A model is not a structure: Using chemistry and physics to right wrongs and get useful biological information along the way.

Edward H. Snell CEO Hauptman Woodward Medical Research Institute

Pessimists, Optimists, and Crystallographers

Consider a glass of water

Pessimist (the glass is half empty)

> Crystallographer (the glass is completely full)

Optimist (the glass is half full)

Only approximately 11% of the proteins we target for crystallography yield a crystallographic structure.

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On the need for an international effort to capture, share and use crystallization screening data

When crystallization screening is conducted many outcomes are observed but typically the only trial recorded in the literature is the condition that yielded the crystal(s) used for subsequent diffraction studies. The initial hit that was optimized and the results of all the other trials are lost. These missing results contain information that would be useful for an improved general understanding of crystallization. This paper provides a report of a crystallization data exchange (XDX) workshop organized by several international large-scale crystallization screening laboratories to discuss how this information may be captured and utilized. A group that administers a significant fraction of the world's crystallization screening results was convened, together with chemical and structural data informaticians and computational scientists who specialize in creating and analysing large disparate data sets. T Acta Cryst. (2012). F68 crystallization ontology for the crystallization community was proposed. This paper (by the attendees of the workshop) provides the thoughts and rationale leading to this conclusion. This is brought to the attention of the wider audience of crystallographers so that they are aware of these early efforts and can contribute to the process going forward.

At least 99.8% of crystallization experiments produce an outcome other than crystallization.

Fantasy

Crystallize

Crystallography Requires Crystals

No crystal …

No crystallography ….

No crystallographer ….

High-throughput crystallization is easy

Efficient High-Throughput Crystallization is hard

- Successful high-throughput crystallization approaches require efficiency
	- The methodology must be equal or better to any other methods
	- The amount of sample used should be minimal
	- The amount of information obtained needs to be maximal and interpretable.
	- The results must be useable, reproducible and if necessary scalable.
	- Single point failures must be eliminated or minimized

The Crystallization Screening laboratory at the Hauptman-Woodward Medical Research Institute

Since February of 2000 the High Throughput Search (HTS) laboratory has been screening potential crystallization conditions as a high-throughput service

The HTS lab screens samples against three types of cocktails:

- 1. Buffered salt solutions varying pH, anion and cation and salt concentrations
- 2. Buffered PEG and salt, varying pH, PEG molecular weight and concentration and anion and cation type
- 3. Almost the entire Hampton Research Screening catalog.

The HTSlab has investigated the crystallization properties of over 15,000 individual proteins archiving approximately 140 million images of crystallization experiments.

The HWI crystallization cocktail screen.

The 1536 diverse chemical cocktails (Luft et al., 2003). The 984 in-house conditions comprise a incomplete factorial sampling of 36 salts, eight buffers, and 5 different PEGs.

The remainder of 1536 cocktails are comprised of commercial screens available from Hampton Research. Specifically, in order of use; the Natrix Screen, Quick Screen, Nucleic Acid Screen, Sodium Malonate Grid, PEG/Ion, PEG 6000 Grid, Ammonium Sulfate Grid, Sodium Chloride Grid, HT Screen, Index and the SaltRx screen.

Simplified phase diagram for crystallization

Precipitant Concentration

The crystallization method used is micro-batch under oil with 200 nl of protein solution being added to 200 nl of precipitant cocktail in each well of a 1536 well plate.

Wells are imaged before filling, immediately after filling then weekly for six weeks duration with images available immediately on a secure ftp server.

Several software utilities for viewing and analyzing data are available.

Outcomes

The protein data bank

- The Protein Data Bank contains depositions for 108,607 biological macromolecules.
- Some 90,506 of those are from data derived by Xray crystallography.
- Simple validation tests are available but a deposition is still accepted even if a test is failed.
- How accurate are the 'structures' in the PDB?

What are the errors, if any?

- Residues have well defined geometries.
- Sequence information is well known.
- Potential problems are:
	- Structural perturbation due to radiation damage
	- Incorrect ligand identification
	- Missing ligands
	- Just generally bad refinement
	- Crystallographic oligomer

How common is the problem?

- More common than you may think
- The examples presented on the previous slide are in the PDB and all come from here ….
- Despite care and diligence, errors still get through
- There are serious problems in many models yet the non-crystallographic community use these as 'structures' on the assumption that the model accurately represents the structure

How can we over come these problems?

- Structural perturbation due to radiation damage
	- Radiation damage studies, knowledge of the chemical processes and signatures
- Incorrect ligand identification
	- Better ligand treatment during refinement
	- Careful analysis of the crystallization conditions
	- Analysis of the sample pre or post crystallization
- Missing ligands
	- Similar approaches to the above
- Just generally bad refinement
	- To paraphrase Bernard Rupp, sometimes is worthwhile to look at the map!
- Crystallographic oligomer
	- Solution scattering

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Careful analysis of crystallization conditions

Molecular Fingerprints

Molecular fingerprints are representations of chemical structures designed to capture molecular activity.

We use atomic properties and a SMILES string to capture six components:

- 1. Atomic number
- 2. Number of directly-bonded neighbors
- 3. Number of attached hydrogens
- 4. The atomic charge
- 5. The atomic mass
- 6. If the atom is contained in a ring

These components are calculated for the whole molecule in an iterative manner starting from an arbitrary non-hydrogen.

Example: Sodium chloride, NaCl

Sodium [11,0,0,1,22.99,0] Chlorine [17,0,0,-1,35.45,0]

Starting from Na two, properties are associated with Na and encoded by: $(3,855,292,234,1)$ and $(3,737,048,253, 1)$ ^{*}

One property is associated with Cl and encoded by: (2,096,516,726,1)

This information is stored in single integer with bits 3,855,292,234, 3,737,048,253 and 2,096,516,726 set to on.

* Rodgers and Hahn, J. Chem. Inf. Model. 2010, 50, 742-754

Cocktail Fingerprints

Cocktail fingerprints combine the molecular fingerprints and account for the molarity of each in the crystallization cocktail.

For example, consider a very simple example: 0.1 M sodium chloride and 0.1 M ammonium sulfate

Molecular fingerprint: Sodium chloride [(3855292234, 1),(3737048253, 1),(2096516726, 1)] Ammonium chloride [(847680145, 1), (3855292234, 1),(2214760707, 1)]

Bit (3855292234, 1) is common in both so we set the bit count to 2 and multiply by the molar concentration

Cocktail fingerprint: [(3855292234, 0.2),(3737048253, 0.1),(2096516726, 0.1) (847680145, 1),(2214760707, 0.1)]

The bits are stored in a single 64 bit number with the bit counts stored in a sequential array

Comparing Cocktail Fingerprints

Take a real example of two crystallization screening cocktails as stored in our database

First convert all concentrations to molarity

Cocktail C1249 contains 30% (v/v) MPD. This is converted to 2.349 M. PEGs are more problematic as they can be polydispersive in which case the average molecular weight is used.

The cocktail fingerprint is calculated using the molecular fingerprint for each component and its molar concentration

Where F_k is the cocktail fingerprint, *i* is the number of components, *f* the molecular fingerprint and *c* the concentration

An example of two cocktail fingerprints

```
C1249 = [ (2245273601, 2.35), (2214760707, 0.02), (3537123720, 4.70), (864942730, 0.10)(1614748561, 2.35), (786100370, 2.35), (864666390, 0.34), (3537119515, 2.35),
(3925650716, 0.02), (2246728737, 7.15), (864662311, 4.70), (1582611257, 2.35),
(3737048253, 0.10), (3855292234, 0.04), (864942795, 0.10), (2245384272, 2.35),
(3992738647, 2.35), (1510323402, 0.10), (248253150, 2.35), (1542633699, 2.35),
(3219326737,0.10), (2246699815,0.10), (2355142638,2.35), (2245277810,2.35),
(1542631284, 2.35), (2096516726, 0.10), (3545365497, 0.10), (1510328189, 0.10)C0160 = [ (864942730, 0.20), (951748626, 0.10), (2143075994, 0.10), (2227993885, 0.10),(2968968094, 0.40), (192851103, 0.10), (2092489639, 0.10), (2604889258, 0.10),
(2880892204, 0.10), (1535166686, 0.10), (4226502584, 0.20), (825302073, 0.10),
(3855292234, 4.48), (1412710081, 0.20), (2828037323, 0.10), (2228063684, 0.20),
(569967222, 0.10), (2105180129, 0.10), (2803848648, 0.20), (4055698890, 0.10)(864942795, 0.10), (2808066764, 0.20), (2245384272, 0.40), (4023654873, 0.10),
(3336755162, 0.10), (999334238, 0.10), (1789200865, 0.10), (864662311, 0.10),
(3737048253, 4.48), (2096516726, 4.48), (2257970297, 0.10), (1634606847, 0.10)
```
Each is encoded in a single hashed number.

Comparing Cocktail Fingerprints (worked)

Take a real example of two crystallization screening cocktails

- 1. Convert all component concentrations to molarity. Cocktail C1249 contains 30 % (v/v) of MPD which we must first convert to molarity using the following equation: $molarity = %v/v * ((density/mw) * 1000)$. Plugging in the values for MPD we get: $2.349 = 0.30 * ((0.9254/118.1742) * 1000)$
- 2. Compute cocktail fingerprints using the molecular fingerprints for each component and it's molar concentration, as described in the previous section and equation (1). Cocktail fingerprints for C1249 and C0160 are listed below (each component fingerprint was computed using RDKit):

 $C1249 = [(2245273601, 2.35)$, $(2214760707, 0.02)$, $(3537123720, 4.70)$, $(864942730, 0.10)$ $(1614748561, 2.35)$, $(786100370, 2.35)$, $(864666390, 0.34)$, $(3537119515, 2.35)$, $(3925650716, 0.02)$, $(2246728737, 7.15)$, $(864662311, 4.70)$, $(1582611257, 2.35)$, $(3737048253, 0.10)$, $(3855292234, 0.04)$, $(864942795, 0.10)$, $(2245384272, 2.35)$, $(3992738647, 2.35)$, $(1510323402, 0.10)$, $(248253150, 2.35)$, $(1542633699, 2.35)$ (3219326737,0.10), (2246699815,0.10), (2355142638,2.35), (2245277810,2.35), $(1542631284, 2.35)$, $(2096516726, 0.10)$, $(3545365497, 0.10)$, $(1510328189, 0.10)$ $C0160 = [(864942730, 0.20), (951748626, 0.10), (2143075994, 0.10), (2227993885, 0.10),$ $(2968968094, 0.40)$, $(192851103, 0.10)$, $(2092489639, 0.10)$, $(2604889258, 0.10)$, $(2880892204, 0.10)$, $(1535166686, 0.10)$, $(4226502584, 0.20)$, $(825302073, 0.10)$, $(3855292234, 4.48)$, $(1412710081, 0.20)$, $(2828037323, 0.10)$, $(2228063684, 0.20)$, $(569967222, 0.10)$, $(2105180129, 0.10)$, $(2803848648, 0.20)$, $(4055698890, 0.10)$ $(864942795, 0.10)$, $(2808066764, 0.20)$, $(2245384272, 0.40)$, $(4023654873, 0.10)$, $(3336755162, 0.10)$, $(999334238, 0.10)$, $(1789200865, 0.10)$, $(864662311, 0.10)$, $(3737048253, 4.48)$, $(2096516726, 4.48)$, $(2257970297, 0.10)$, $(1634606847, 0.10)$

- 3. Compute the Bray-Curtis dissimilarity measure as described in equation (2) from the paper. Using the cocktail fingerprints in step 2 we obtain: $0.97 = \frac{|.1-.2|+|.04-.4.48|+|2.349-.4|+|4.698-.1|+|.1-.4.48|+|.1-.4.48|+|.4.698-|.1|+|.1-.4.48|+|.1-.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+|.4.48|+$
- 4. Compute the pH distance: $0.207 = \frac{|4.6 7.5|}{14}$
- 5. The final cocktail distance coefficient using $w = \{1, 1\}$ is: $CD_{coeff} = 0.589 = \frac{1}{2}(0.207 + 0.97)$

Cocktail similarity measures are not new.

We build on the original work by Janet Newman's in Melbourne, Australia who originated the concept of a similarity measure (termed C6) within crystallization to compare individual cocktails and different screening kits. (Newman J, Fazio VJ, Lawson B, Peat TS (2010) The C6 Web Tool: A Resource for the Rational Selection of Crystallization Conditions. Crystal Growth & Design 10: 2785-2792).

Our internal 1,536 screens are reformatted on a yearly basis to remove any conditions that produce salt crystals, to incorporate the latest screening developments, and building on internal research into crystallization processes.

In this example we apply both the C6 and our new similarity measure to two generations of screen where 96 conditions have been replaced with a new commercially available screen/

The Bray-Curtis dissimilarity measure is used to compute the dissimilarity.

$$
BC(F_i, F_j) = \sum_k |F_{ik} - F_{jk}| / \sum_k |F_{ik} + F_{jk}|
$$

This pH is incorporated along with the ability to weight individual components and the Cocktail Dissimilarity coefficient calculated.

$$
CD_{coeff} = \frac{1}{sum(w)} \left(\left(\frac{E(pH_i) - E(pH_j)}{14} \right) w_1 + BC(F_i, F_j) w_2 \right)
$$

The Cocktail Similarity coefficient given by:

$$
CS_{coeff}=1\!-\!CD_{coeff}
$$

Clustering then using a hierarchal display

The Dissimilarity Measure Over the Whole Screen

Automatic Clustering of the Results

Hierarchical clustering using a default max cophenetic distance cutoff of one standard deviation identified 28 clusters.

PEG based conditions

Salts with different anions and cations

A structural genomics target.

BfR192, is a 343 residue protein with a molecular weight of 39.77 kDa. For crystallization screening the protein was prepared at 7.4 mg/ml in a 5 mM DTT, 100 mM NaCl, 10 mM Tris-HCl, pH 7.5, 0.02% NaN₃ buffer.

Several potential crystallization conditions for BfR192 SelMet labeled protein were identified

The optimized conditions for crystallization combined 5µl of the protein at 7.4 mg/ml concentration was mixed with the precipitant containing 320mM potassium acetate, 100 mM sodium acetate, pH 6.5 in 1:1 ratio. Crystals appeared in one week.

PDB ID 3DMA as deposited in the PDB

Overlaying crystallization data

Overlaying Crystal Hits on the Cocktail Clustering

Conditions showing crystal hits are given for each cluster along with the total number of cocktails in that cluster.

A selection of cocktails that showed hits are listed on the outside of the dendogram. For clarity not all hits are shown

Cluster 20, PEG based, only 3 hits

Zoom in on Cluster 13

Identifies a pipette error

Clustering samples the phase diagram

Comparing Chemistry to Outcome: The Development of a Chemical Distance Metric, Coupled with Clustering and **Hierarchal Visualization Applied to Macromolecular** Crystallography

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Abstract

Many bioscience fields employ high-throughput methods to screen multiple biochemical conditions. The analysis of these becomes tedious without a degree of automation. Crystallization, a rate limiting step in biological X-ray crystallography, is one of these fields. Screening of multiple potential crystallization conditions (cocktails) is the most effective method of probing a proteins phase diagram and quiding crystallization but the interpretation of results can be time-consuming. To aid this empirical approach a cocktail distance coefficient was developed to quantitatively compare macromolecule crystallization conditions and outcome. These coefficients were evaluated against an existing similarity metric developed for crystallization, the C6 metric, using both virtual crystallization screens and by comparison of two related 1,536-cocktail highthroughput crystallization screens. Hierarchical clustering was employed to visualize one of these screens and the crystallization results from an exopolyphosphatase-related protein from Bacteroides fragilis, (BfR192) overlaid on this clustering. This demonstrated a strong correlation between certain chemically related clusters and crystal lead conditions. While this analysis was not used to quide the initial crystallization optimization, it led to the re-evaluation of unexplained peaks in the electron density map of the protein and to the insertion and correct placement of sodium, potassium and phosphate atoms in the structure. With these in place, the resulting structure of the putative active site demonstrated features consistent with active sites of other phosphatases which are involved in binding the phosphoryl moieties of nucleotide triphosphates. The new distance coefficient, CD_{coeff}, appears to be robust in this application, and coupled with hierarchical clustering and the overlay of crystallization outcome, reveals information of biological relevance. While tested with a single example the potential applications related to crystallography appear promising and the distance coefficient, clustering, and hierarchal visualization of results undoubtedly have applications in wider fields.

Citation: Bruno AE, Ruby AM, Luft JR, Grant TD, Seetharaman J, et al. (2014) Comparing Chemistry to Outcome: The Development of a Chemical Distance Metric, Coupled with Clustering and Hierarchal Visualization Applied to Macromolecular Crystallography. PLoS ONE 9(6): e100782. doi:10.1371/journal.pone.0100782

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Data Availability: The authors confirm that all data underlying the findings are fully available without restriction. The code used to evaluate the CDcoeff is open source and freely available at http://ubccr.github.io/cockatoo/ or directly from the authors. The crystallization images and cocktail data are large files (1,536 different images and metafiles) and available from the authors.

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Incorporating the correct ligands reduced the R and R_{free} from to 23.5% and 26.4% to 20.7% and 24.3% respectively.

The software is publically available and while it takes some time to run for each generation of screen it only has to be run once.

A Revised Structure Illustrating Mechanism

Biological implication of the phosphates identified

- The structure consists of two domains (N-terminal domain; residues 2 -212 and Cterminal domain residues 217-343) which are connected by a short loop – seen in the initial structure
- The N-terminal domain contains the DHH (Asp224-His225-His226) motif and the C-terminal domain contains a glycine-rich (GGGH-Gly308-Gly309-Gly310-His311) phosphate binding motif – seen but not identified in the initial structure.
- Three of the phosphates (presumably carried with the protein), and the potassium and the sodium ion are bound in the cl
-
- might anchor in this pocket.
-
- and polarization of the phosph nucleophilic attack.
-

• The phosphate ions interact with the important point here is not the details of the \overline{a} , • The location of the phosphate new information but that this information was • The putative active site has fea Potential function and mechanism was revealed. The which are involved in binding t While on could argue that these could have been • The possible roles of the active identified earlier many examples in the PDB have $\frac{d}{dx}$ • The space around the phosphal sample of structures and seen problems in many of $\frac{1}{3}$. obtained after the correct ligands were identified. ambiguous atoms – we have explored only a small them.

Elemental Analysis

Particle Induced X-ray Emission

The energy of an X-ray emitted when an atomic electron undergoes an energy transition between its shell and a vacant electron site in a lower energy shell (e.g. for an M to L shell transition, sulphur gives a 2.3 keV X-ray) gives an unambiguous identification of atoms.

Emission of the characteristic X-rays from a sample can be induced by an incident beam of high energy protons (Particle Induced X-ray Emission: PIXE).

High-throughput Sample Preparation

Dispense samples with a noncontact microarray printer

Up to 144 samples dispensed into a 384 well plate and printed into a 12x12 array of 60 um drops with 200 um spacing.

Up to five arrays can be mounted into a single sample holder giving a total of 720 samples per slide.

Scanning Proton Microprobe for PIXE analysis. 2-3 MeV protons emerge from the van de Graaff accelerator and are focussed by high precision magnets onto the sample. The whole beamline is kept under vacuum.

High-Throughput PIXE

- 34 samples analyzed chosen from NESG samples submitted to the highthroughput crystallization screening laboratory on the basis of a PDB model available and that the model in the PDB contained at least one metal ion.
- The samples used were split into four groups based on PIXE analysis
	- Those where the PDB was inconsistent with the PIXE data
	- Those where extra metals were seen in the PIXE data (but not present in the PDB)
	- Those that were consistent with the PIXE data.
	- Those that produced no signal.

Re-refinement

- 34 samples analyzed chosen on the basis of a PDB structure available and that structure containing at least one metal ion.
- The samples used were split into four groups based on PIXE analysis
	- Those where the PDB was inconsistent with the PIXE data
	- Those where extra metals were seen in the PIXE data (but not present in the PDB)
	- Those that were consistent with the PIXE data.
	- Those that produced no signal.

High-Throughput PIXE

- MicroPIXE can be used to determine the proportion of methionine substitution where no sulfur is present in the buffer.
- The concentration of an element is determined by fitting the area of the Xray peak corresponding to the element.
- If the total number of Se atoms per protein molecule is $\alpha_{\text{S}_{\text{e}}}$, the total number of S atoms left per protein molecule is α_{s} , and the original number of S atoms (cysteines + methionines) in the sequence was α then $\alpha = \alpha_{s} + \alpha_{s}$ and we can write: α_S = $c_S A_{Se} (\alpha - \alpha_{Se})$

$$
\alpha_{Se} = c_{Se} A_S \qquad \alpha_{Se}
$$

• Where A_s and A_{Se} are the atomic masses of S and Se respectively and c_s and $c_{\varsigma_{\mathsf{e}}}$ are the mass concentrations.

High-Throughput PIXE

- In our case the NESG buffer has Sulfur.
- However, all the proteins studied were expressed with SeMet for phasing purposes.
- The number of atoms of element Z per protein can be determined by

$$
\alpha_Z = \frac{c_Z}{c_{Se}} \frac{A_{Se}}{A_Z} \alpha_{Se}
$$

• Where $A_{\rm z}$ and $A_{\rm Se}$ are the atomic masses of element Z and Se respectively and c_z and c_Se are the mass concentrations determined from the PIXE spectrum.

Model in the PDB containing a metal from the crystallization cocktail and not protein

Model in the PDB containing an incorrect metal

Model in the PDB containing an extra misidentified metal

◆Presence of sodium and magnesium could not be confirmed at the proton energies used in these experiments. *Selenium signal was below 3 times the limit of detection, so accurate stoichiometries could not be established.

- Of the 34 samples analyzed, 9 were inconsistent with the PDB results, 9 had extra metals present, 18 were consistent, and 2 were unsuitable for analysis due to low protein concentration on the sample.
- In total, 18 of the 32 analyzable samples (56%) were not correctly or fully described in the PDB deposition.

Re-Refining 3DCP

including biophysical chemistry & molecular biology

Article pubs.acs.org/biochemistry

A closely related protein

LOURD RAIN VON COMMUNENT REMANDED OF HOME WANT ZANC AN THE ROLL site and had a catalytic efficiency of $\sim 10^3$ M⁻¹ s⁻¹. Expression of

the protein under iron-free conditions resulted in the production of an enzyme with a 2 order of magnitude improvement in catalytic efficiency and a mixture of zinc and manganese in the active site. Solvent isotope and viscosity effects demonstrated that proton transfer steps and product dissociation steps are not rate-limiting. X-ray structures of HPP were determined with sulfate, L-histidinol phosphate, and a complex of L-histidinol and arsenate bound in the active site. These crystal structures and the catalytic properties of variants were used to identify the structural elements required for catalysis and substrate recognition by the HPP family of enzymes within the amidohydrolase superfamily.

S Supporting Information

ABSTRACT: L-Histidinol phosphate phosphatase (HPP) catalyzes the hydrolysis of L-histidinol phosphate to L-histidinol and inorganic phosphate, the penultimate step in the biosynthesis of Lhistidine. HPP from the polymerase and histidinol phosphatase (PHP) family of proteins possesses a trinuclear active site and a distorted (β/α) -barrel protein fold. This group of enzymes is closely related to the amidohydrolase superfamily of enzymes. The mechanism of phosphomonoester bond hydrolysis by the PHP family of HPP enzymes was addressed. Recombinant HPP from Lactococcus lactis subsp. lactis that was expressed in Escherichia coli contained a mixture of iron and zinc in the active site and had a catalytic efficiency of $\sim 10^3$ M⁻¹ s⁻¹. Expression of

L-histidir

Metal content measured with an inductively coupled mass spectrometer

the protein under iron-free conditions resulted in the production of an enzyme with a 2 order of magnitude improvement in catalytic efficiency and a mixture of zinc and manganese in the active site. Solvent isotope and viscosity effects demonstrated that proton transfer steps and product dissociation steps are not rate-limiting. X-ray structures of HPP were determined with sulfate, L-histidinol phosphate, and a complex of L-histidinol and arsenate bound in the active site. These crystal structures and the catalytic properties of variants were used to identify the structural elements required for catalysis and substrate recognition by the HPP family of enzymes within the amidohydrolase superfamily.

Accurate Metal identification is important

- The original structure contained Fe and Zn.
- The revised structure shows the phosphate and Co
- The phosphate and tri-nuclear metal center are important for mechanism

Work in progress

- All the structures in the table are currently being re-refined
- Each is improved with the correct metal placed
- All will be revisited once completed to determine if there are any 'clues' to mechanism with the correct metal in place.

Important notes about the technique

- Because PIXE is an elemental analysis the sample does not have to be in any preserved state.
- Samples from years ago can be used to collect experimental data.
- The number and ratio of different metals (or other atoms) per protein molecule can be determined.
- Not discussed today, but the data reveals clear signatures in protein models that identify suspect metals.

Summary

- Crystallization analysis and elemental analysis have great potential in improving structural models.
- This improvement is needed as our limited study shows a greater than 50% error rate.
- Experimentally identifying errors defines signatures of those same errors in other structural models.
- The work leads to a potential quality control mechanism to identify suspect structural models.
- It also allows native metals (at least from expression) to be distinguished from opportune ones.

The Team

Andrew Bruno, Elspeth Garman, Geoffrey Grime, Joseph Luft, Amanda Ruby, Edward Snell, Elizabeth Snell, and Oliver Zeldin

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Available online at www.sciencedirect.com SCIENCE \bigcap DIRECT⁺

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Review

Elemental analysis of proteins by microPIXE

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Thank you and questions?

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