





cryo-EM Course Laboratory for BioMolecular Structure (LBMS) Friday, June 6<sup>th</sup> 2025

# Predicted models for cryo-EM



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# Outline

- About AlphaFold predictions
- Using predicted models in cryo-EM
  - 1. Docking
  - 2. Model completion
  - 3. Reference model restraints
- Automated workflow: Iterating prediction and model building

### Predicting models with AlphaFold



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Sequence Multiple sequence alignment

**3D** prediction

### 1. pLDDt (predicted Local Distance Difference Test)

- pLDDt identifies where errors are more likely.
- Per-residue confidence measure.
- Scales from 0 100 (pLDDt > 90: predicted with high accuracy).



Data from 7mjs, Cater, R.J., et al. (2021). Nature 595, 315–319

7*mjs* (3 Å, EMDB 23883)

Residues 100-120

Low sequence coverage, low confidence, low accuracy

7mjs

**AlphaFold** 

Residues 1-100 High sequence coverage and confidence

Data from 7mjs, Cater, R.J., et al. (2021). Nature 595, 315–319

### 1. pLDDt (predicted Local Distance Difference Test)

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AlphaFold confidence (pLDDT)	Median prediction error (Å)	Percentage with error over 2 Å
>90	0.6	10
80 - 90	1.1	22
70 - 80	1.5	33
<70	3.5	77

Terwilliger et al. (2023), AlphaFold predictions are valuable hypotheses, and accelerate but do not replace experimental structure determination. Nature Methods 2023: https://doi.org/10.1038/s41592-023-02087-4

### 1. pLDDt (predicted Local Distance Difference Test)



(PDB entry 6L5L)

Blue: pLDDt > 90 Green: pLDDt 80 - 90





### 2. Predicted aligned error (PAE)

- Certainty of relative positions between two residues.
- Identifies accurately-predicted domains.
- Dark blue: uncertainty in relative positions < 5 Å.</li>



### Using predicted models: B-factors



high pLDDT (high confidence) low pLDDT (low confidence, uncertain)



high B-factor (disordered, uncertain) low B-factor (ordered)

Oeffner RD, Croll TI, Millán C, Poon BK, Schlicksup CJ, Read RJ, Terwilliger TC. Acta Cryst. D, 2022 (78):1303-1314; <a href="https://doi.org/10.1107/S2059798322010026">https://doi.org/10.1107/S2059798322010026</a>

### Using predicted models: B-factors

high pLDDT (high confidence) low pLDDT (low confidence, uncertain)



high B-factor (disordered, uncertain) low B-factor (ordered)

B-factor may be used in downstream calculations, e.g. to calculate weights for docking. Residues with high B-factors are downweighed.

 $\rightarrow$  Convert pLDDT to pseudo B-factors.

$$\Delta = 1.5 \exp[4(0.7 - \text{pLDDT})]$$
  $B = \frac{8\pi^2 \Delta^2}{3}$ 

Oeffner RD, Croll TI, Millán C, Poon BK, Schlicksup CJ, Read RJ, Terwilliger TC. Acta Cryst. D, 2022 (78):1303-1314; <a href="https://doi.org/10.1107/S2059798322010026">https://doi.org/10.1107/S2059798322010026</a>

### AlphaFold predictions are great hypotheses

# AlphaFold models can be....



Terwilliger et al. (2023), AlphaFold predictions are valuable hypotheses, and accelerate but do not replace experimental structure determination. Nature Methods 2023: https://doi.org/10.1038/s41592-023-02087-4

### How to use predictions?

Incorporate predictions into the typical cryo-EM workflow.



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Incorporate predictions into the typical cryo-EM workflow.



### 1. Use a predicted model for cryo-EM docking



# cryo-EM docking

Cryo-EM maps typically lack the necessary resolution and quality for *ab initio* model building.

 $\rightarrow$  dock a pre-existing model into the map.



Model assumed to look like the sample.

## Use a predicted model for cryo-EM docking

### Example:

Cryo-EM map (30160 – 7brm) 3.6 Å

>chain ' A' XXXXXXXXXXXXXXCLTAPPKEAARPTLMPRAQSYKDLTHLPAP TGKIFVSVYNIQDETGQFKPYPASNFSTAVPQSATAMLVTALKDS RWFIPLERQGLQNLLNERKIIRAAQENGTVAINNRIPLQSLTAAN IMVEGSIIGYESNVKSGGVGARYFGIGADTQYQLDQIAVNLRVVN VSTGEILSSVNTSKTILSYEVQAGVFRFIDYQRLLEGEVGYTSNE PVMLCLMSAIETGVIFLINDGIDRGLWDLQNKAERQNDILVKYRH MS



### cryo-EM map

#### sequence

# Get a prediction



### **Process prediction**



AlphaFold model

### **Process prediction**



### Dock processed model



## Docking in Phenix

• Dock-in-map (T. Terwilliger) – phenix.dock\_in\_map





# Docking in Phenix

### Likelihood-based EM docking:

- Use likelihood scores to dock a model into a map
- Works at low resolution (8.5 Å)

$$\begin{aligned} \text{LLG}(\mathbf{E}_{\text{mean}};\mathbf{E}_{\text{C}}) &= \frac{2}{1 - D_{\text{obs}}^2 \sigma_{\text{A}}^2} D_{\text{obs}} \sigma_{\text{A}} E_{\text{mean}} E_{\text{C}} \cos(\Delta \varphi) \\ &- \frac{D_{\text{obs}}^2 \sigma_{\text{A}}^2 (E_{\text{mean}}^2 + E_{\text{C}}^2)}{1 - D_{\text{obs}}^2 \sigma_{\text{A}}^2} - \ln(1 - D_{\text{obs}}^2 \sigma_{\text{A}}^2). \end{aligned}$$





Read RJ, Millán C, McCoy AJ, Terwilliger TC. Likelihood-based signal and noise analysis for docking of models into cryo-EM maps. Acta Cryst. D 2023 271–80. Millán C, McCoy AJ, Terwilliger TC, Read RJ. Likelihood-based docking of models into cryo-EM maps. Acta Cryst. D 2023 Apr 1;79(Pt 4):281–9.

# Docking in Phenix/ChimeraX

- Likelihood-based docking can be done via ChimeraX.
- Can select the region into which the model should be docked.



## Docking in Phenix/ChimeraX

- Likelihood-based docking can be done via ChimeraX.
- Can select the region into which the model should be docked.



### Dock processed model



Some parts don't fit into the map

### Dock processed model



Some parts don't fit into the map  $\rightarrow$  fit loops and rebuild



# Fit loops and rebuild



Next step: real space refinement

#### 2. Use a predicted model to complete your structure



Use predicted model as hypothesis for missing parts.

### Can AF predictions help if the structure is already solved?

Repressor-DNA complex, solved with 2.6 Å SeMet SAD data & refined against 3.1 Å native data

Before AlphaFold, R/Rfree = 0.27/0.29

AlphaFold model: A **hypothesis** about this structure

After AlphaFold, R/Rfree = 0.21/0.24 (it was a good hypothesis)

Jamie Wallen, Western Carolina University

### **3.** Use an Alphafold model for reference model restraints

### Restraints: *a priori* knowledge

- Restraints increase the number of observations.
- Restraints modify the target function by creating relationships between independent parameters.
- Example: restrained bond lengths



- the coordinates of the two atoms are independent
- restraint keeps their distance within a certain target value
- imposes a penalty if it deviates too much.

### **Reference model Restraints**

#### Concept

- Use a related model to generate a set of torsion restraints.
- Restrain each torsion angle in the working model to the corresponding torsion angle in the reference model.
- Allows for structural differences.

#### When to use

Low resolution (worse than 3Å).

If no high resolution homologue available, could use AF model for reference model (AF models have good geometry).

### **Reference model Restraints**



### Reference model Restraints: example

**1GTX**: 3.0 Å **10HV**: 2.3 Å



4-aminobutyrate-aminotransferase

### Reference model Restraints: example



4-aminobutyrate-aminotransferase

### **Reference model Restraints**

#### How to use

Supply a reference model in phenix.refine; check the corresponding box. (Oleg Sobolev: working on finding reference automatically)

Real-space refinement (Project: 7rpq_AF_reference_m Preferences Help	Run Abort Save Help
Input/Output Refinement Settings	$\triangleleft \triangleright$
By default real-space refinement applies Ramachandran restraints to the model to maintain good stereochem input model contains Ramachandran outliers these restraints may lead to a non-optimal local geometry, whic validation metrics to detect, such as CABLAM and the Rama-Z score. We recommend that you manually fix th real-space refinement. Job title :	nistry. However, if your ch will require other nese outliers before running Input/Output Refinement Settings Strategy
Input       Format       Data type         File path       Format       Data type         \u03c4       /Users/dcliebschner/Documents/7rpq_AF_reference       PDB       Model file         \u03c4       /Users/dcliebschner/Documents/7rpq_AF_reference       PDB       Reference model         \u03c4       /Users/dcliebschner/Documents/7rpq_AF_reference       ccp4_mtz       M12 mic	Image: Select Atoms Image: Select Atoms     Image: Select Ato
Add file Remove file Modify file data type   Resolution : Map coefficients label :   Ignore symmetry conflicts	Strategy Options Morphing : first  Options  Reference model restraints Options Other Options
Output          Output       Prefix :         Vrite initial geo file       Write final geo file         Write all states       Run validation	Scattering table :       electron image: Weight :       Resolution factor :       0.25         Nproc :       1       Random seed :       0         Image: Refine ncs operators       Image: Show per residue         Model interpretation       Rotamers       Automatic linking       All parameters

#### Use your working model to get a new AlphaFold prediction



# Example: Fab heavy chain



### 7MJS

Single-Particle Cryo-EM Structure of Major Facilitator Superfamily Domain containing 2A in complex with LPC-18:3

PDB DOI: https://doi.org/10.2210/pdb7MJS/pdb EM Map EMD-23883: EMDB EMDataResource

### 3.03 Å resolution



A loop that interacts with other chains is not correctly predicted.

### AF2 prediction of chain H



### AF2 prediction of chain H



### **Process AF2 prediction**



### Dock processed model into the map



### AF2 prediction of chain H

Loop predicted with low confidence Does not fit into map.

### AF2 prediction of chain H



### Dock and rebuild model



"predicted-processed-docked-rebuilt" model

### Make a new prediction



### Using a template improves prediction



### Iterate prediction and rebuilding



### Iterate prediction and rebuilding





Data from 7mjs, Cater, R.J., et al. (2021). Nature 595, 315–319

## Using predicted models

### Updated approach: Iterate prediction and model building





# Phenix Server for running AlphaFold

put/Output Predictio	on Settings						4				
rediction strategy											
Number of models :	5	✓ Include templates from PDB	🗸 Use MSA	Skip all M	SA after first cycl	le					
rediction Server											
Prediction Server :	PhenixServer 🔇	Allow precalculated results	✓ Allow precalculate	d MSA 🛛 🗸	Stop if internet	not available					
Additional prediction inputs (Note: this GUI is only for prediction, not docking or model-building)											
Predict and Build	Input Files	Prediction Control A	All parameters								

# No need for a local AlphaFold installation

SERVER STATUS: UP RUNNING JOBS: 2 WAITING JOBS: 0	
RUNNING JOBS: 2 WAITING JOBS: 0	:
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This server is used in PredictModel and PredictAndBu Colab can be used as an alternative	ild.
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# Process predicted model

Process Predicted Model (Project: 7rpq_AF_reference_model)	Preferences Prefe	Run Abort	C Help
Configure			$\triangleleft \triangleright \mathbf{x}$
ProcessPredictedModel: Prepare predicted model for structure determination Replace values in B-factor field with estimated B values. Optionally remove low-confidence residues and split into domains. Inputs: Model file (PDB, mmCIF)			
Job title :			
Predict	ed model		
Predicted model :	Browse		
Contents of B-value field for input models : plddt 📀			
Optional input files			
Output			
Output file prefix (optional) :			
Options			
✓ Remove low-confidence residues ✓ Split model into compact regions Maximum output B :	999.0		
✓ Remove hydrogen Use only single-letter chain ID Maintain continu	uous chain		
Processing options All parameters			



# Iterate with Predict and Build

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Input/Output	Prediction ar	d Building Sett	ngs PredictAndBu	uildCryoEM_17						4 ▷
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Carry on pre	vious run (Resto	re a completed r	un from Job history	to fill this in automatic	ally. You can also selec	t the CarryOn s	ubdirectory i	n a Predict_	_and_Build_xx	directory)
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Advanced in	puts									
Predict a	nd Build	Input Files	Box info	Processing	Search	Building				
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Fully automatic – AF prediction, processing, building, refinement.

# Strategy for structure determination

### **1. Predict your structure**

Design your experiment accordingly

(choose experimental approach, consider trimming at domain boundaries)

### **2. Solve your structure**

Cryo-EM: docking X-ray: MR; SAD

### 3. Update your prediction

Run AlphaFold again with your best model as a template

#### 4. Improve your structure

Use your new prediction as hypothesis, rebuild parts

Iterate

### Summary

- AlphaFold models are great hypotheses.
- Can be used for cryo-EM docking (need to interpret the confidence measures), model completion, reference model restraints.
- Iterating prediction and model building can lead to improved models.
- Still need experiment to get a model that best explains the data.



# Further reading/material



#### Tom Terwilliger: AlphaFold changes everything

https://youtu.be/ugMPYdPo8Bc?feature=shared





Liebschner D, *et al.*, Macromolecular structure determination using X-rays, neutrons and electrons: recent developments in *Phenix*. Acta Cryst. 2019 **D75**:861–877