Low Resolution Refinement

Macromolecular Crystallography School Madrid, May 2017

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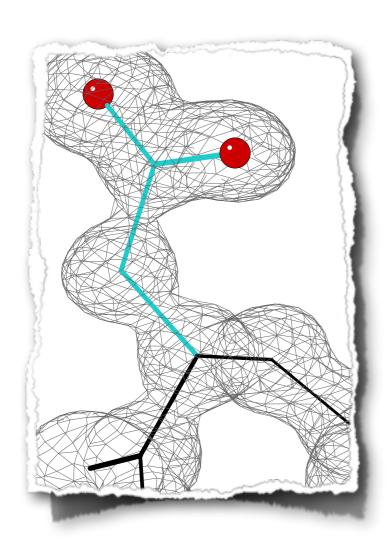
Macromolecular Crystallography

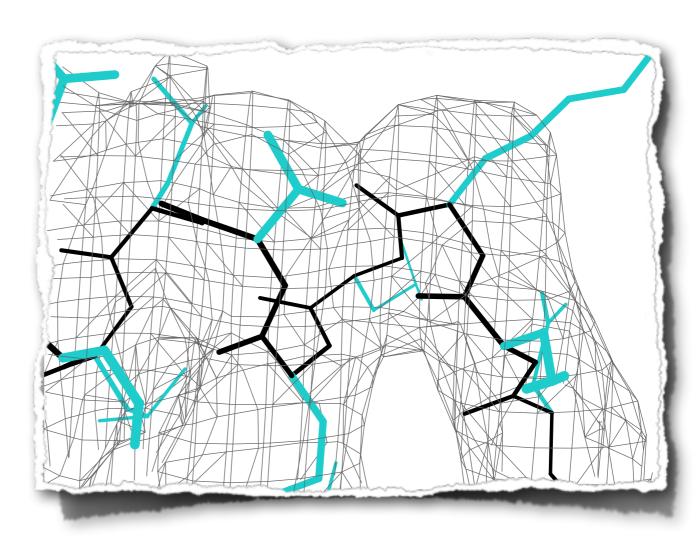
PDBID: 2gkg

Resolution: I.00Å

PDB ID: 3k7a

Resolution: 3.80Å





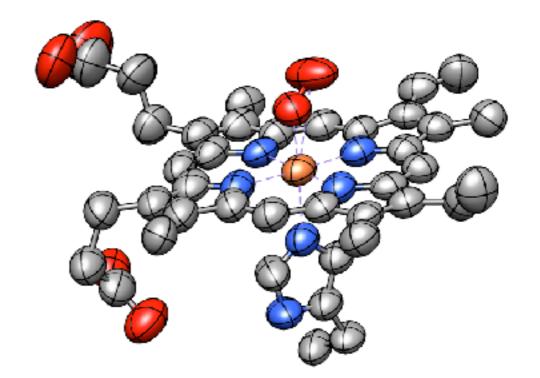
- Many challenges, but low resolution data is increasingly an issue:
- How to interpret "featureless" maps (pattern matching, chemical constraints)
- How to optimize models with sparse data (prior information)

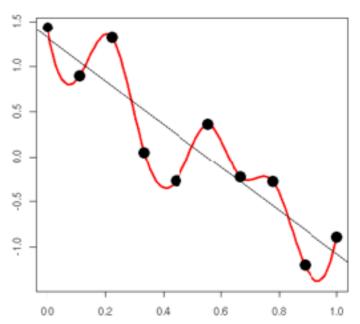




The Challenge of Too Few Data

- With only low resolution data we typically have too many parameters to optimize
 - Atomic coordinates, displacement parameters
- Underdetermined optimization problems lead to overfitting of the data
- To help address overfitting we can:
 - Add prior information to reduce the number of effective parameters
 - Remove parameters
- Current refinement methods do not define a reasonable chemical result in the absence of data









New Methods in Phenix for Improving Models

- Using prior structural knowledge as additional restraints:
 - Secondary structure
 - Protein mainchain conformations (Ramachandran)
 - Related high resolution structures as restraints
 - Multiple copies of the same molecule as restraints (c.f. local NCS restraints in SHELX)
- Automated correction of models during refinement using prior knowledge of stereochemistry:
 - Fixing of rotamers
 - Flipping of side chains





Reference model restraints (Jeff Headd)

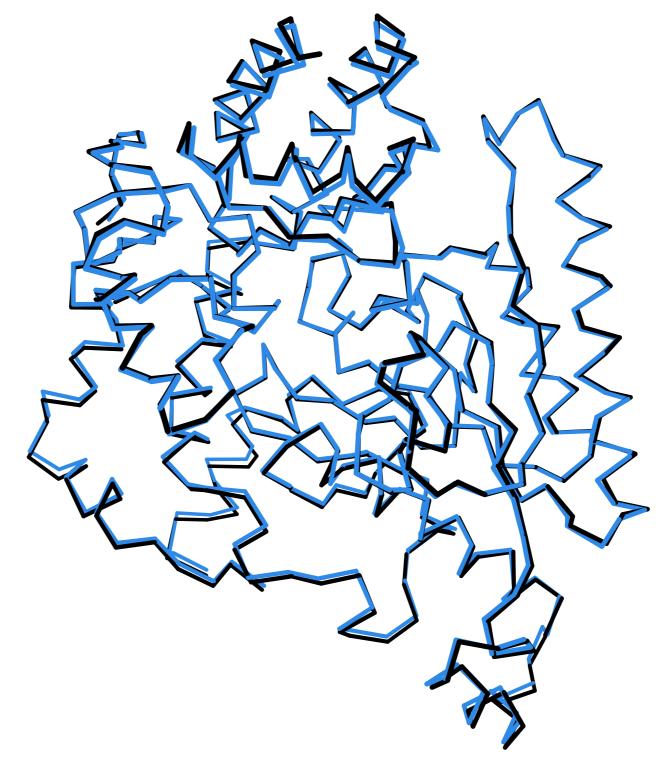




IGTX and IOHV

IGTX: 3.0 Å

IOHV: 2.3 Å

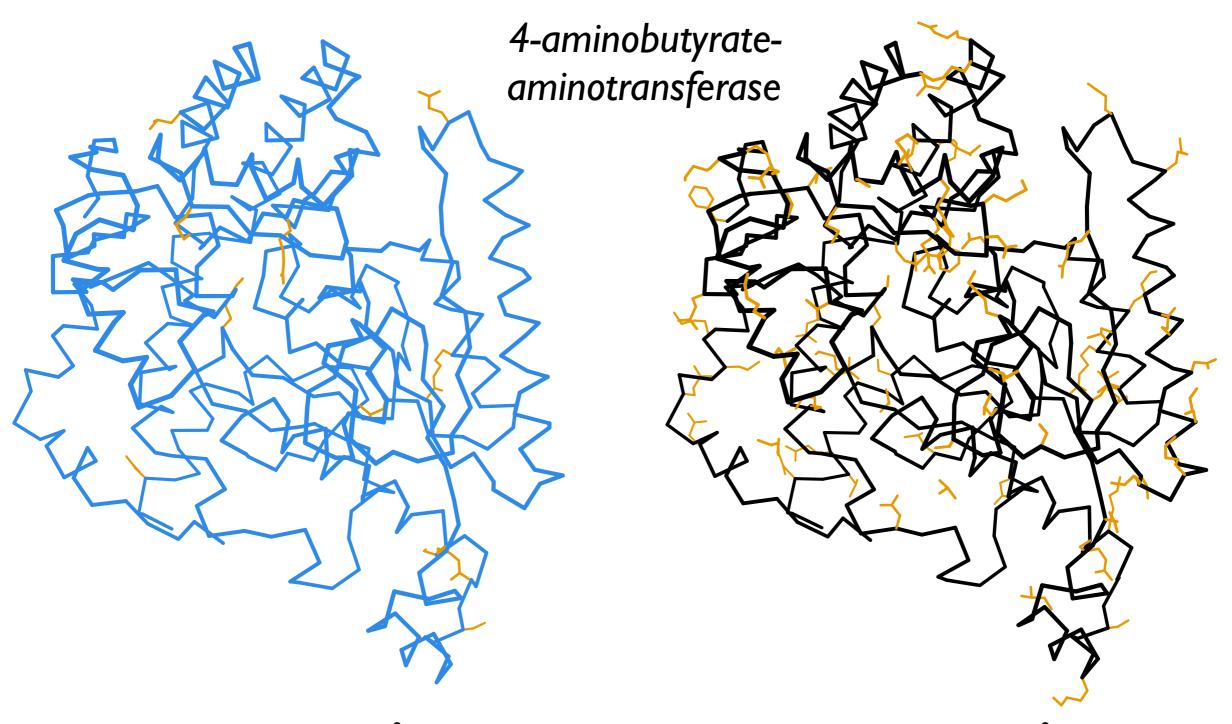




4-aminobutyrate-aminotransferase



IGTX and IOHV



IOHV: 2.3 Å

IGTX: 3.0 Å





Reference Model Restraints

Combines two concepts:

- Pre-correct rotamer outliers
 - Set rotamer outliers in the model to match the torsion angles of the reference model if the reference model has an acceptable rotamer at that position and there is no significant clash or density mismatch
- Generate reference torsion restraints
 - Restrain each torsion angle in the working model to the corresponding torsion angle in the reference model
 - Chains are aligned using SSM alignment to allow for sequence differences
 - Restraints take the form of a modified harmonic 'top-out' potential that allows for structural differences





Reference model restraints

$$E_{total} = \sum_{i=1}^{n} E_i$$

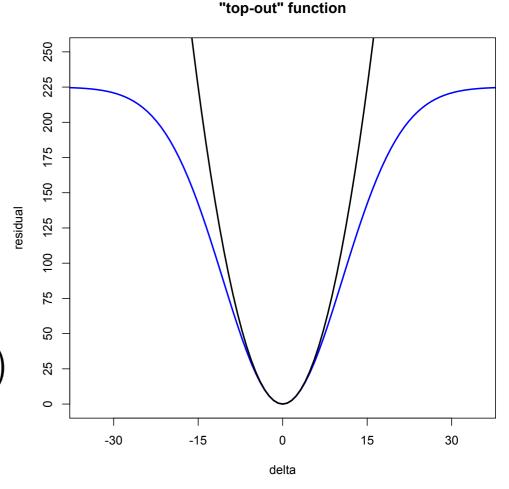
Simple harmonic potential: $E_i = w\Delta_i^2$

$$L_i - \omega \Delta_i$$

'Top-out' potential:
$$E_i = au(1.0 - e^{rac{-\Delta_i^2}{l^2}})$$

$$\tau = wl^2$$

$$w = \frac{1}{\sigma^2}$$



Similar potentials are used in **REFMAC5** and BUSTER -Geman-McClure robust estimator function

where σ is the ESD, Δ is the difference between the model dihedral and reference dihedral, and I is a 'limit' parameter that limits how far the model dihedral may vary from the reference dihedral before being shut off.



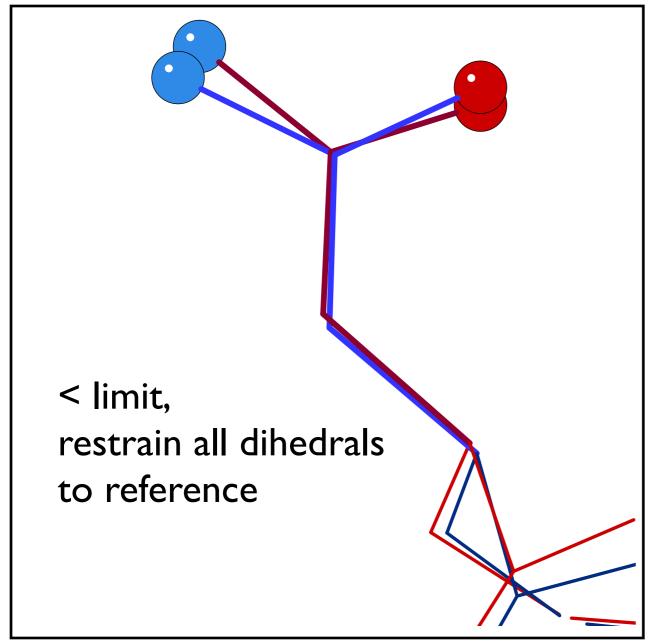
developed by Ralf Grosse-Kunstleve

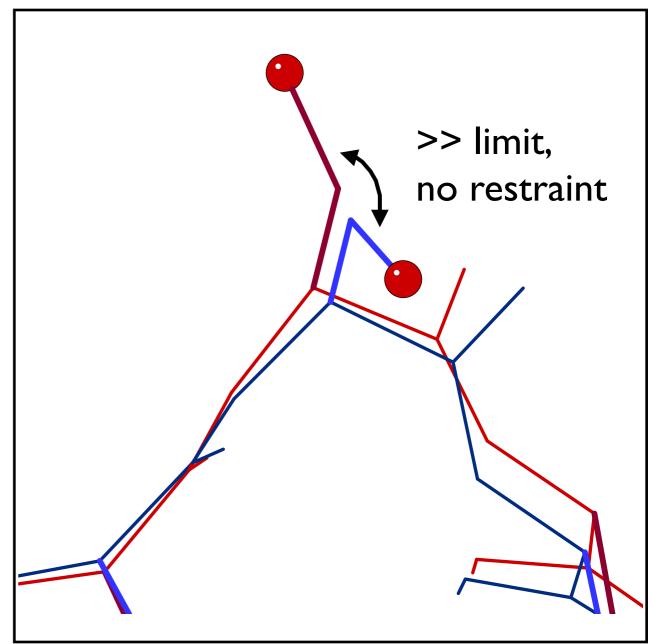
default: $limit = 15.0^{\circ}$



The 'limit' parameter

default: $limit = 15.0^{\circ}$

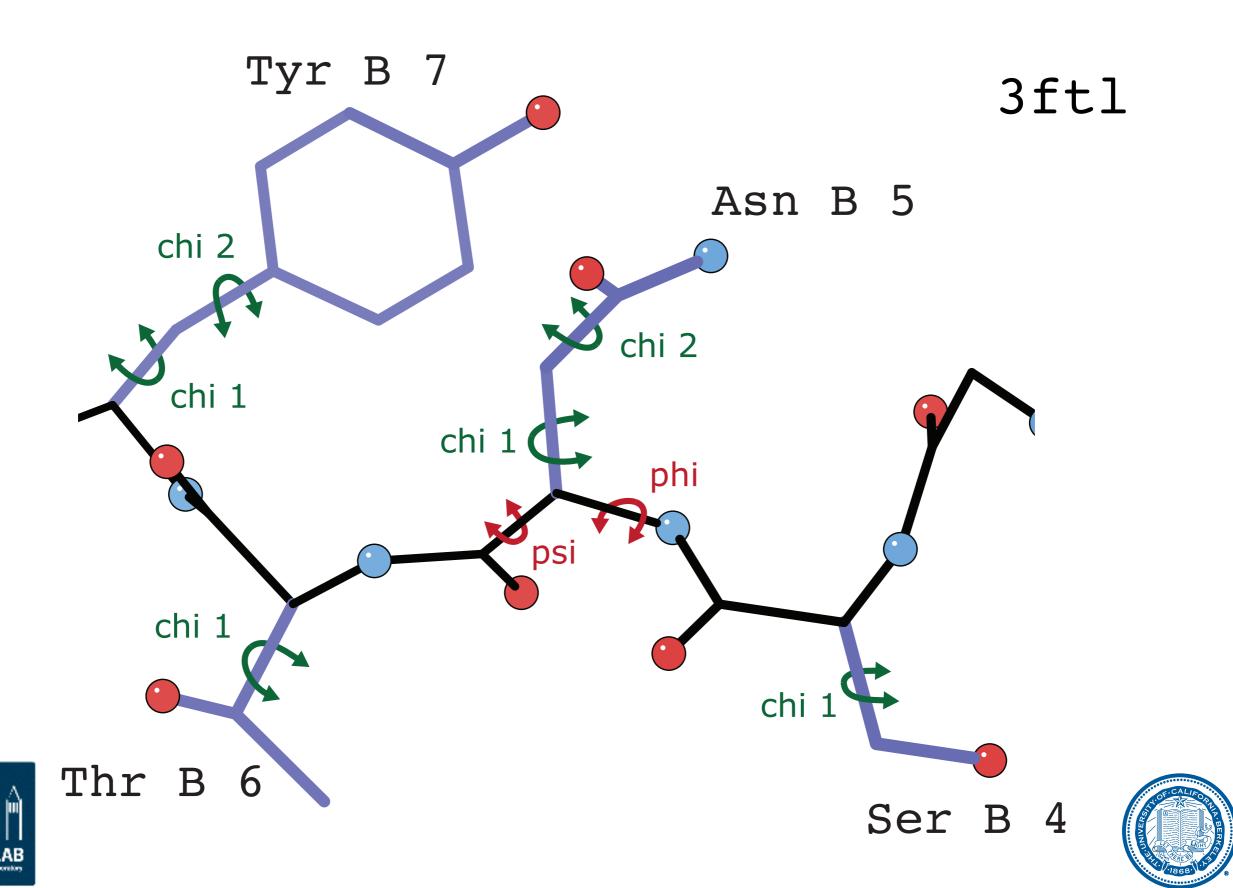








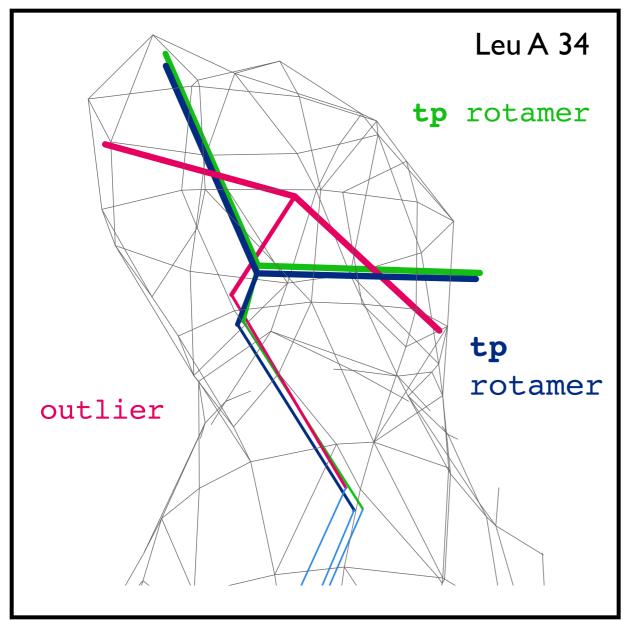
Why torsion angles?

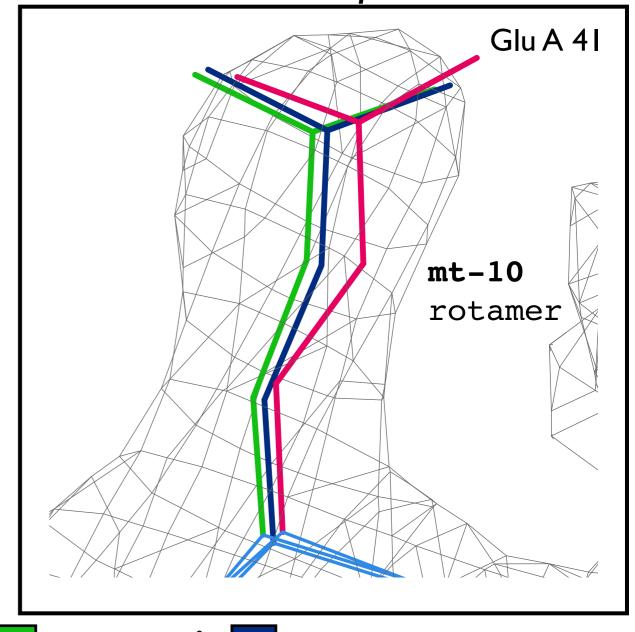


IGTX/IOHV reference example

outlier correction

restrained refinement





IGTX (3.0Å) IOHV (2.3Å) IGTX w/ IOHV reference

5 macrocycles of phenix.refine w/ reference restraints

 $R_{\text{free}}: 0.2379 \rightarrow 0.2186$

 $\Delta R: 0.833 \rightarrow 0.60$

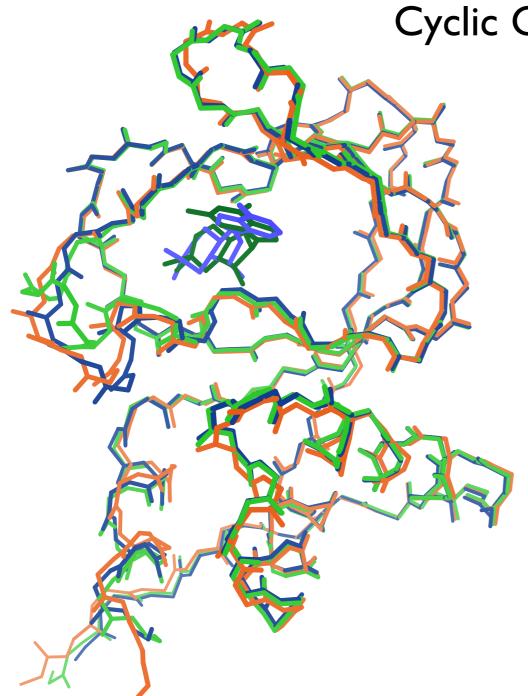
MolProbity: $64^{\text{th}} \rightarrow 96^{\text{th}}$





Practical Example

Cyclic GMP-dependent protein kinases (PKG's)



camp bound: 2.49Å

cGMP bound: 3.20Å

APO form: 2.69Å

JJ Kim et al. (2011) Crystal structures of PKG I β (92-227) with cGMP and cAMP reveal the molecular details of cyclic nucleotide binding. *PLoS ONE*.



Cyclic GMP-dependent protein kinase

cAMP bound: 2.49Å

cGMP bound: 3.20Å

APO form: 2.69Å

	Validation Criteria	cAMP bound	
All-Atom Contacts	Clashscore, all atoms:	16.53	
	Clashscore percentile	81st	
Protein Geometry	Poor rotamers:	2.61%	
	Rama outliers:	0.00%	
	Rama favored:	98.80%	
	Cβ dev. > 0.25Å:	0	
	MolProbity score:	2.04	
	MP score percentile	95th	
	Res w/ bad bonds:	0.00%	
	Res w/ bad angles:	0.00%	
Residual	R-work	0.1960	
	R-free	0.2264	

	1 1	
cGMP bound		cGMP bound
56.57		24.56
15th		87th
18.58%		4.00%
2.02%		0.40%
85.48%		96.00%
23		0
3.84		2.61
12th		96th
2.38%		0.00%
5.95%		1.18%
0.2102		0.1980
0.2582		0.2397
7 DO	· 	7 DO
APO		APO
28.52		19.5
46th		15th
10.53%		3.66%
3.19%		0.60%
89.02%		96.61%
3		0
3.29		2.43
12th		89th
0.79%		0.00%
0.98%		0.20%
0.2205		0.2166

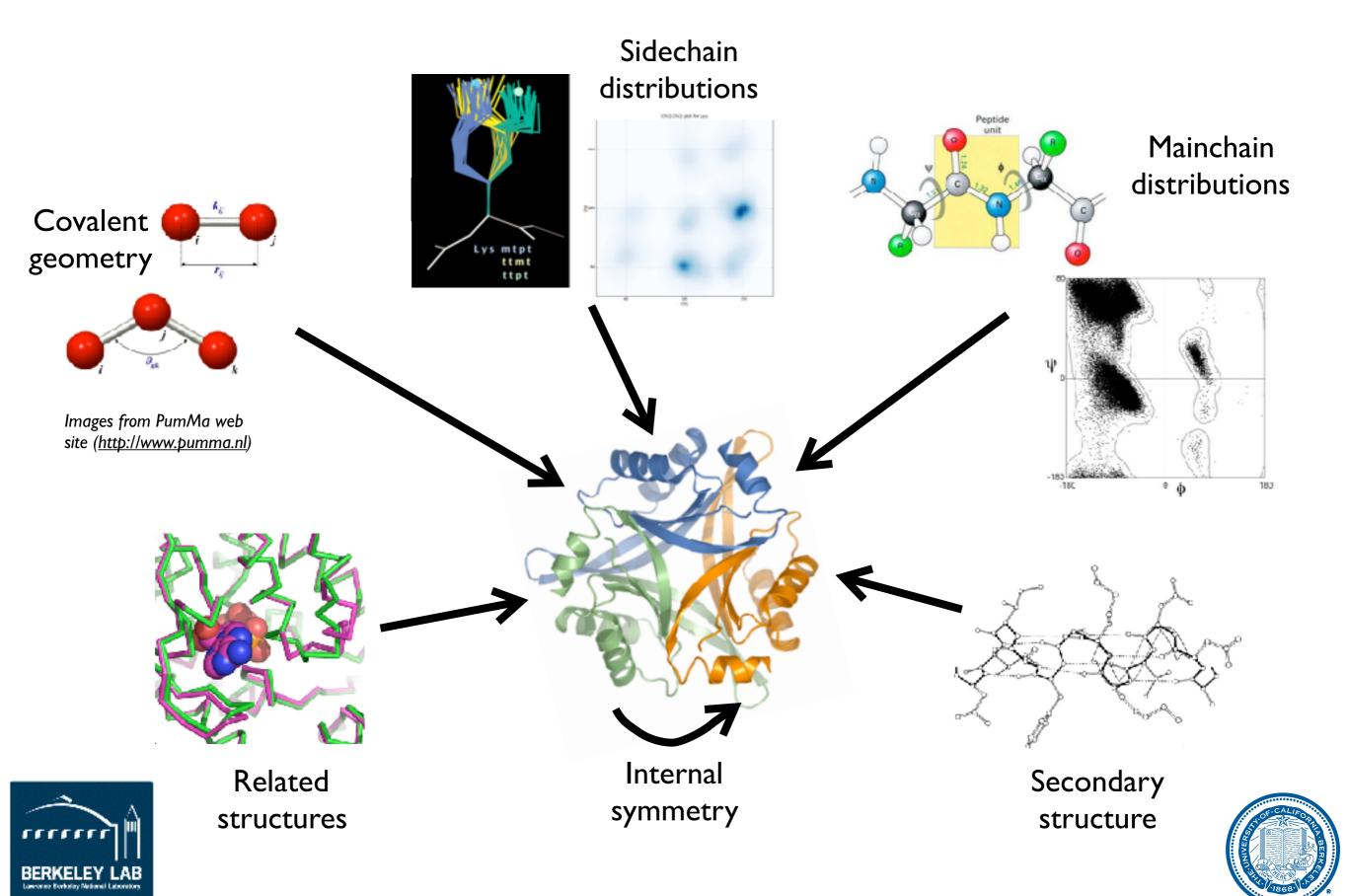
0.2612





0.2525

Sources of Prior Information



Torsion space NCS restraints (Jeff Headd)



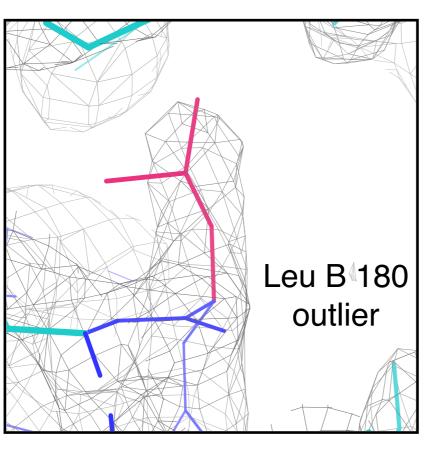


1b04: 2.8 Å

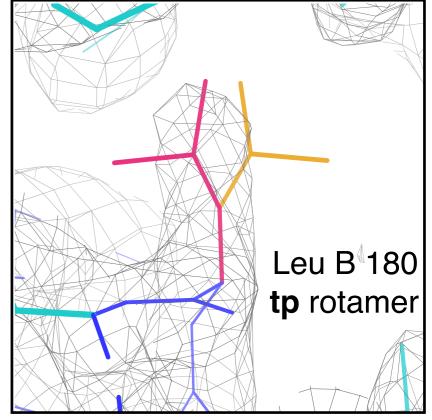
rotamer outlier correction

DNA ligase

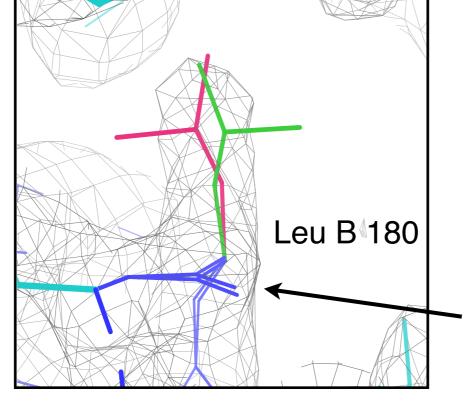
Identify rotamer outlier



2. correct to corresponding rotamer in NCS-related chain by matching χ angles

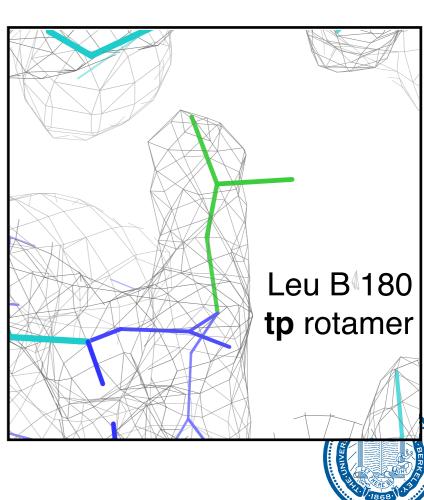


3. 'backrub' search, then limited χ angle torsion search



4. verify rotamer is still correct match

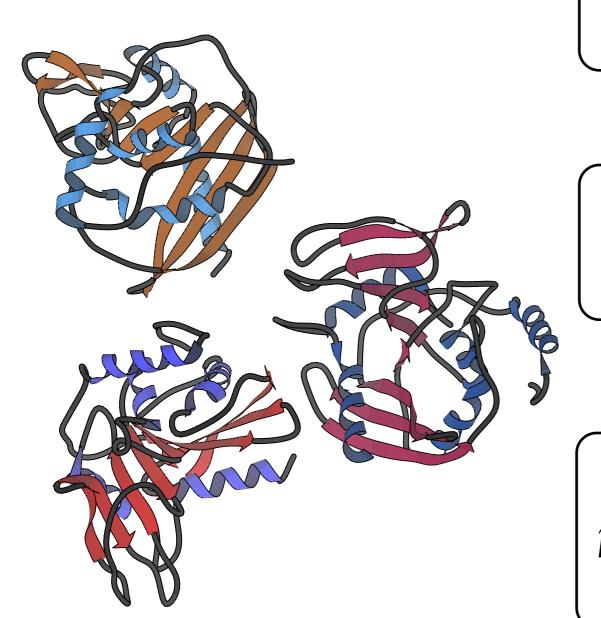
"backrub"



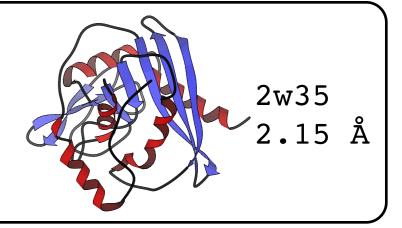


molecular replacement —— refinement

3hd0: 2.70 Å endonuclease



MR w/ Phaser



$$R_{work} = 0.3844$$

AutoBuild

- Rebuild in place
- NCS on for rebuilding
- NCS off for refinement
- No water picking

$$R_{\text{work}} = 0.1895$$

 $R_{\text{free}} = 0.2745$ $R_{\text{gap}} = 0.085$

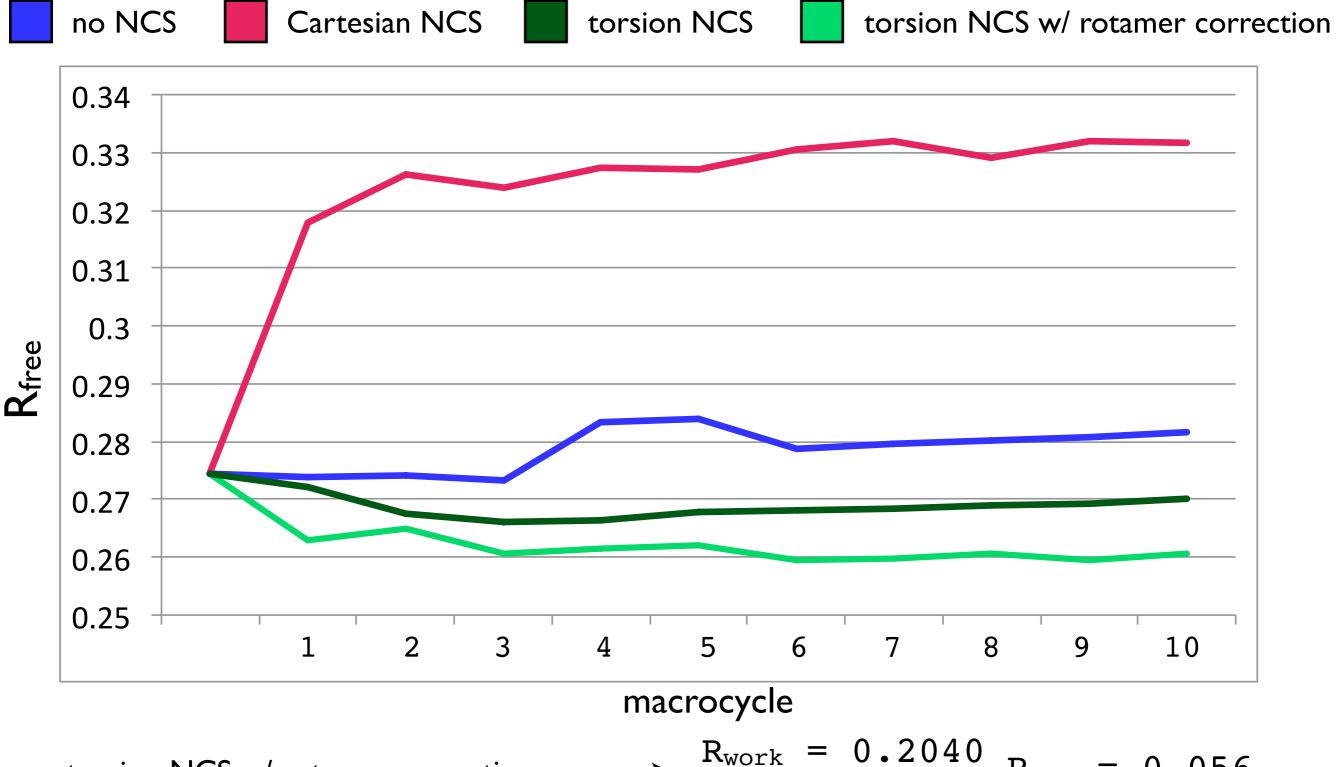
phenix.refine

- 10 macrocycles
- optimize weights
- No NCS, Cartesian NCS, torsion NCS w/ and w/o rotamer correction



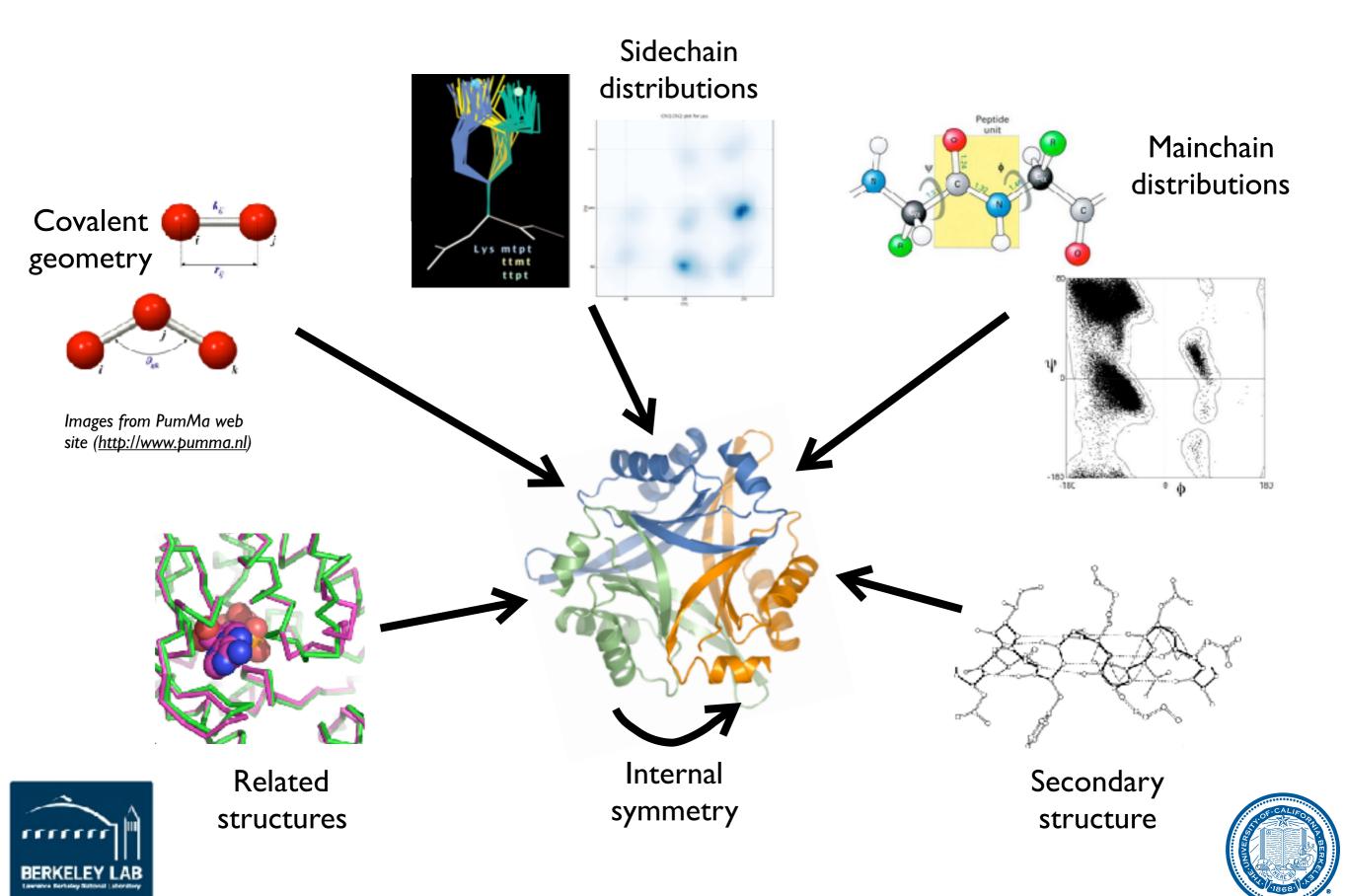


3hd0 refinement



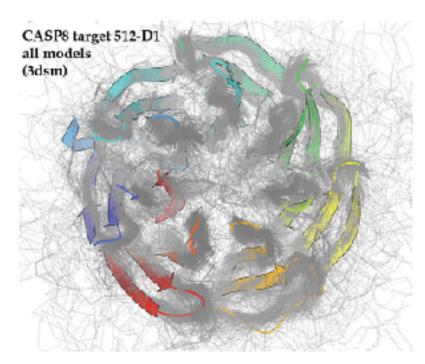
torsion NCS w/ rotamer correction \longrightarrow $R_{\text{mork}} = 0.2040$ $R_{\text{gap}} = 0.056$

Sources of Prior Information

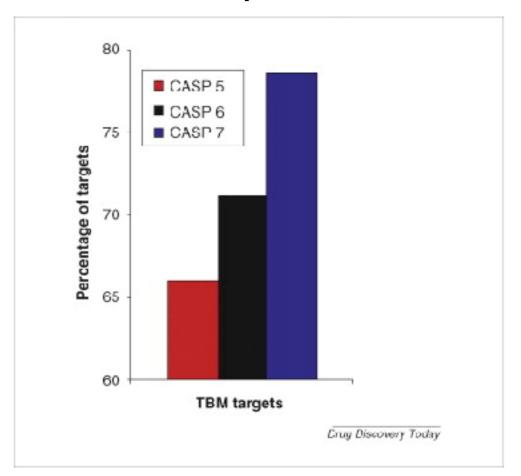


More Prior Information

- As the number of observations decreases we need to increase the amount of prior information we include (or the number of constraints we apply)
- At the extreme what if we had no data?
- Other fields have been trying to address this problem:
 - Structure prediction
 - Homology modelling
 - Protein folding



http://www.predictioncenter.org



From: Kryshtafovych & Fidelis, Drug Discovery Today, 2009, 14:386–393





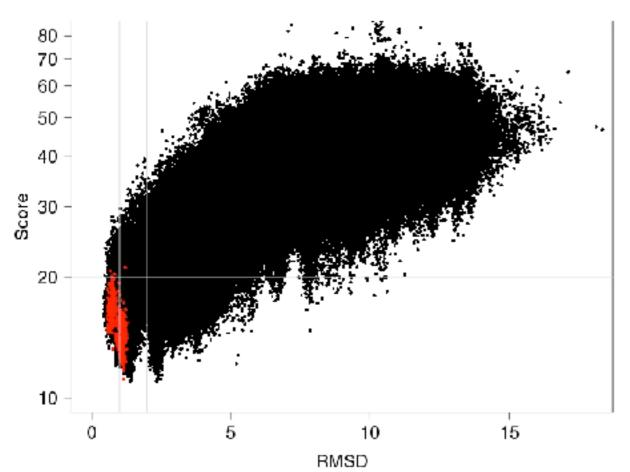
Physically Realistic Potentials (Rosetta) (Nat Echols & Frank DiMaio)



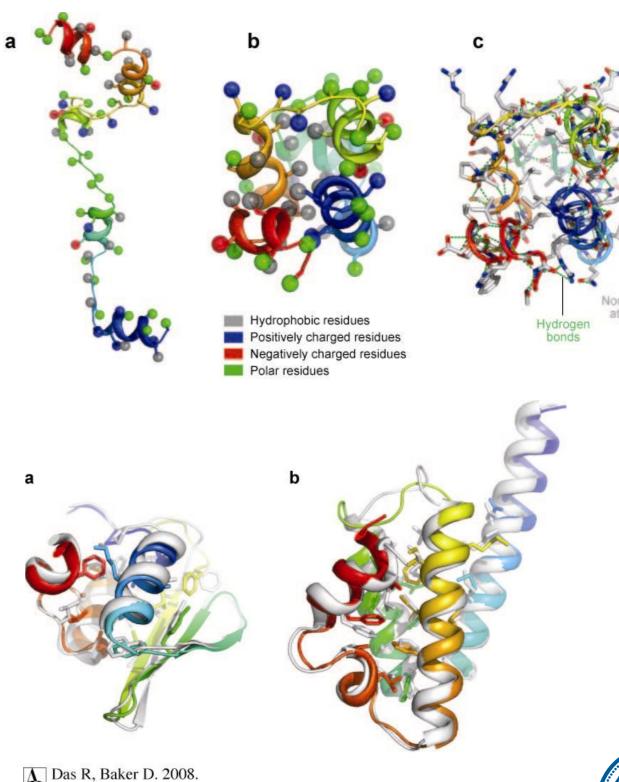


Rosetta

- ab initio model generation and model optimization
- Requires extensive computational sampling



Black - Rosetta ab initio models, Red - Crystal structure after Relax protocol



R Annu. Rev. Biochem. 77:363–82.



Why Rosetta

- Designed to recognize near-native structures among many possible models; combines empirical and physical potentials
 - All-atom force field, incorporates solvation effects, attractive forces, hydrogen bonds, knowledge-based dihedral restraints
- Can yield chemically realistic ab initio models without experimental data to guide assembly
 - Occasionally good enough for molecular replacement
- Shown to be useful for NMR structure determination with sparse data (CS-Rosetta), MR solution improvement (MR-Rosetta), RNA structure refinement (ERRASER)





Complementary Algorithms

Phenix

- Reciprocal space X-ray target functions (ML, MLHL, LS-twin)
- Bulk solvent correction
- B-factor refinement (including TLS)
- Map calculation
- Density modification (using RESOLVE)

Rosetta

- Physically realistic potentials
- Repacking to remove steric clashes and building rotameric sidechains
- Torsion-angle minimization
- Real-space target (refinement against electron density)
- Fragment-based rebuilding (optional, not currently used)





Low Resolution Protocol

- Sidechain repacking (using density)
- Coordinate
 refinement
 (reciprocal space
 torsion angle
 minimization and
 reduced nonbonded
 penalty)
- B-factor refinement

- Sidechain repacking (using density)
- Coordinate
 refinement (real
 space and
 reciprocal space
 torsion angle
 minimization)
- B-factor refinement

- Sidechain repacking (using density)
- Coordinate
 refinement
 (reciprocal space
 minimization with
 restrained bonds
 and angles)
- B-factor refinement

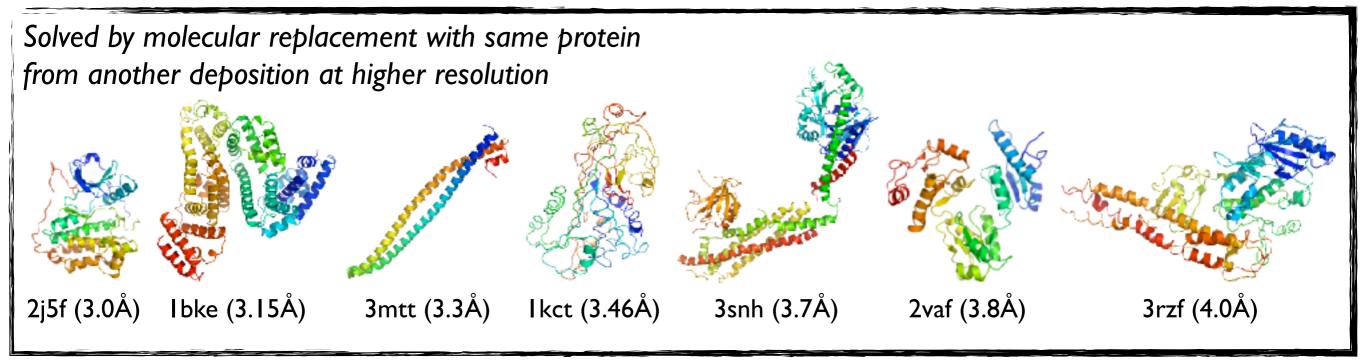
3 Cycles 5 Cycles 2 Cycles

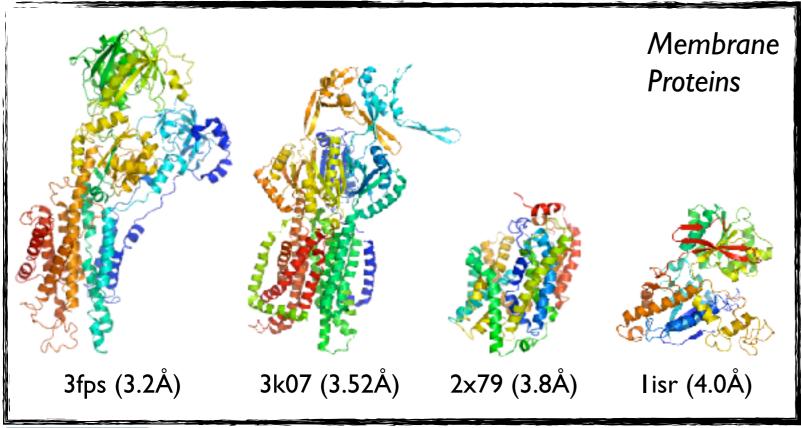


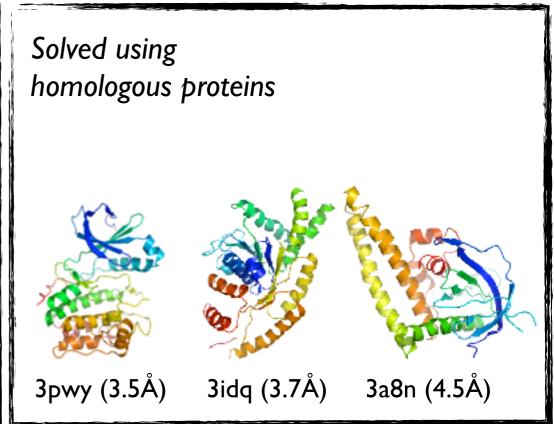
Protocol run 5 times in parallel and the best model selected based on R-free



Test Cases



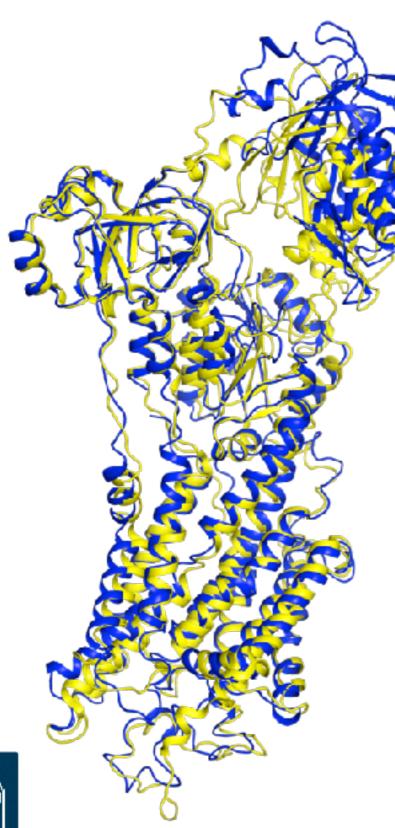


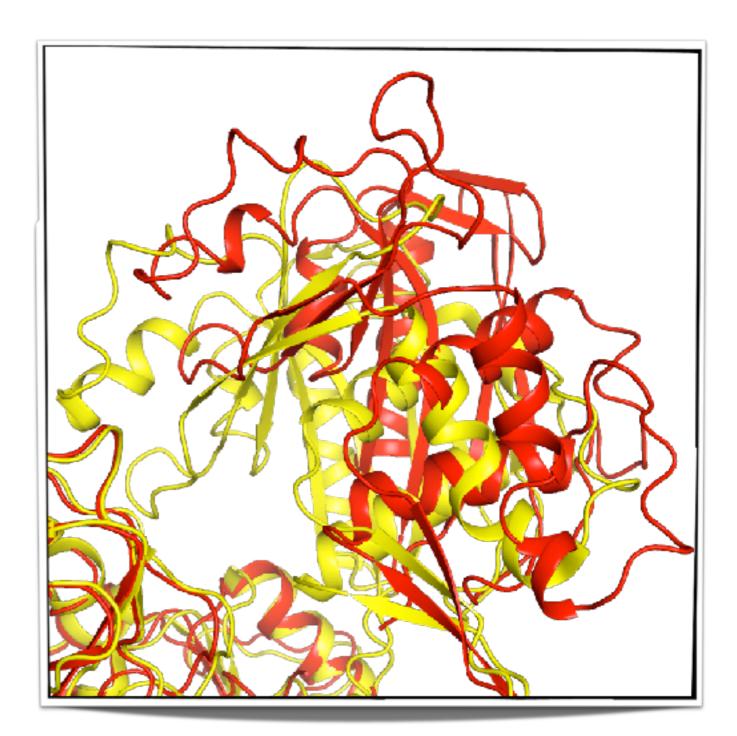






Calcium ATPase - phenix.refine



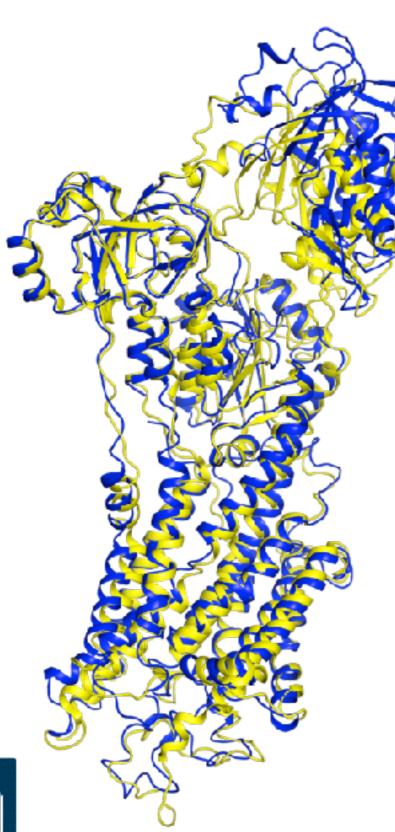


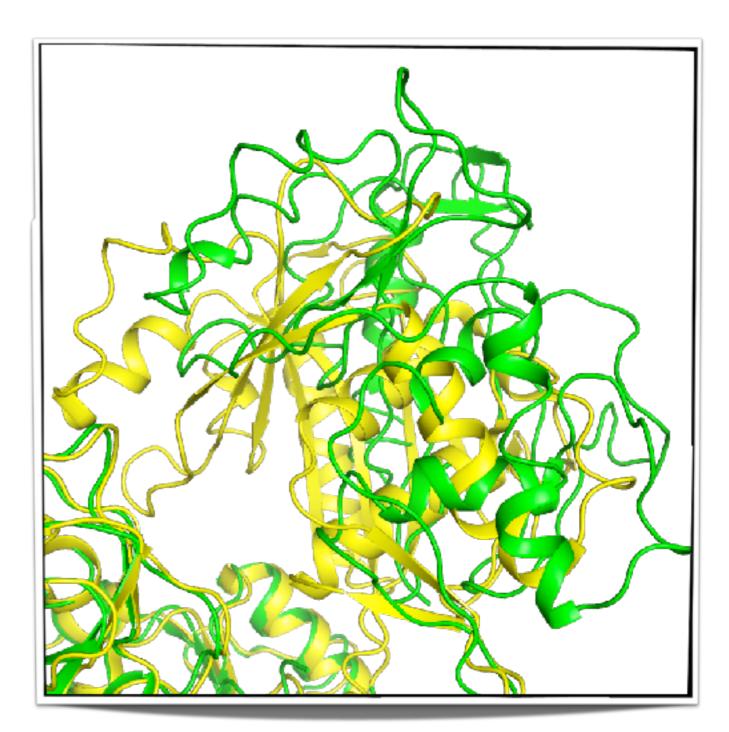
	R	R-free	mp score	RMSD
start	0.47	0.51	3.21	6.1
phenix	0.43	0.48	2.66	6.2





Calcium ATPase - DEN



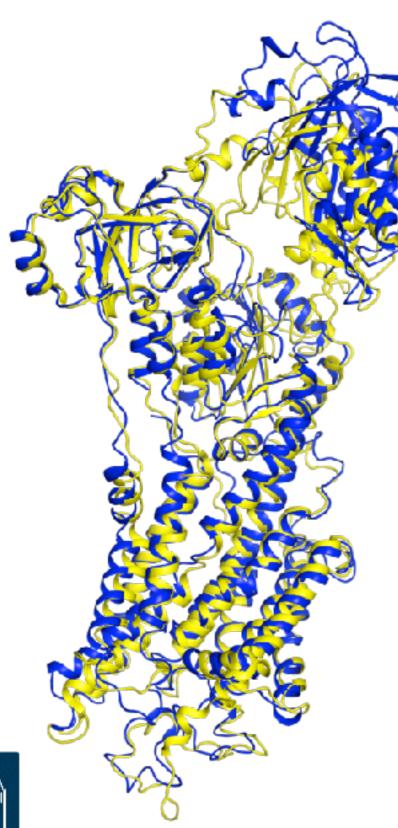


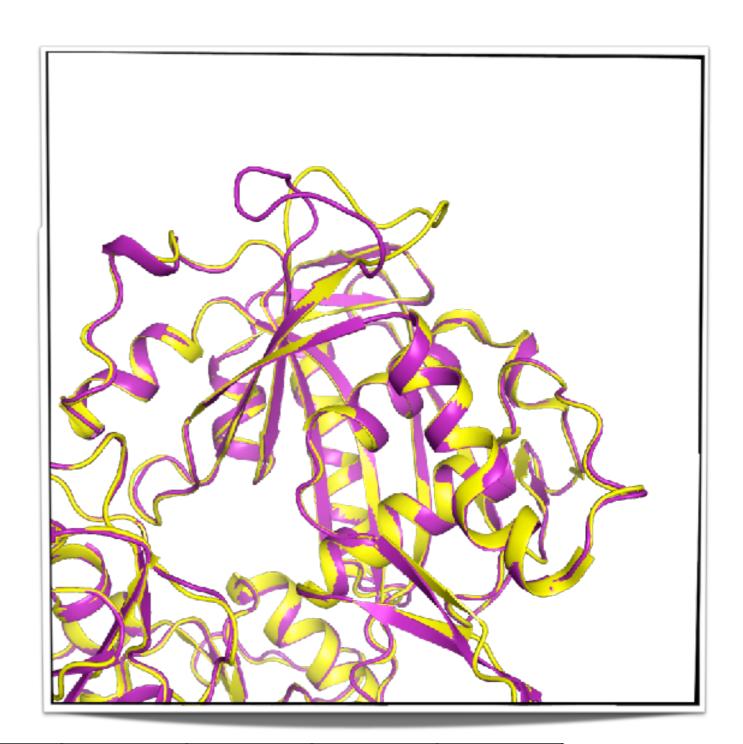
	R	R-free	mp score	RMSD
start	0.47	0.51	3.21	6.1
DEN	0.38	0.44	3.79	6.1





Calcium ATPase - Phenix-Rosetta



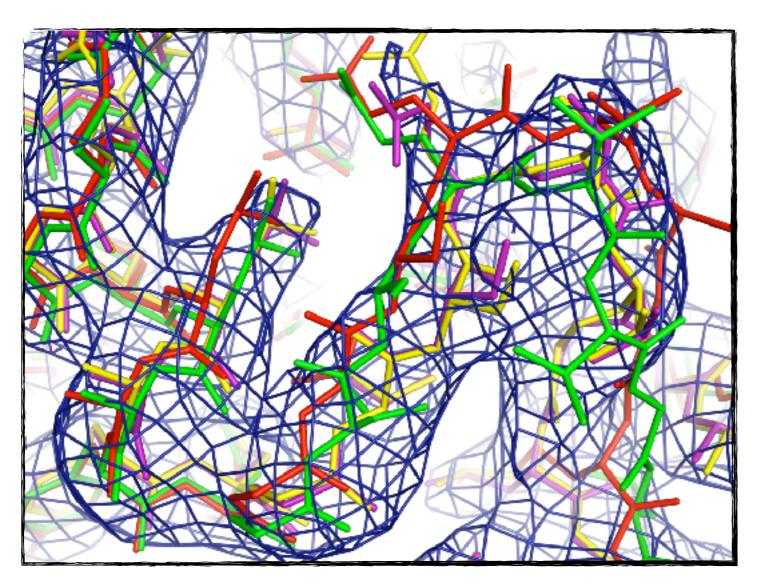


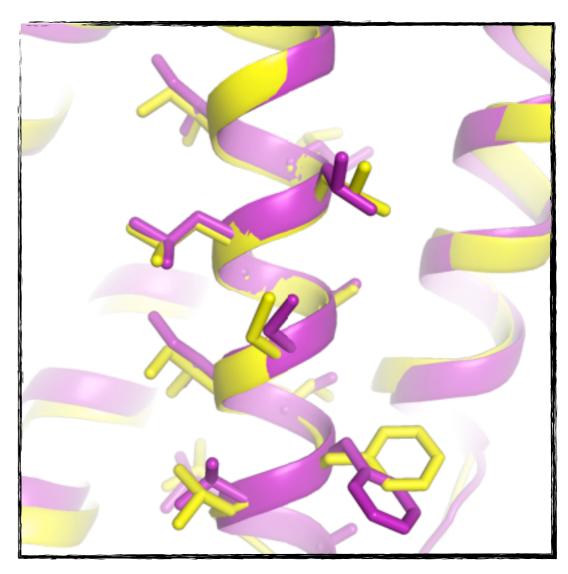
	R	R-free	mp score	RMSD
start	0.47	0.51	3.21	6.1
Rosetta	0.24	0.28	1.55	1.7





Calcium ATPase - Detail



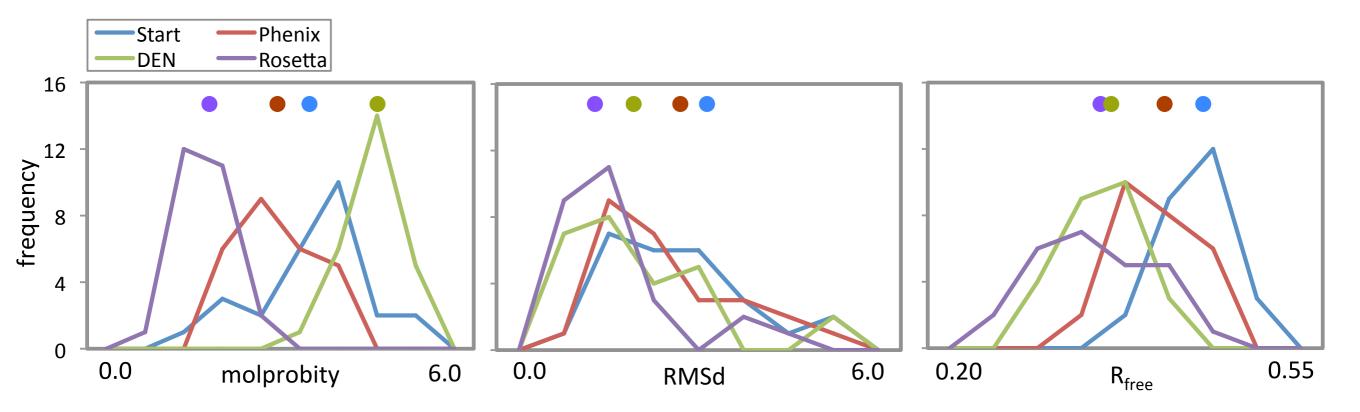


 Phenix-Rosetta model is very close to the deposited structure (even at the level of side chains) with better fit to density





Improved Models



- Phenix-Rosetta typically has improved fit to the crystallographic data and models are closer to the known structure
- Phenix-Rosetta always has improved model quality, as judged by Molprobity
- Generally similar to DEN results but with much improved geometry, and generally faster





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University of Washington

Frank DiMaio, David Baker

Oak Ridge National Laboratory

Marat Mustyakimov, Paul Langan

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- Garib Murshudov & Alexi Vagin
- Kevin Cowtan, Paul Emsley, Bernhard Lohkamp
- David Abrahams
- PHENIX Testers & Users: James Fraser, Herb Klei, Warren Delano, William Scott, Joel Bard, Bob Nolte, Frank von Delft, Scott Classen, Ben Eisenbraun, Phil Evans, Felix Frolow, Christine Gee, Miguel Ortiz-Lombardia, Blaine Mooers, Daniil Prigozhin, Miles Pufall, Edward Snell, Eugene Valkov, Erik Vogan, Andre White, and many more

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