

Model Refinement: cryo-EM

Pavel Afonine



phenix-online.org



lbl.gov



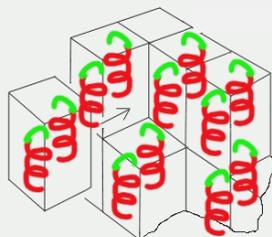
qrefine.com

Missoula, MT

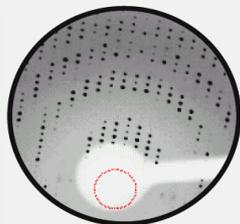
June 28th 2024

Refinement in Phenix

Crystallography



Initial model



Experimental
data

A priori
knowledge

Score

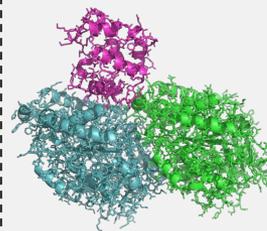
Modify model
parameters

Improved
model

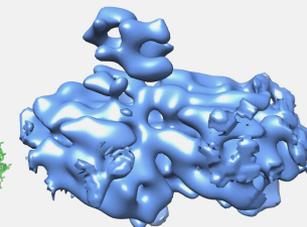
phenix.refine

Available since 2005

Cryo-EM



Initial model



Experimental
data

A priori
knowledge

Score

Modify model
parameters

Improved
model

phenix.real_space_refine

Available since 2013

Atomic model refinement: crystallography vs cryo-EM

Crystallographic refinement

- Improving model improves map
 - (2mFo-DFc, Model phase), (mFo-DFc, Model phase)
 - Better model leads to better map
 - Better map leads to more model built
 - Improving model in one place lets build more model elsewhere in the unit cell
 - Refine all model parameters (XYZ, B) from start to end of structure solution
 - Build solvent (ordered water) early
- Experimental data never changed
- Data / restraints weight is global and time expensive to find best value
- Whole model needs to be refined

Cryo-EM refinement

- Changing model does not change map
 - Build solvent (water) last
 - Get as complete and accurate model as possible before refining B factors and occupancies
- Experimental data changes a lot during the process (filtering, boxing, using maps with implied symmetry or not, etc.)
 - What map to use in refinement?
 - Refined B factors depend on map used
- Data / restraints weight can be local and is always optimal
- Boxed parts of the model can be refined

Atomic model refinement: *phenix.real_space_refine*



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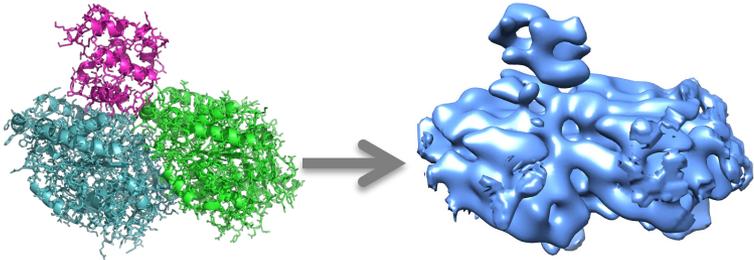
Real-space refinement in *PHENIX* for cryo-EM and crystallography

Pavel V. Afonine,^{a,b*} Billy K. Poon,^a Randy J. Read,^c Oleg V. Sobolev,^a Thomas C. Terwilliger,^{d,e} Alexandre Urzhumtsev^{f,g} and Paul D. Adams^{a,h}

How we evaluate refinement progress (model-to-map fit) or what's the analogue of crystallographic R-factor?

Model-to-map fit validation: CC_{MASK}

Model to map fit



$$CC_{\text{MASK}} = \frac{\sum \rho_{\text{obs}} \rho_{\text{calc}}}{(\sum \rho_{\text{obs}}^2 \sum \rho_{\text{calc}}^2)^{1/2}}$$

ρ_{obs} = experimental map

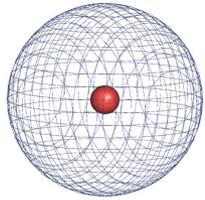
ρ_{calc} = model calculated map

- Easy interpretation: -1: anticorrelation, 0: no correlation, 1: perfect correlation
- Uses all atomic model parameters (XYZ, B-factors, occ, atom type)
- Not specific to map type (any map: x-ray, neutron, electron, cryo-EM, ...)
- Can be calculated locally (per atom, residue, chain, molecule, whole box, ...)
 - Local resolution can be trivially taken into account

Metric	Expected value
CC_{MASK}	Poor: < 0.3
	So-so: 0.3-0.6
	Good: > 0.6

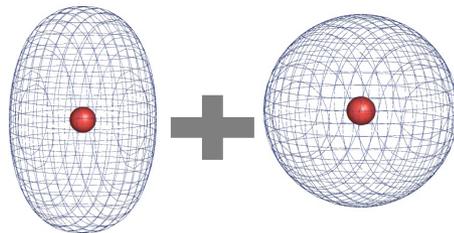
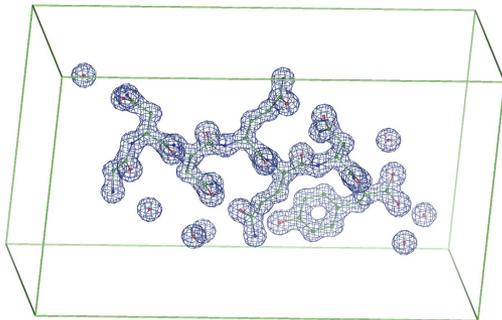
Model-to-map fit validation: CC_{MASK}

- Gaussian IAM (Independent Atom Model)



$$\rho_{atom}(\mathbf{r}, \mathbf{r}_0, B, q) = q \sum_{k=1}^5 a_k \left(\frac{4\pi}{b_k + B} \right)^{3/2} \exp\left(-\frac{4\pi^2 |\mathbf{r} - \mathbf{r}_0|^2}{b_k + B} \right)$$

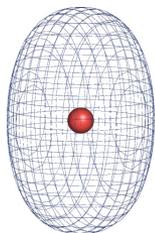
ATOM	25	CA	PRO	A	4	31.309	29.489	26.044	1.00	57.79	C
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$$\rho_{\text{MODEL}}(\mathbf{r}) = \sum_{i=1}^{N_{\text{atoms}}} \rho_{\text{atoms}}(\mathbf{r})$$

Model map

- Gaussian IAM (Independent Atom Model)
- Anisotropic:



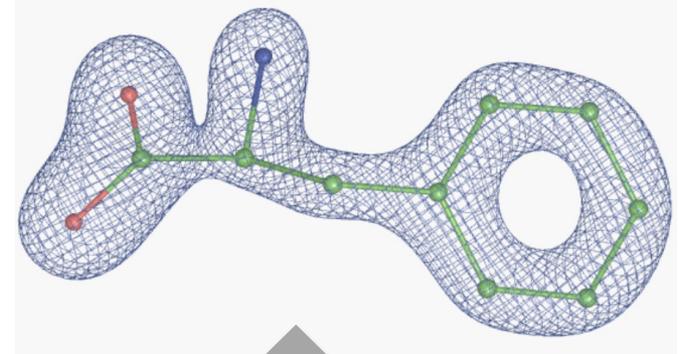
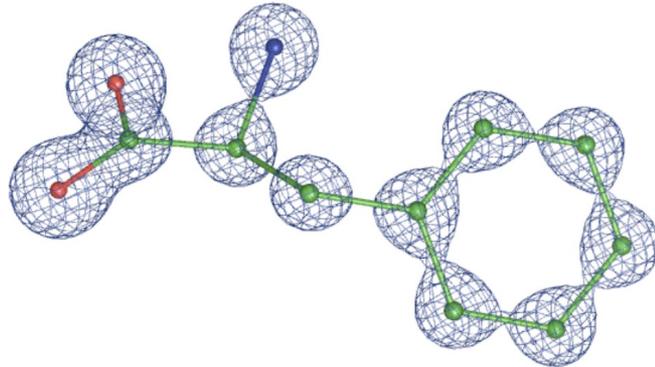
$$\rho_{atom}(\mathbf{r}, \mathbf{U}, q) = q \sum_{j=1}^5 \frac{q a_j (4\pi)^{3/2}}{|8\pi^2 \mathbf{U}_{cart} + b_j \mathbf{I}|^{1/2}} \exp\left(-4\pi^2 (\mathbf{r} - \mathbf{r}_0)^T \mathbf{A}^T [8\pi^2 \mathbf{U}_{cart} + b_j \mathbf{I}]^{-1} \mathbf{A} (\mathbf{r} - \mathbf{r}_0)\right)$$

ATOM	25	CA	PRO	A	4	31.309	29.489	26.044	1.00	57.79	C	
ANISOU	25	CA	PRO	A	4	8443	7405	6110	2093	-24	-80	C

Model-to-map fit validation: CC_{MASK}

3Å model-calculated map

Exact model map



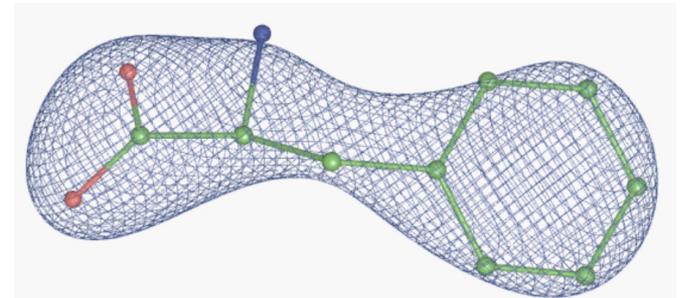
CC_{MASK}



CC_{MASK}



3Å experimental map



$$\rho_{\text{MODEL}}(\mathbf{r}) = \sum_{i=1}^{N_{\text{atoms}}} \rho_{\text{atoms}}(\mathbf{r})$$

- FT exact model map
- Remove terms up to specified resolution
- FT back to real space to get a Fourier image = “Model map”

**Other popular model-to-map fit metrics and reasons
why they are not as good as CCmask**

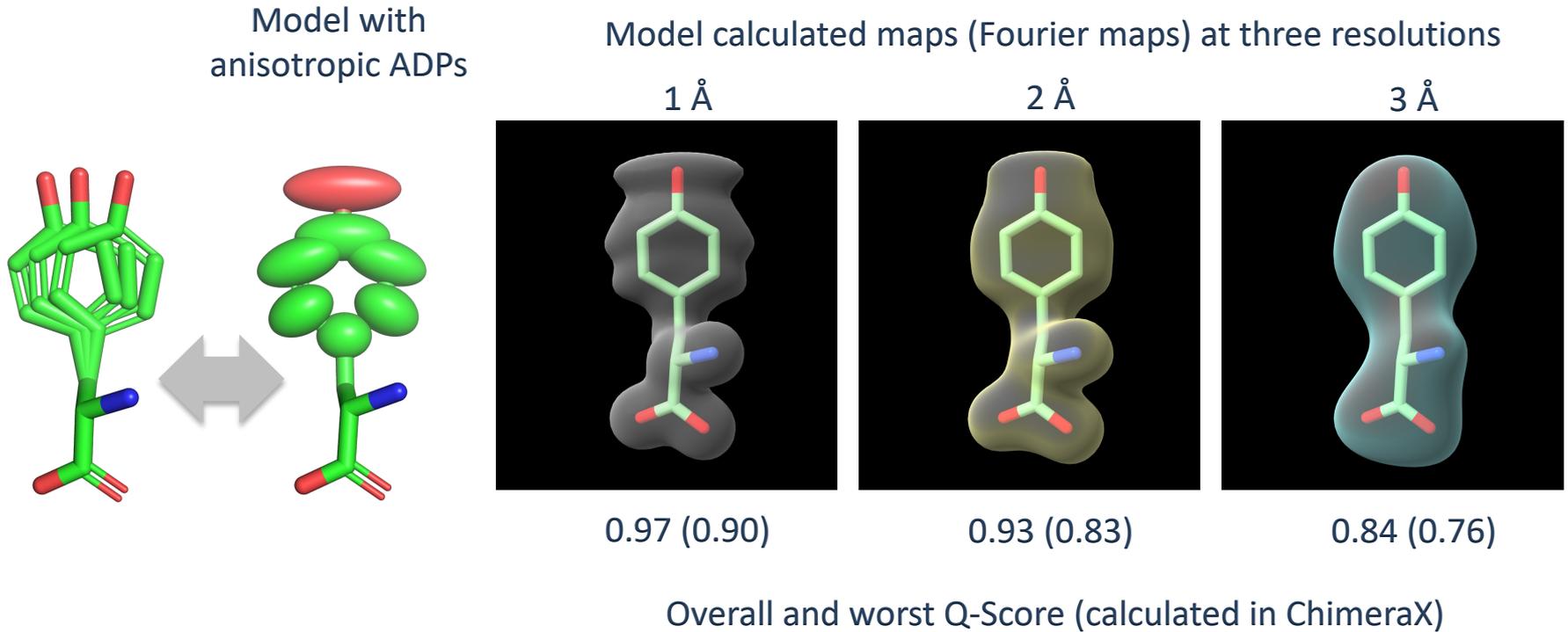
Atom inclusion

- **Atom inclusion:** fraction of atoms inside molecular envelope contoured at a given level
 - Contouring threshold: Arbitrarily? What is optimal level?
 - No use of atomic model parameters such as ADP, occupancy, atom type, ...
 - Does not compare shape of density:
 - How SER placed into PHE density is going to score?
 - How water O placed into Mg peak will score?
 - Does not account for missing atoms
 - Does not use map type (x-ray, neutron, electron)
 - Partially occupied atoms (alternative conformations):
 - Chosen level for fully occupied atoms needs to be scaled by occupancy for partially occupied atoms

Q-Score

- **Q-score:** measure the resolvability of individual atoms in a cryo-EM map, using an atomic model fitted to or built into the map
 - No use of atomic model parameters such as ADP, occupancy, atom type, ...
 - Shape of density:
 - How SER placed into PHE density is going to score?
 - How water O placed into Mg peak will score?
 - Does not account for missing atoms (it shouldn't given the definition)
 - Alternative conformations are **not** handled
 - How anisotropic atoms are **not** handled
 - Does not use map type (x-ray, neutron, electron)

Example: Q-Score for exact (model-generated) map



- Why Q-Score is not perfect (=1) given these are exact model-generated maps?
- Why it varies with the resolution?

Validation reports (RCSB): only Q-score and atom inclusion

Structure Summary 3D View Annotations Experiment Sequence Genome Versions

Biological Assembly 1

6KIQ

Complex of yeast cytoplasmic dynein MTBD-High and MT with DTT

PDB DOI: 10.2210/pdb6KIQ/pdb EM Map EMD-9997: EMDB EMDataResource

Classification: **MOTOR PROTEIN/STRUCTURAL PROTEIN**

Organism(s): *Sus scrofa*, *Saccharomyces cerevisiae* S288C

Expression System: *Escherichia coli*

Mutation(s): Yes

Deposited: 2019-07-19 Released: 2020-03-04

Deposition Author(s): Komori, Y., Nishida, N., Shimada, I., Kikkawa, M.

Funding Organization(s): Japan Science and Technology, Japan Agency for Medical Research and Development (AMED)

Experimental Data Snapshot

Method: ELECTRON MICROSCOPY

Resolution: 3.62 Å

Aggregation State: FILAMENT

Reconstruction Method: HELICAL

wwPDB Validation

3D Report Full Report

Metric	Percentile Ranks	Value
Clashscore		10
Ramachandran outliers		10.7%
Sidechain outliers		12.4%

Worse Better

■ Percentile relative to all structures

□ Percentile relative to all EM structures

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Full wwPDB EM Validation Report

EMD-

wwPDB Validation

3D Report

Full Report

9.5 Map-model fit summary ⓘ

The table lists the average atom inclusion at the recommended contour level (0.125) for the entire model and for each chain.

Chain	Atom inclusion	Q-score
All	0.9062	0.4550
M	0.5810	0.3210
a	0.9659	0.4790
b	0.9656	0.4730

Model-to-map fit statistics is insufficient and very well hidden!

Refinement: practical considerations

- Final stages
 - Refine B-factors (Atomic Displacement Parameters)
 - Group B factor or individual
 - Refine occupancies
 - Use Hydrogen atoms (and keep them in the final model!)
 - Add water (phenix.douse: command line and GUI):
 - Also available in ChimeraX

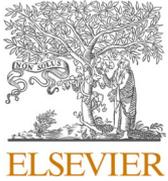
mmCIF

- mmCIF file format for atomic models
 - Mandatory use for crystallographic models since July 2019
 - PDB formatted files are not accepted any more
 - Some cryo-EM models may be too large to fit into PDB file format
 - *Phenix* provides full support for mmCIF I/O

	letters to the editor
 <p>STRUCTURAL BIOLOGY</p> <p>ISSN 2059-7983</p>	Announcing mandatory submission of PDBx/mmCIF format files for crystallographic depositions to the Protein Data Bank (PDB)
Received 21 February 2019 Accepted 3 April 2019	<p>Paul D. Adams,^{a,b} Pavel V. Afonine,^a Kumaran Baskaran,^c Helen M. Berman,^d John Berrisford,^e Gerard Bricogne,^f David G. Brown,^g Stephen K. Burley,^{d,h,i,*} Minyu Chen,^j Zukang Feng,^d Claus Flensburg,^f Aleksandras Gutmanas,^e Jeffrey C. Hoch,^{k,*} Yasuyo Ikegawa,^j Yumiko Kengaku,^j Eugene Krissinel,^l Genji Kurisu,^{j,*} Yuhe Liang,^d Dorothee Liebschner,^a Lora Mak,^e John L. Markley,^{c,*} Nigel W. Moriarty,^a Garib N. Murshudov,^m Martin Noble,ⁿ Ezra Peisach,^d Irina Persikova,^d Billy K. Poon,^a Oleg V. Sobolev,^a Eldon L. Ulrich,^c Sameer Velankar,^{e,*} Clemens Vornrhein,^f John Westbrook,^d Marcin Wojdyr,^{f,l} Masashi Yokochi^j and Jasmine Y. Young^d</p>
Edited by R. J. Read, University of Cambridge, England	

Variability refinement

Treasuring conformational changes



Contents lists available at [ScienceDirect](https://www.sciencedirect.com)

BBA - Biomembranes

journal homepage: www.elsevier.com/locate/bbamem



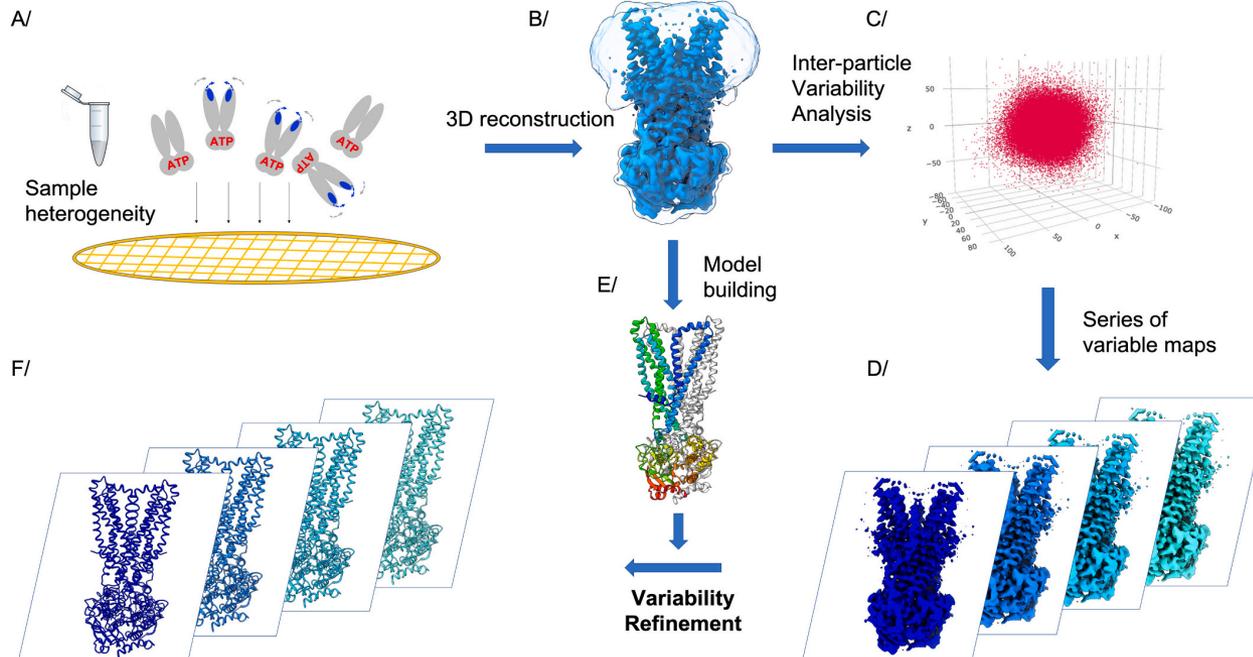
Review

Conformational space exploration of cryo-EM structures by variability refinement

Pavel V. Afonine^{a,*}, Alexia Gobet^b, Loïck Moissonnier^b, Juliette Martin^b, Billy K. Poon^a, Vincent Chaptal^{b,*}

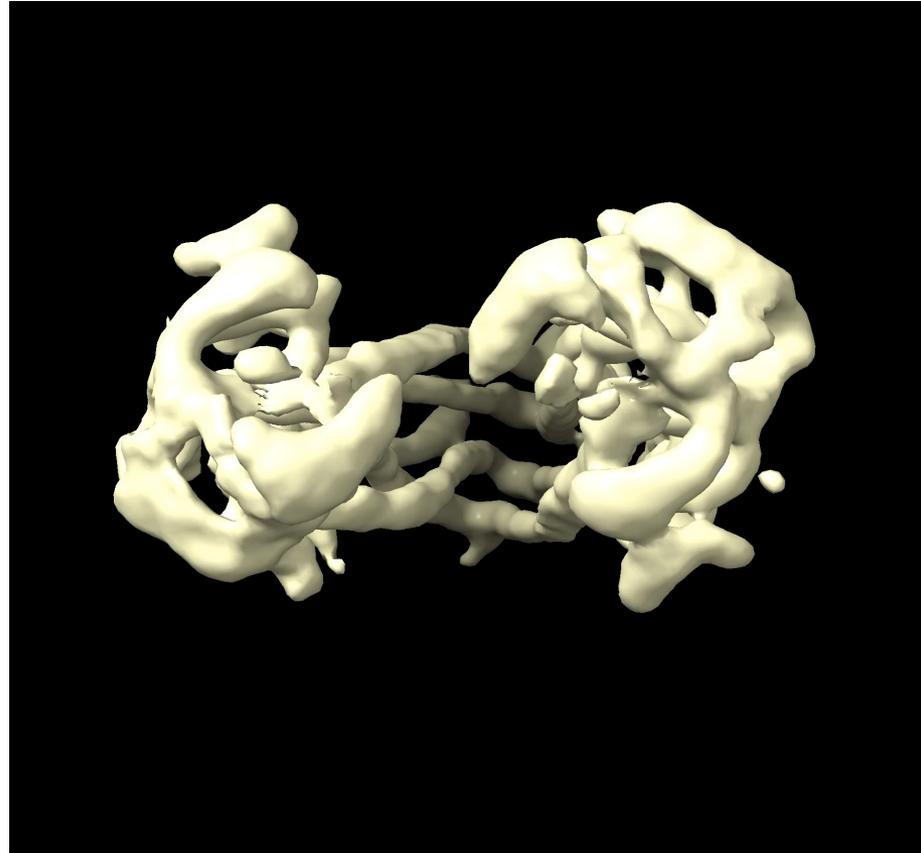
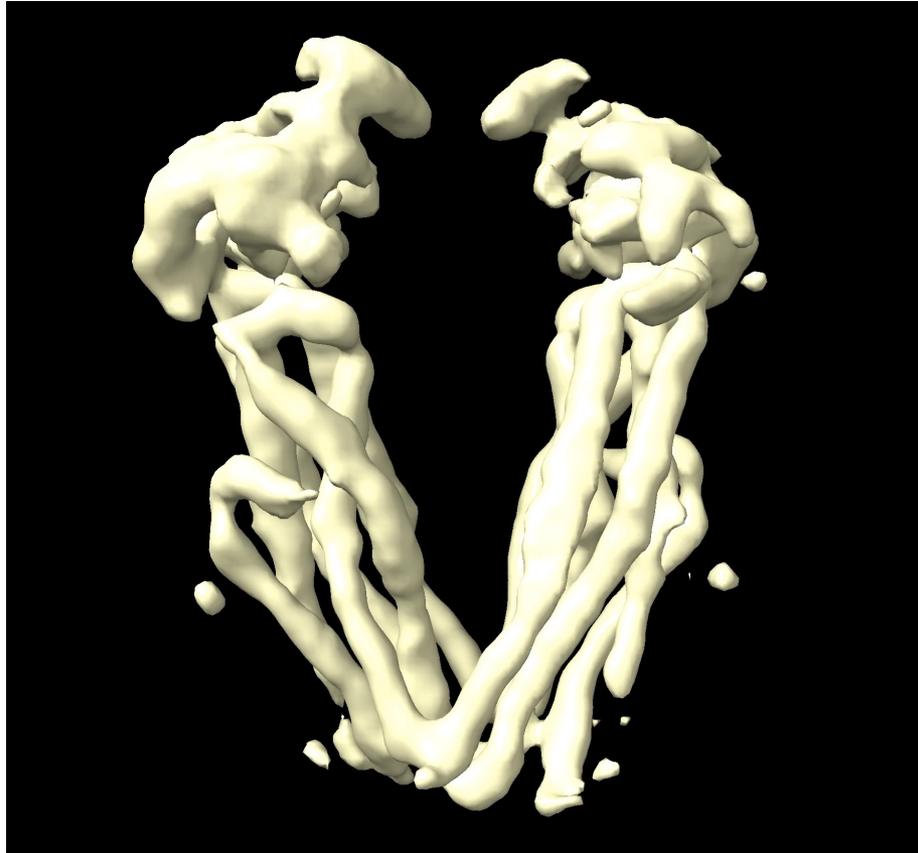
^a Molecular Biosciences and Integrated Biomaging, Lawrence Berkeley National Laboratory, 1 Cyclotron Road, Berkeley, CA 94720, USA

^b Molecular Microbiology and Structural Biochemistry, UMR5086 CNRS University Lyon1, 7 passage du Vercors, 69007 Lyon, France



Maps

ABC transporter BmrA (unpublished!)



phenix.varref – Phenix tool to represent ensemble of maps with ensemble of atomic models

phenix.varref

map1.mrc ... mapN.mrc

model.pdb

resolution=3

nproc=100

models_per_map=100

Output: ensemble of refined models that represents all maps

Workflow

- Input model and maps
- Order maps by similarity using CC_{box}
- Identify the map that is closest to input model (by CC_{mask})
 - This is the starting point for the first refinement
 - Generate ensemble of 100 perturbed models (by MD)
 - Refine each model with *phenix.real_space_refine*
 - Combine all refined models to yield overall best fitting model
- Refine ensemble of refined models against the next closest map
 - Combined all refined models to yield overall best fitting model
- ...and so on for all maps.
- Result:
 - N models corresponding to N maps
 - 100 models per map (can be used to estimate uncertainty)

Refined ensembles of models

