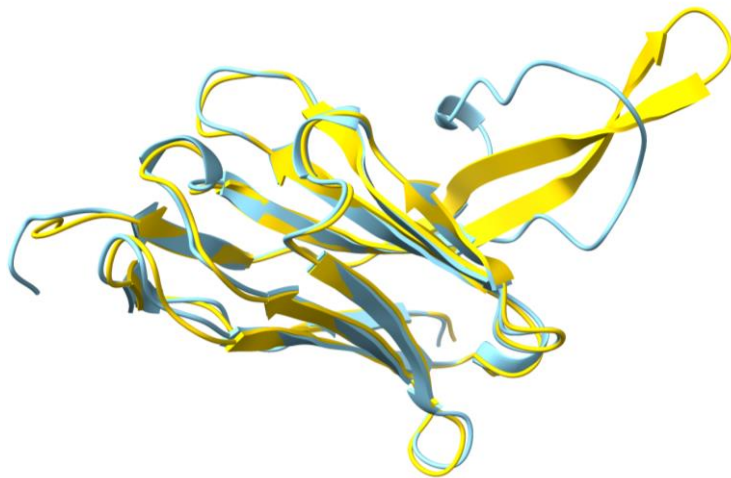


Using AlphaFold predictions for structure determination

*Phenix Workshop, AsCA 2025
Taipei, Taiwan, Dec. 6, 2025*

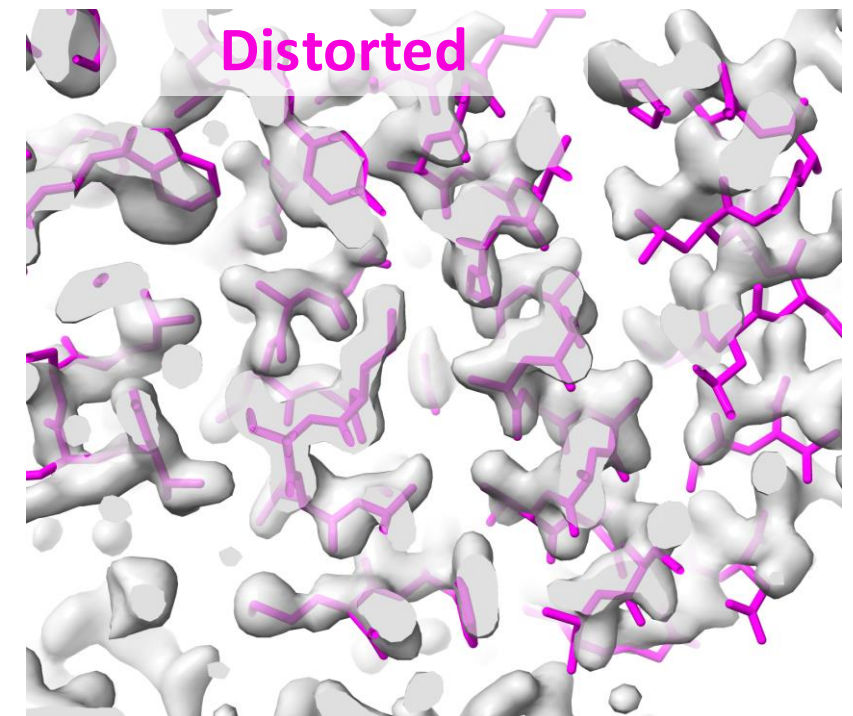
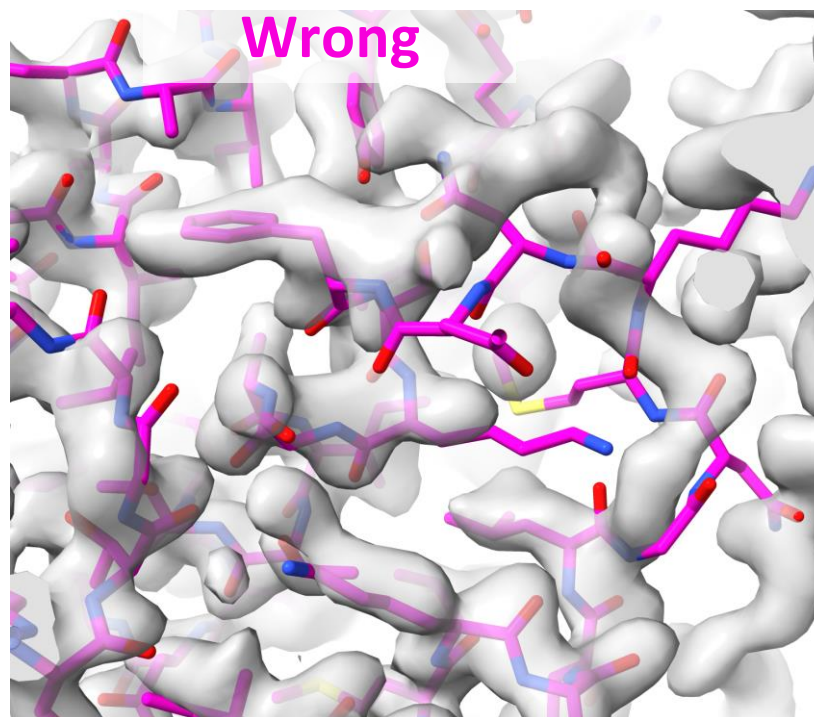
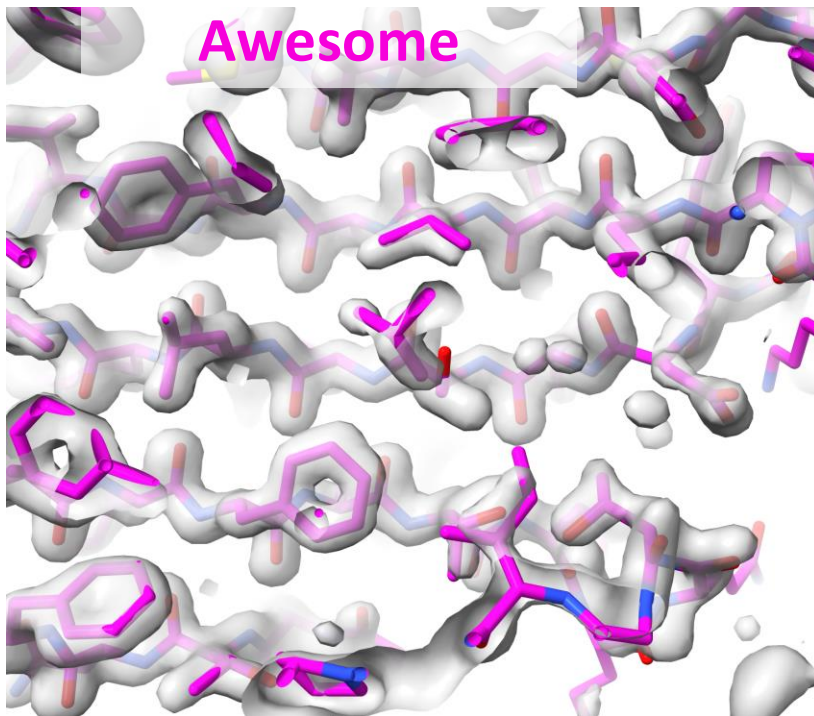
Tom Terwilliger

The New Mexico Consortium
Los Alamos National Laboratory



AlphaFold predictions are great hypotheses

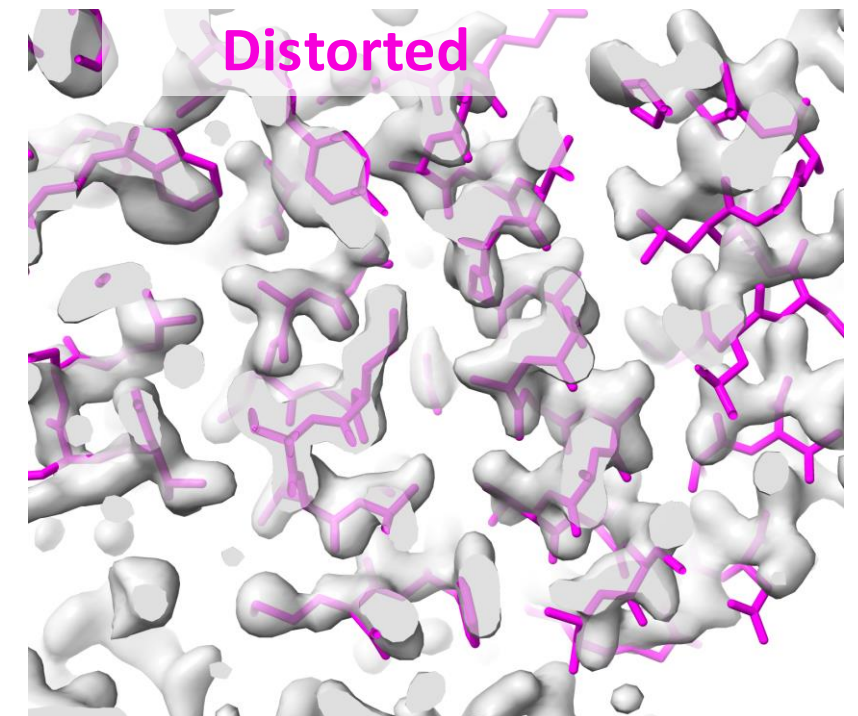
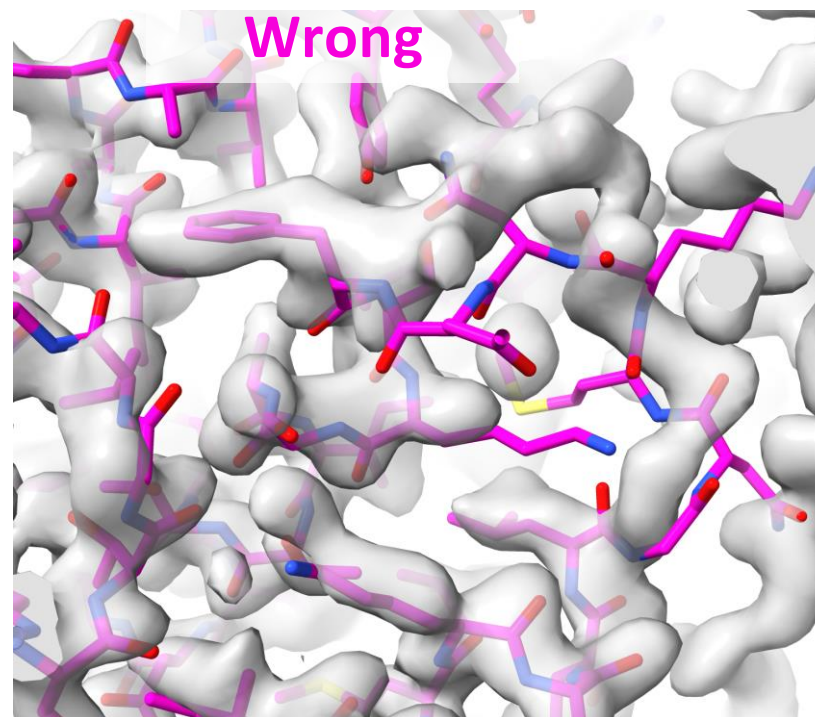
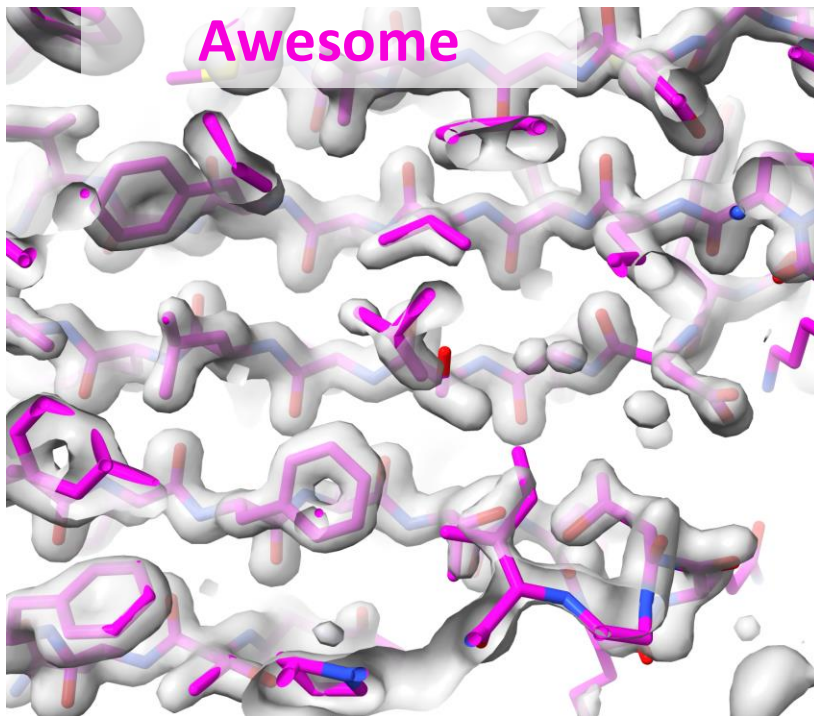
*AlphaFold models
can be*



AlphaFold predictions and confidence estimates

Residue-specific confidence (pLDDT) identifies where errors are more likely

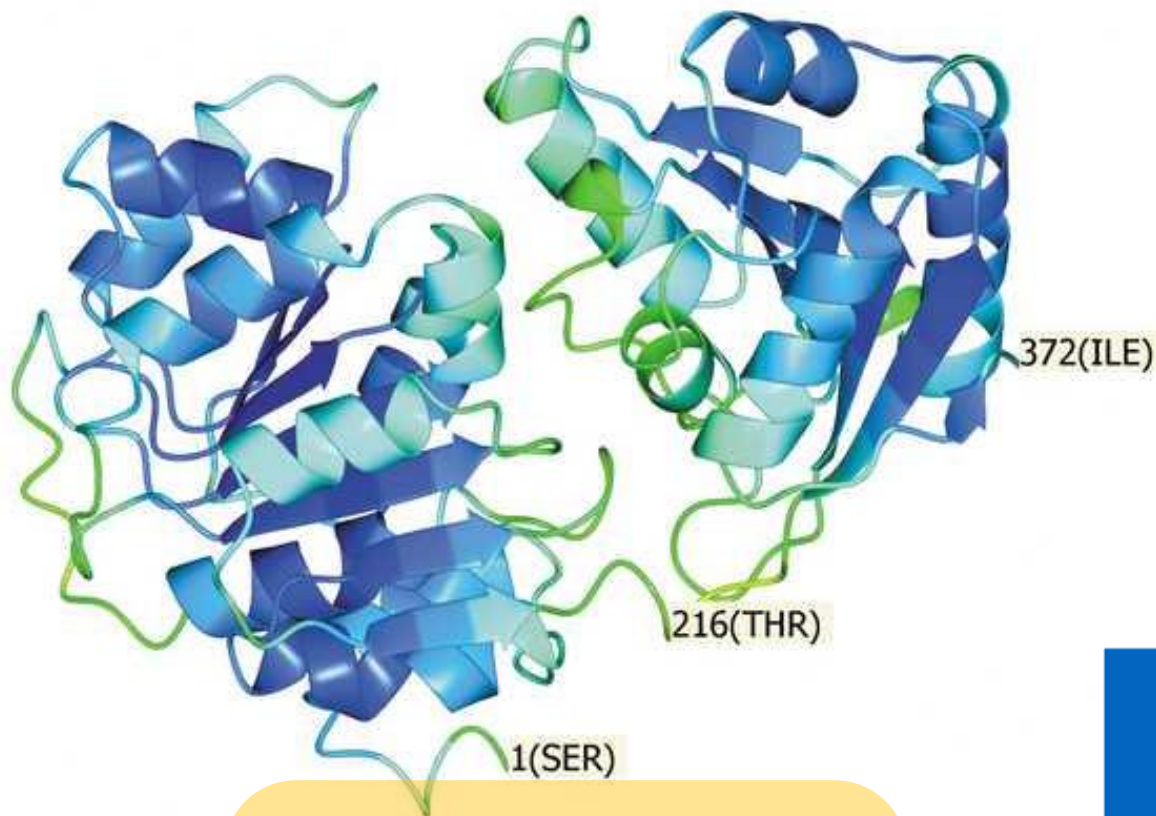
AlphaFold confidence (pLDDT)	Median prediction error (Å)	Percentage with error over 2 Å
>90	0.6	10
80 - 90	1.1	22
70 - 80	1.5	33
<70	3.5	77



Terwilliger et al. (2024), AlphaFold predictions are valuable hypotheses, and accelerate but do not replace experimental structure determination. Nature Methods 21, 110-116.

AlphaFold confidence measure

(pLDDT, Predicted difference distance test)



Confidence:

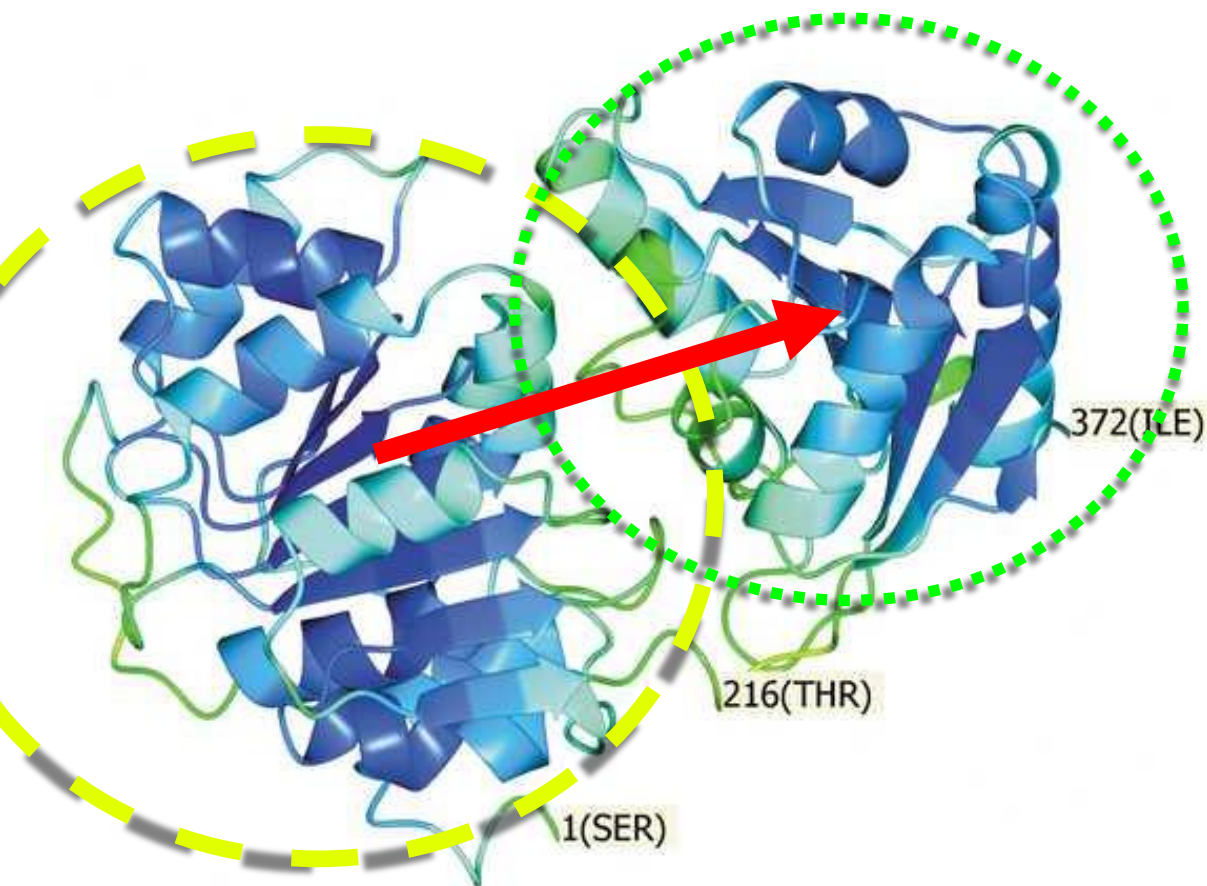
Blue: > 90

Green: 80 - 90

AlphaFold prediction for
RNA helicase
(PDB entry 6i5i)

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

PAE matrix (Predicted aligned error)



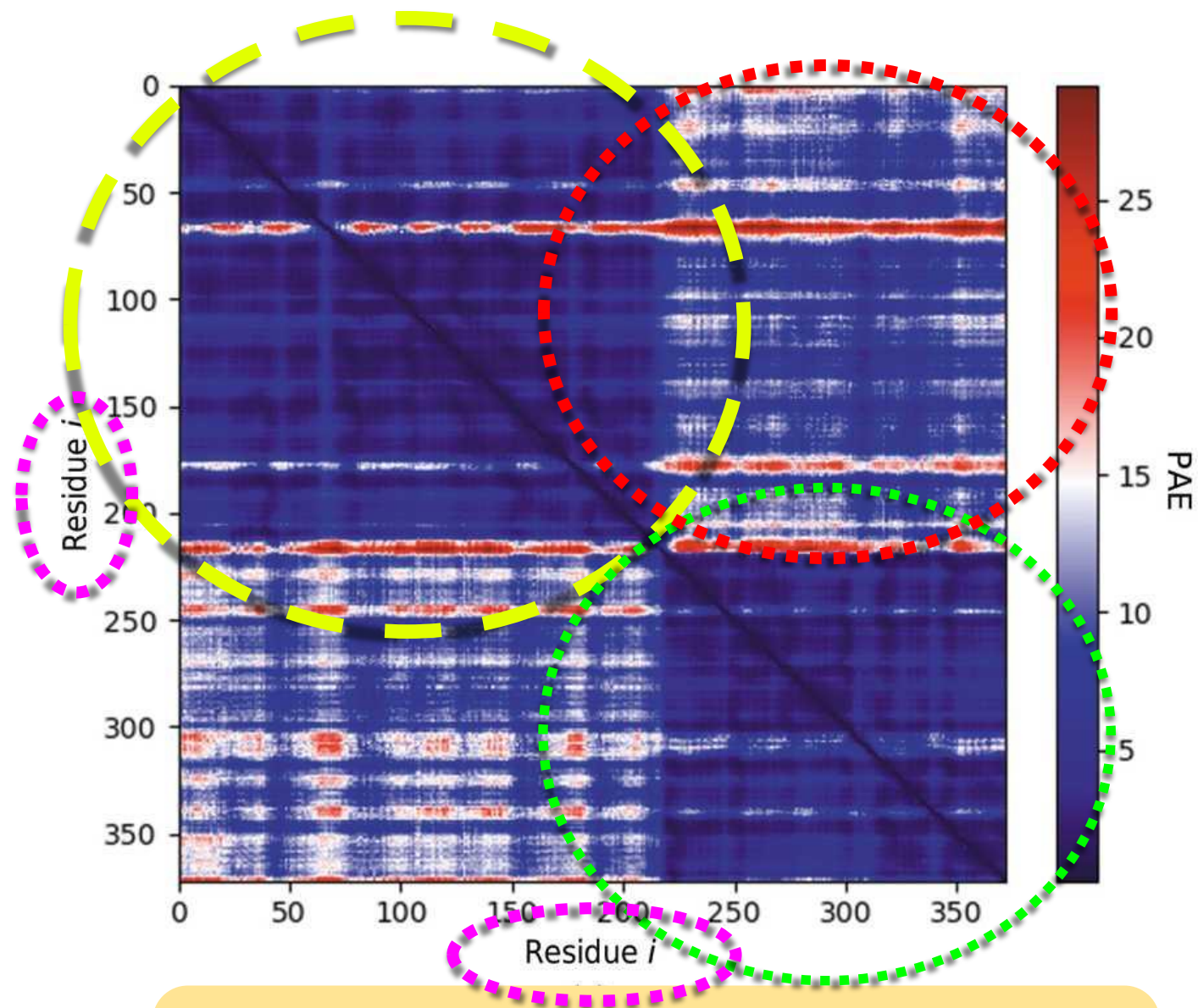
Confidence:

Blue: > 90

Green: 80 - 90

*AlphaFold prediction for
RNA helicase
(PDB entry 6i5i)*

*PAE matrix identifies
accurately-predicted domains*



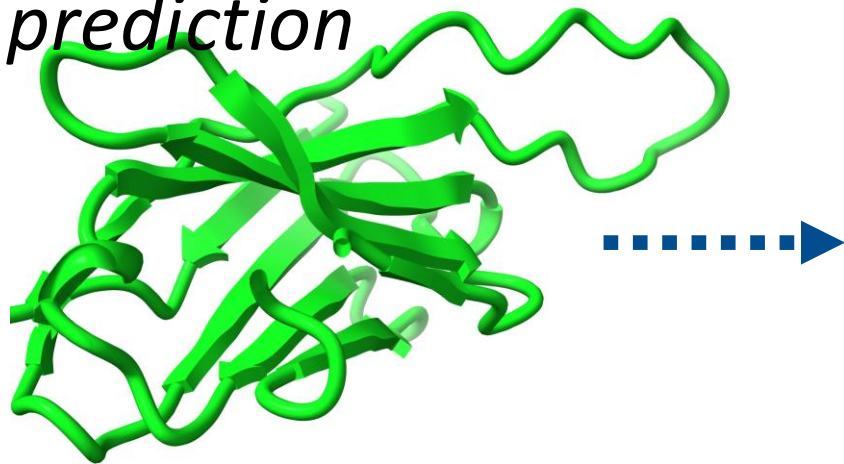
*Dark blue: uncertainty in
relative positions < 5 Å*

Using your best model as a template in AlphaFold prediction

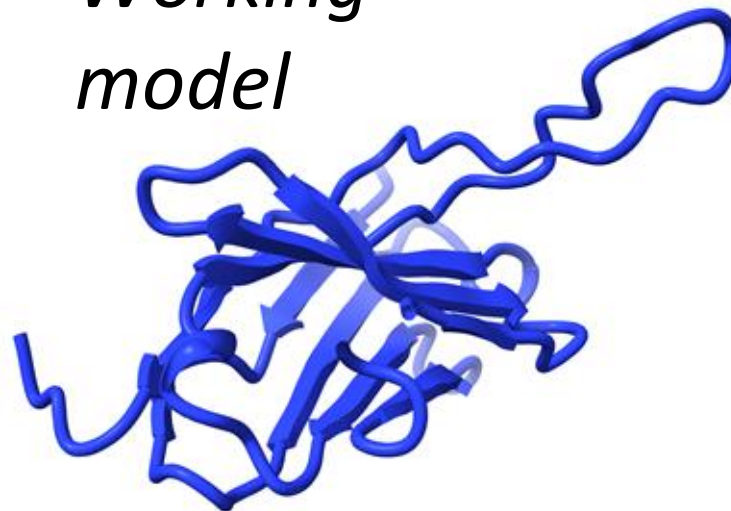
Why?

Because your new prediction might be better than your model ...and better than your original AlphaFold prediction

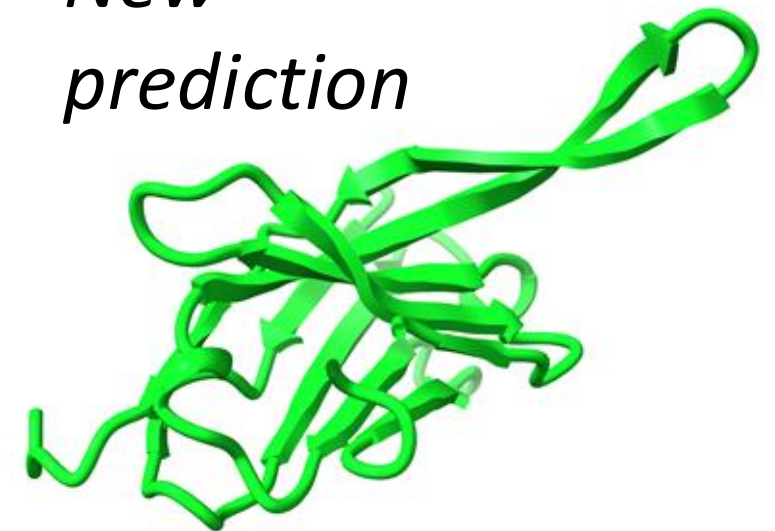
*AlphaFold
prediction*



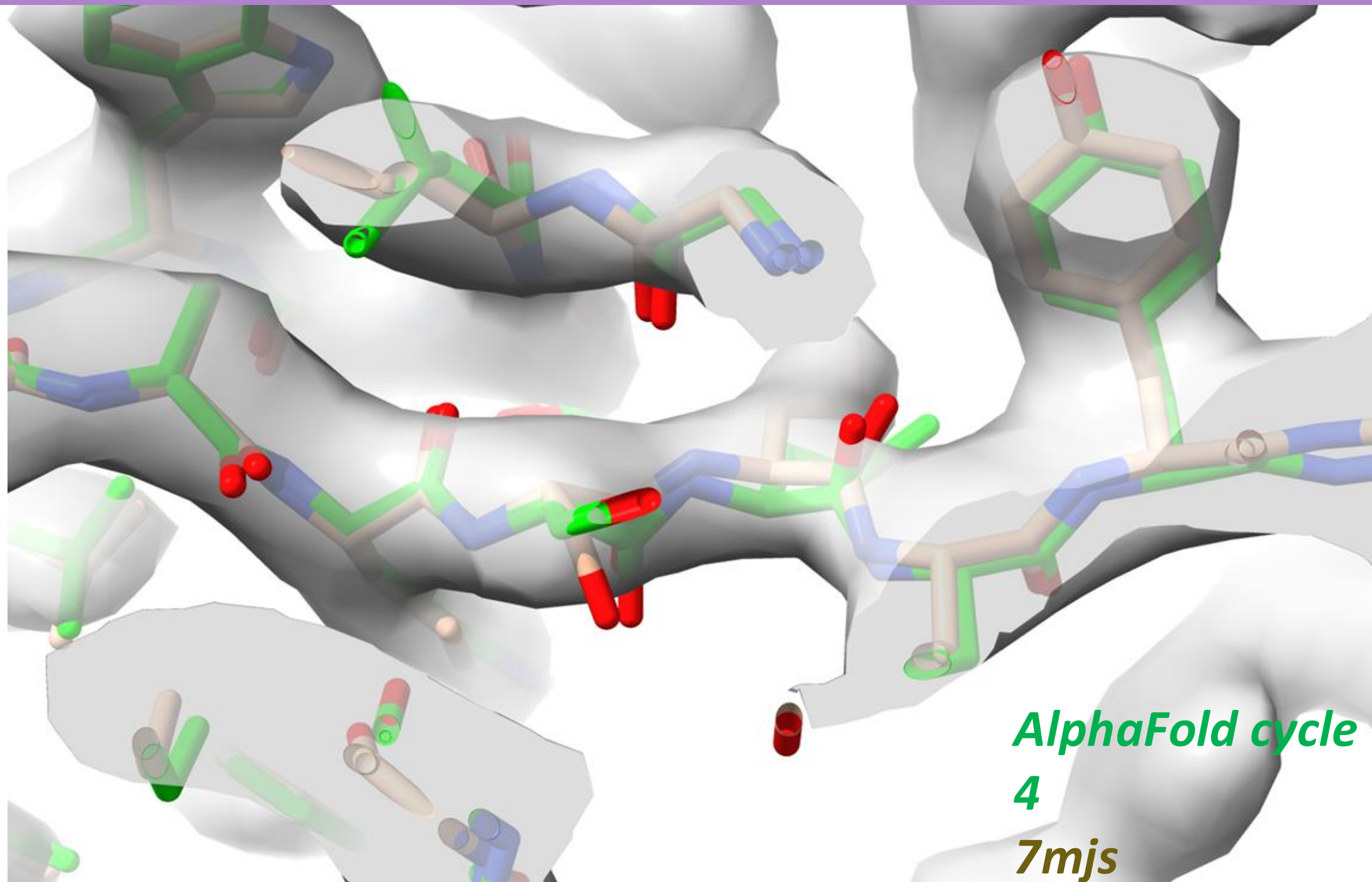
*Working
model*



*New
prediction*



Improving AlphaFold prediction using partial models as templates (Cryo-EM)



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

Strategy for structure determination in the AlphaFold era

1. Predict your structure

Design your experiment based on predicted models
Choose experimental approach
Consider trimming at domain boundaries (Perrakis & Sixma, 2021)

2. Solve your structure

Cryo-EM or X-ray MR with trimmed predicted model, SAD

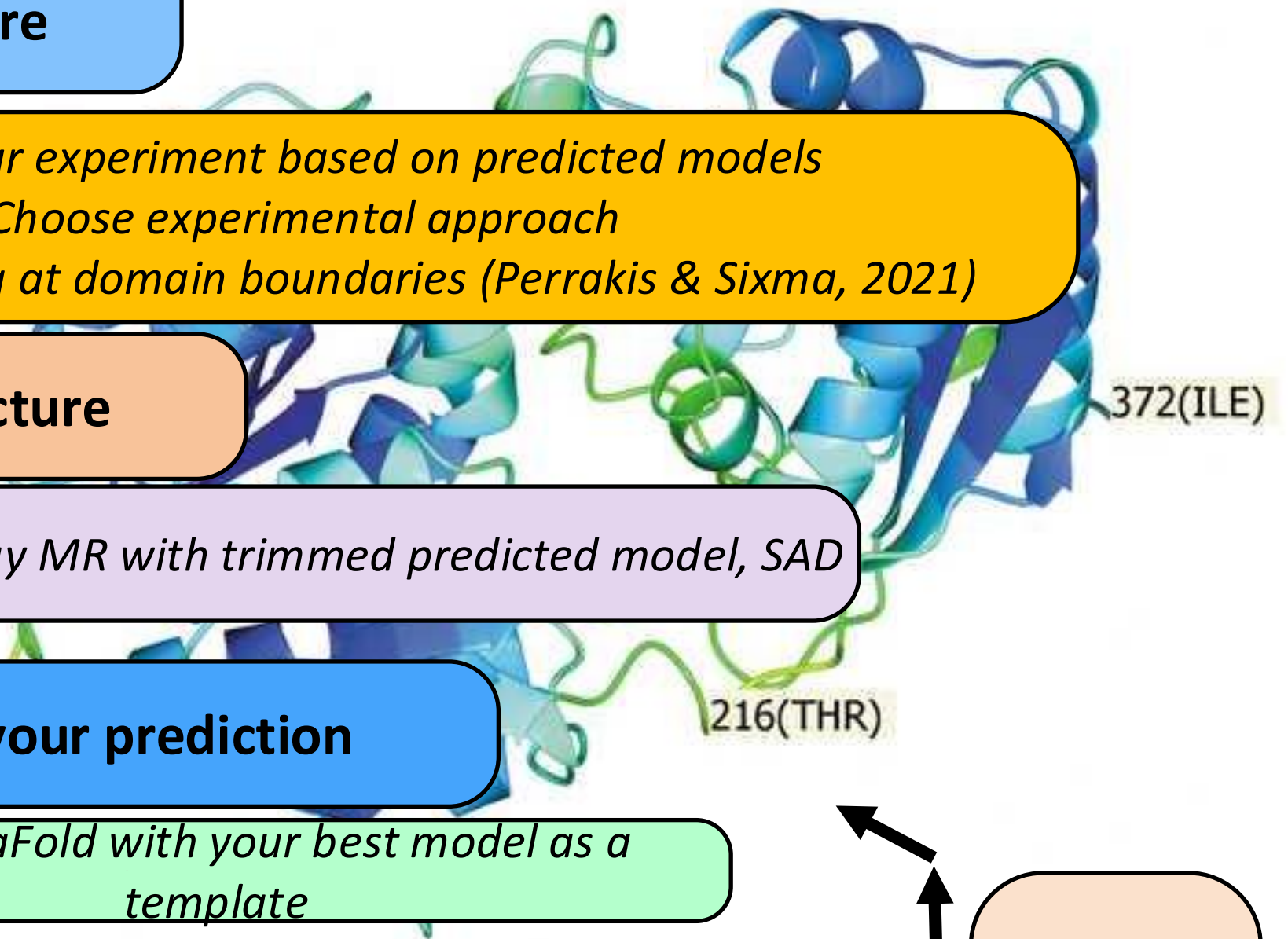
3. Update your prediction

Run AlphaFold with your best model as a template

4. Improve your structure

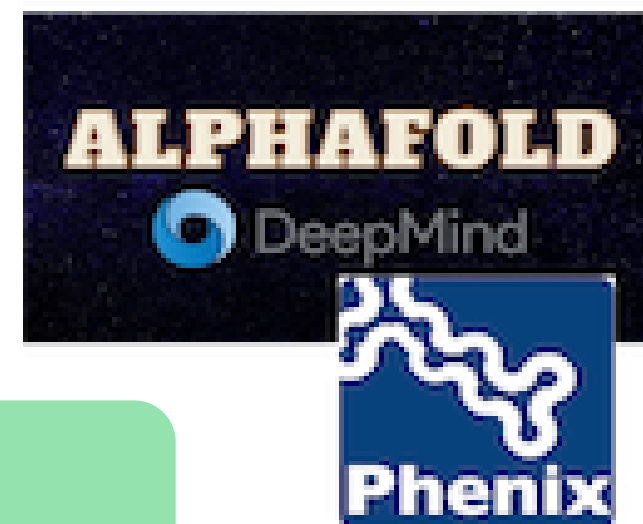
Use your new predictions as hypotheses

Iterate



Phenix AlphaFold prediction server

Available from the Phenix GUI



*Predicts structures of protein chains
(one at a time)*

Can use a template to guide the prediction

You do not need an MSA (multiple sequence alignment) if you supply a template

The template should not be an AlphaFold model

Many thanks for AlphaFold, ColabFold scripts, and the MMseqs2 server for MSAs

Process predicted model

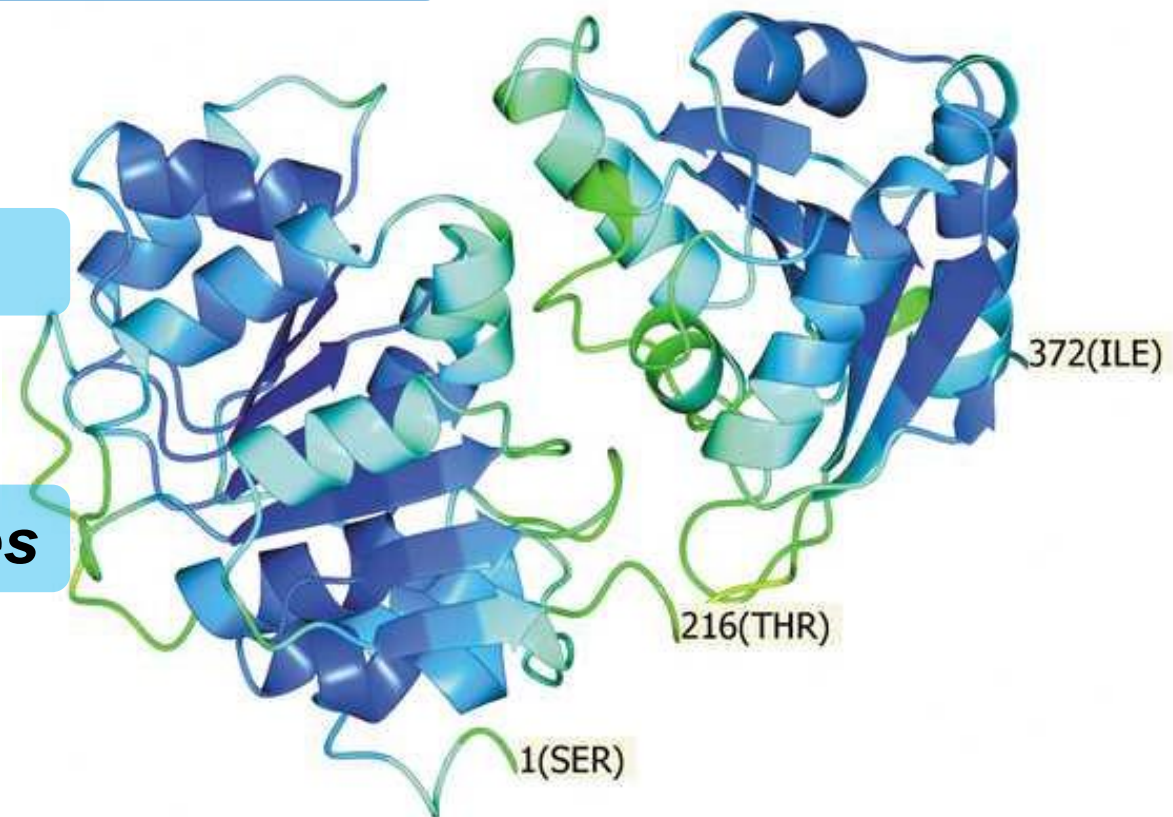
Convert pLDDT to B-value

Trim low-confidence parts of model

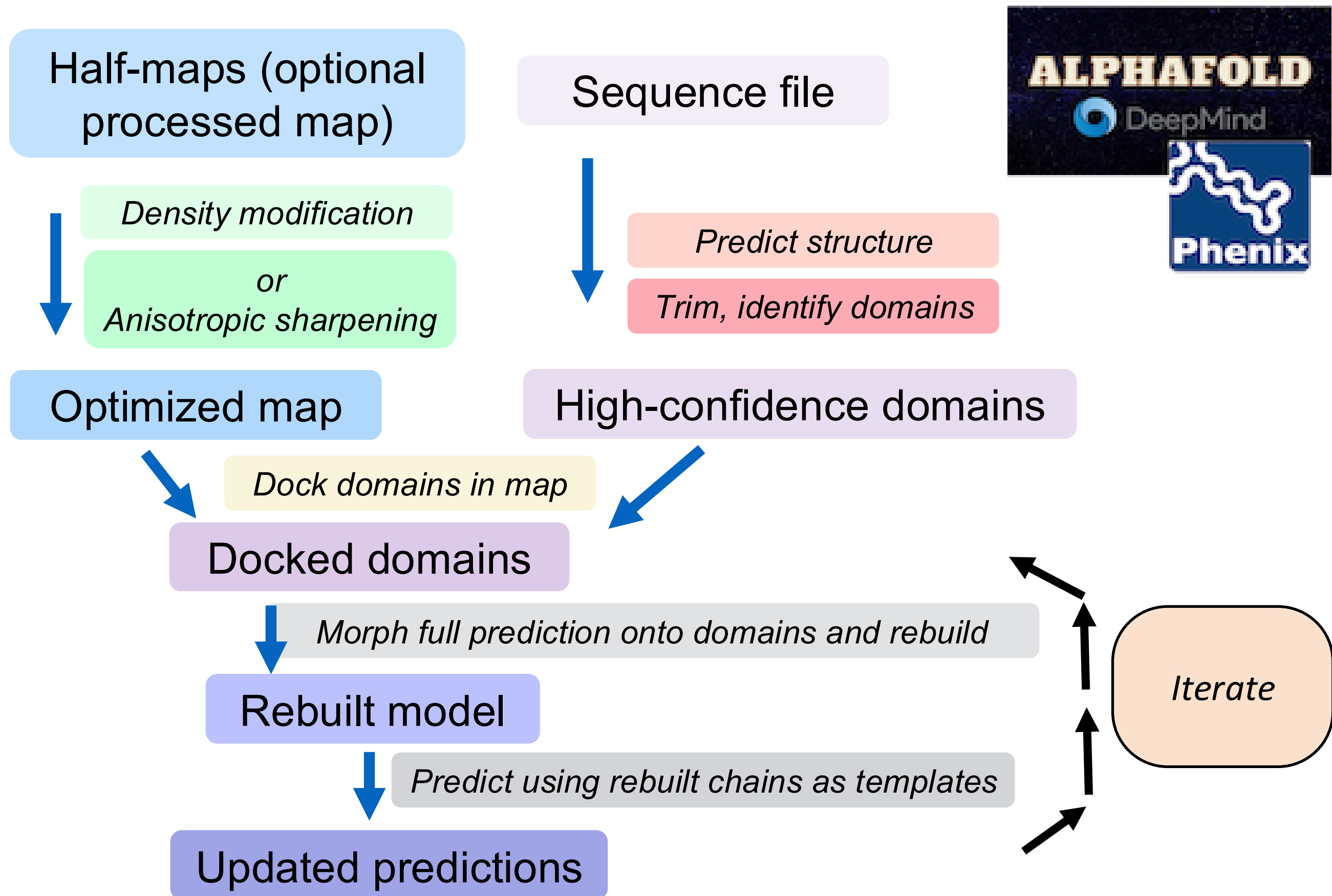
Identify high-confidence domains

Compact high-confidence regions

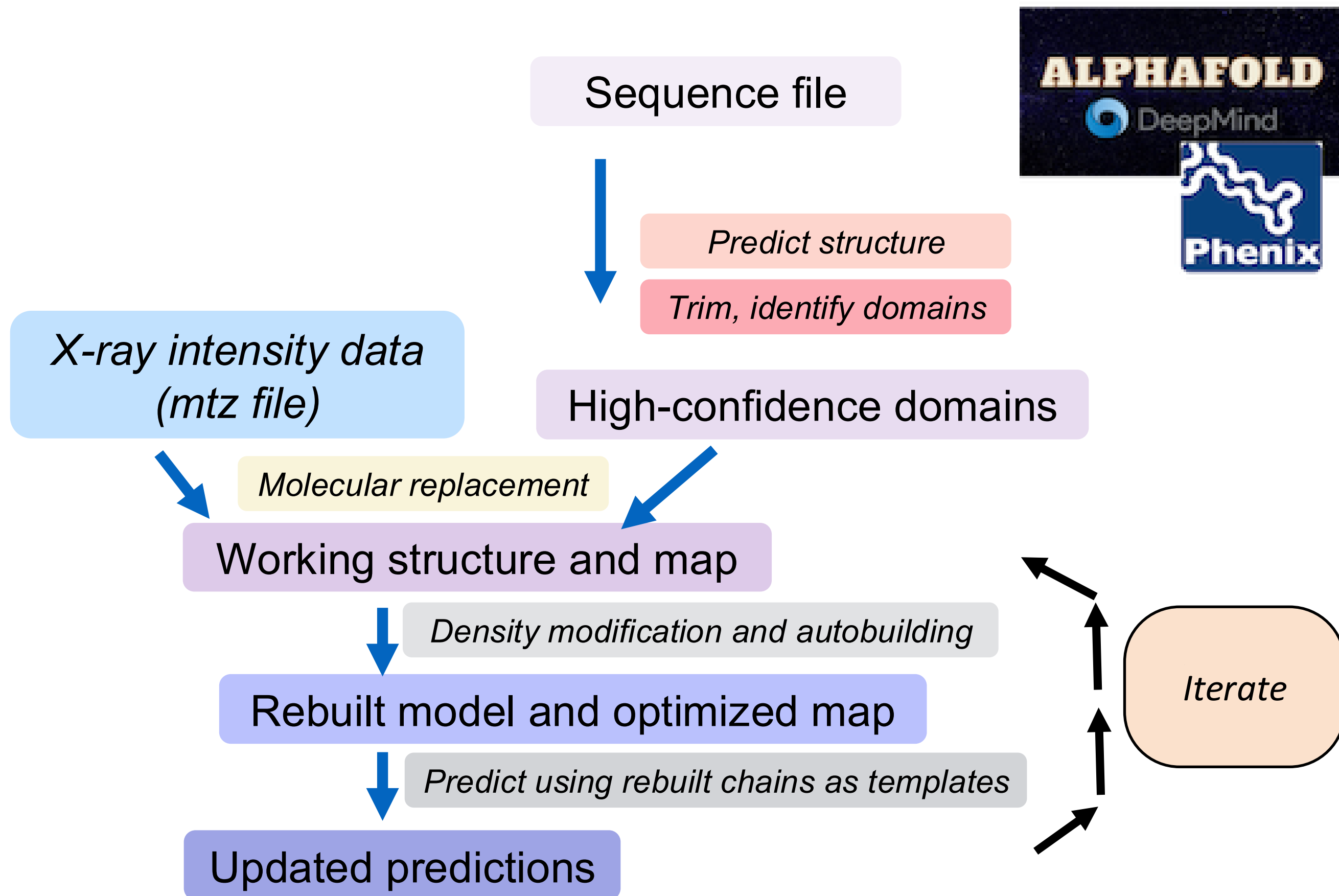
Groupings of residues with low PAE values



Cryo-EM structure determination with AlphaFold



X-ray structure determination with AlphaFold



Input and output from structure determination with AlphaFold

Input

Experimental data (maps or X-ray data)

Contents of asymmetric unit (sequence file)

Output

*Rebuilt model
Optimized map*

*Map and model ready
for next steps*

Docked predicted models

*Useful as high-quality
reference models*

Phenix tools for structure determination with AlphaFold

PredictModel (Predict with AlphaFold)

*AlphaFold
models*

ProcessPredictedModel (Trim and identify domains)

ResolveCryoEM, LocalAnisoSharpen (map improvement)

EMPlacement, DockInMap (Docking of single, multiple chains)

Cryo-EM

DockAndRebuild (Morphing and rebuilding)

RealSpaceRefine (Refinement)

Phaser-MR (Molecular replacement)

AutoBuild (Density modification and rebuilding)

X-ray

Phenix.refine (Refinement)

PredictAndBuild (Prediction and structure determination)

*Full
automation*

Phenix AI tools

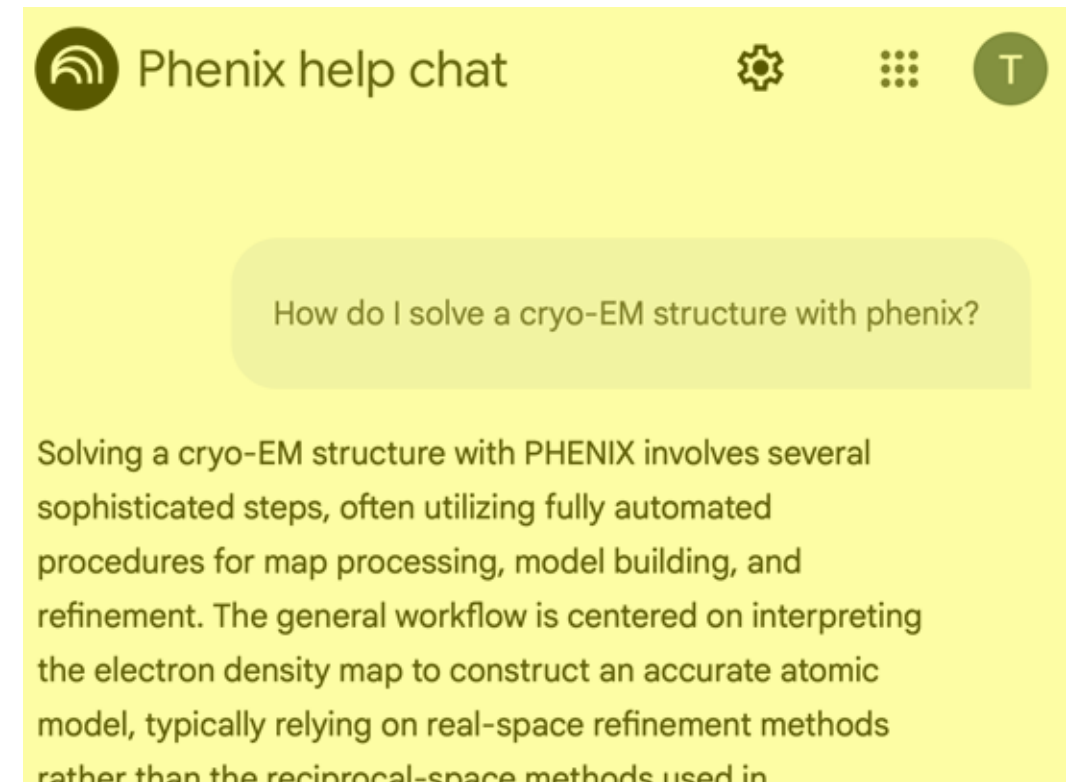
Phenix chatbot

<https://phenix-online.org/chatbot>

*Chat button in main
Phenix GUI*

AI analysis of results

*AI analysis
button in
Results panel
of Phenix GUI*



3. Evaluation of the Run:

- The run appears to have been useful, as indicated by
 - **R-value (0.26)** and **R-free (0.31)** are within acceptable ranges, suggesting a good fit to the experimental data.
 - **RMSD (0.27 Å)** indicates a high level of structural accuracy.
 - **Map-model CC 2mFo-DFc (0.78)** and **Map-model CC mFo-DFc (0.65)** indicate a strong correlation between the model and the electron density map.
 - **pLDDT (95.48)** indicates high confidence in the predicted atomic positions.
- These metrics suggest that the goals of the program, to solve and refine a high-resolution cryo-EM structure, were largely achieved.

4. Suggested Next Steps:

- **Refinement with 'phenix.refine':**
 - **Input:** The best overall predicted model ('Predicted Model')



The Project



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Oleg Sobolev,
Christopher Schlicksup



University of Cambridge

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Alisia Fadini



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Tom Terwilliger, Li-Wei Hung



UTHealth

Matt Baker



Duke University

Jane Richardson, Vincent
Chen, Michael Prisant,
Christopher Williams,



An NIH/NIGMS funded
Program Project

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