_logs=True
    input.dispatcher=cctbx.xfel.xtc_process input.target=LD91-lyso-t000.phil input.trial=0
    input.run_num=$i
    dispatch.integrate=False; done
```

This for loop in bash submits runs 95 through 114, inclusive, for processing. The experiment name is specified to allow `psana` to find the XTC streams with the data. In the output directory, a run directory for each run will be created and under that a three-digit trial directory (named 000 for trial 0) will be created. Logs will be saved to a stdout directory in the trial directory. Here, we use split logs, which means that in addition to the main log file, each of the 36 processors requested here will write to a separate log file so as not to interleave all the output from all the processors. The queue to use is specified as `psanaq`, the public queue available at LCLS. Other queues are available, as described here: https://confluence.slac.stanford.edu/display/PCDS/Submitting+Batch+Jobs. The dispatcher refers to which XTC processing program to run; `cctbx.xfel.xtc_process` invokes the DIALS processing pipeline. The other program available, `cxi.xtc_process`, invokes the LABELIT backend as described elsewhere (Hattne et al. 2014). The PHIL file for processing is specified as a target. Finally, to save time during initial indexing and metrology refinement (the discovery phase), we use `dispatch.integrate=False` to skip the integration step.

After the job completes, indexing results will be available in the results folder for each run, under trial 000 in a folder named out. For each indexed image a CBF file will be created. In addition, the files refined_experiments.json and indexed.pickle will be created, containing information about the indexing solution and the list of indexed reflections, respectively. `dials.show` is useful for displaying some summary information about the contents of these files and `dials.image_viewer` can be invoked with a CBF file and an indexed.pickle file to visualize the indexed reflection positions overlaid on the image data.

`cctbx.xfel.xtc_process` provides user control over which steps in the spotfinding, indexing and integration are executed. For example, the user could dump all images with strong reflections to CBF whether or not they indexed using `dispatch.dump_strong=True`. Use the `-c` (configuration) parameter to show the full set of options available.

4.14. Refinement of tile positions

After initial indexing results are obtained, the program `cspad.cbf_metrology` is used to refine the tile positions of the CSPAD detector. This program aggregates many individual indexing results to do a joint refinement of many crystal models and a single detector.
model using the differences between the observed spot locations and predicted spot locations as a target function. In brief, the steps taken are:

Use `dials.combine_experiments` to concatenate a number of indexing results into a single combined_experiments.json file and a single combined_reflections.pickle file. During this step, each crystal model from each experiment is retained separately, while the detector positions are averaged together to create a single detector model.

- Filter the set of indexing solutions, rejecting images with an RMSD high enough to be considered an outlier using Tukey’s rule of thumb.
- Refine the detector as a whole, including rotation and tilt, using `dials.refine`.
- Filter individual images again by RMSD.
- Refine each quadrant separately using `dials.refine`.
- Filter individual images again by RMSD.
- Refine each 2x1 sensor separately using `dials.refine`.
- Convert the final refined_experiments.json file into a CBF header.
- Convert the CBF header to SLAC format, ready to deploy as a new metrology file.

Use a command like this one to refine the metrology, assuming the results folder is in a directory at the same level as the current folder:

```
cspad.cbf_metrology tag=t000_2k 
../results/r0*/000/out
n_subset=2000
split_dataset=True cxd9114-refine.phil
```

The tag is prepended to every output file name. `n_subset` selects a random group of images for refinement from the input set of folders. `split_dataset` instructs the program to perform the refinement twice with a different random subset of 2000 each time. This is useful to verify the precision of the refinement, meaning how likely it is for the refinement to produce the same detector metrology given a different set of indexing results. The PHIL file provided parameterizes the `dials.refine` steps. Here we only specify a light restraint to the target unit cell as described above:

```json
refinement {
  parameterisation {
    crystal {
      unit_cell {
        restraints {
          tie_to_target {
            values=78.9,78.9,38.1,90,90,90
            sigmas=1,1,1,0,0,0
            apply_to_all=True
          }
        } 
      }
    }
  }
}
```

Depending on the quality of the calibration dataset, the user may decide to forgo filtering individual images with high RMSD out of the dataset. Use `rmsd_filter.enable=False` to do this. An additional parameter can specify which set of reflections to use during metrology refinement. `reflections=indexed` is the default and specifies using only indexed, strong reflections. Refinement can also be done against integrated reflections. During indexing, `dispatch.reindex_strong` can be used to re-index the bright reflections found during spotfinding that weren’t matched during indexing. Then, during metrology refinement, `reflections=reindexed_strong` can be used to include more strong reflections than just the ones from initial indexing. This provides more data to use during refinement without adding weak reflections from integration, which may have poorly determined centroids.

Finally, the user can use two different methods of selecting the subset of data to use for refinement. The default method picks randomly from the available images to create a dataset with which to refine metrology. Alternatively, the user can specify `n_subset_method=n_refl` to instead pick the subset with the greatest number of reflections per image to increase the likelihood of measuring diffraction to the corners of the detector.
The final output file, 0-end.data.t000_2k_1 in this example, is ready to use as a new version of the metrology (here, version 3, or v3). Deploy it to the calibration folder as described above.

4.15. Evaluating metrology convergence
After deploying the metrology, it is recommended that the user repeat indexing and metrology refinement until tile positions converge, meaning until the tile positions are no longer changing significantly and the quality of the positional RMSDs is no longer improving. Two programs are available to assist in evaluating this: cspad.detector_shifts and cspad.detector_statistics.

**cspad.detector_shifts** can be used to see changes in panel positions during refinement:

```
cspad.detector_shifts
t000_2k_1_filtered_experiments.json
t000_2k_1_filtered_reflections.pickle
t000_2k_1_filtered_experiments_level2.json
t000_2k_1_filtered_reflections_level2.pickle
```

This test compares the unrefined metrology (first two files) with the results after sensor refinement (level 2). The CSPAD has four levels of metrology available to refine, level 0 (detector as a whole), level 1 (4 quadrants), level 2 (32 2x1 sensors) and level 3 (64 individual ASICs). As the ASICs are physically connected, level 3 need not be refined. Some example output could look like this:

<table>
<thead>
<tr>
<th>Hierarchy</th>
<th>Delta XY (microns)</th>
<th>Delta XY Sigma (microns)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>6.2</td>
<td>0.0</td>
</tr>
<tr>
<td>1</td>
<td>78.1</td>
<td>53.2</td>
</tr>
<tr>
<td>2</td>
<td>14.5</td>
<td>11.4</td>
</tr>
<tr>
<td>3</td>
<td>0.0</td>
<td>0.0</td>
</tr>
</tbody>
</table>

This output reveals that the detector as a whole (level 0) moved 6 microns during refinement in the XY plane (orthogonal to the beam). Each quadrant moved 78+/−53 microns and each sensor moved 14.5 ± 11.4 microns. As the quadrants are the least well determined, it is reasonable their positions should move the most. The full output contains more information, such as z offsets and rotations.

Use **cspad.detector_statistics** to evaluate the precision of refinement. This program compares the two independent refinements performed by **cspad.cbf_metrology** when split_dataset=True.

```
cspad.detector_statistics
tag=t000_2k
```

This program produces a lot of output, comparing tile positions at each level of the detector hierarchy. We will focus on a few of these statistics.

Each refinement job, two in this case, produces one measurement of the position of any given CSPAD tile. The F offset sigma and S offset sigma values report weighted standard deviations of the two measurements of the each sensor’s fast (F) or slow (S) coordinate. These values are reported for each 2x1 sensor and after several rounds of indexing and refinement using monochromatic data from this experiment, the overall weighted averages of the fast and slow offset sigmas were 1.6 and 1.7 µm, respectively, indicating the overall precision of this refinement was around 1 µm. This is much smaller than the pixel size of the CSPAD, 110 µm.

Also reported is the observational RMSD of the differences between observed and predicted reflection positions, which for this experiment was 38.3 µm for the first split dataset and 37.5 µm for the second. In all cases, weighted averages or standard deviations are computed using the number of reflections observed as the weight.
5. Batch integration, scaling and merging of XFEL data

After metrology refinement is complete, the user can process data in batch mode using `cxi.mpi_submit` as described above, using `dispatch.integrate=True`. During integration, in addition to `experiment.json` and `integrated.pickle` files in DIALS format, the integration results are also stored in `cctbx.xfel` integration files using a convention of `int-0-<timestamp>.pickle`. The files are the direct inputs to scaling and merging programs available through `cctbx.xfel`, namely `cxi.merge` (Sauter 2015) and `prime.postrefine` (Uervirojnangkoorn et al. 2015). Directions for the use of these programs are provided on the `cctbx.xfel` wiki (http://cci.lbl.gov/xfel).

6. Processing serial crystallographic data from other sources

The main program used for processing data presented here, `cctbx.xfel.xtc_process`, is an interface between the LCLS system `psana` used for writing and reading XTC streams, the libraries in `cctbx.xfel` needed for creating and using CSPAD CBF files and the stills-specific algorithms implemented in DIALS for indexing, refinement and integration. Users collecting serial crystallographic data at synchrotron sources on new, high performance detectors such as the Eiger by Dectris have the option of using these same algorithms for still shots with the program `dials.stills_process`. This program accepts a PHIL file as described above, with the exception that the format section is omitted, as the file headers are used directly to assemble the appropriate detector, beam and crystal models. The program can be used on a cluster using `mpirun` if the mp.method parameter is set to mpi. For further detail, use the `-c` parameter as described above to see available options.

7. Future directions and software availability

`cctbx.xfel` is installed for all users at LCLS in `/reg/g/cctbx`. Instructions for setting up the appropriate environment for its use are available at http://cci.lbl.gov/xfel. `cctbx.xfel` bundled with DIALS is completely open source and is distributed through SourceForge and Git (see installation instructions for standalone packages at http://dials.github.io).

A graphical user interface (GUI) for processing XFEL data at LCLS using `cctbx.xfel` has been developed. Users interested in helping beta test the interface are invited to contact the authors. Further, example data useful for practicing the commands presented here are available on request.

8. Acknowledgements

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9. References


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**ARTICLES**

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