Protein structure prediction with in-cell photo-crosslinking mass spectrometry and deep learning

Kolja Stahl, Andrea Graziadei, Therese Dau, Oliver Brock, Juri Rappsilber, Nature Biotechnology, 2023 (from Technische Universität Berlin, Germany)

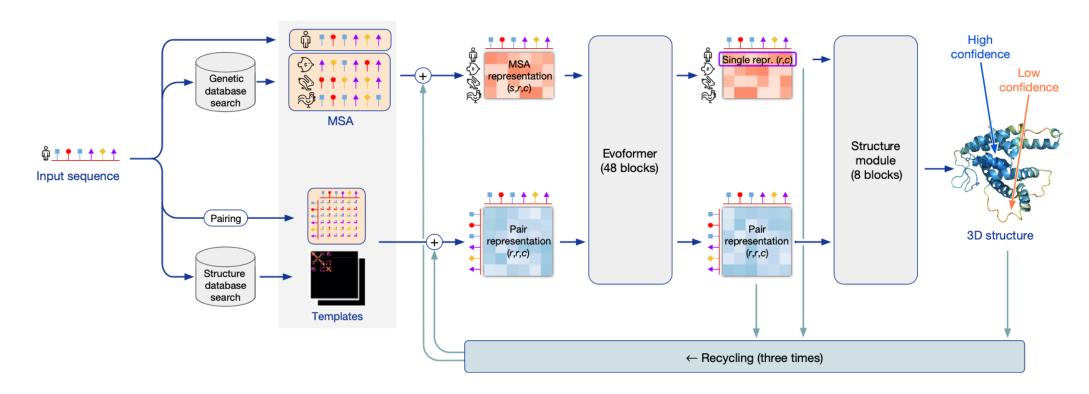
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Presentation Line

- Background & Motivation
- AlphaLink architecture
- AlphaLink methods
- AlphaLink training data
- Main experimental results
- Discussions

Recap AlphaFold2



- Multiple Sequence Alignment (MSA) + Templates of Similar Protein Structures
- Evoformer (distance space)
- Structure Module (3D space)

Background

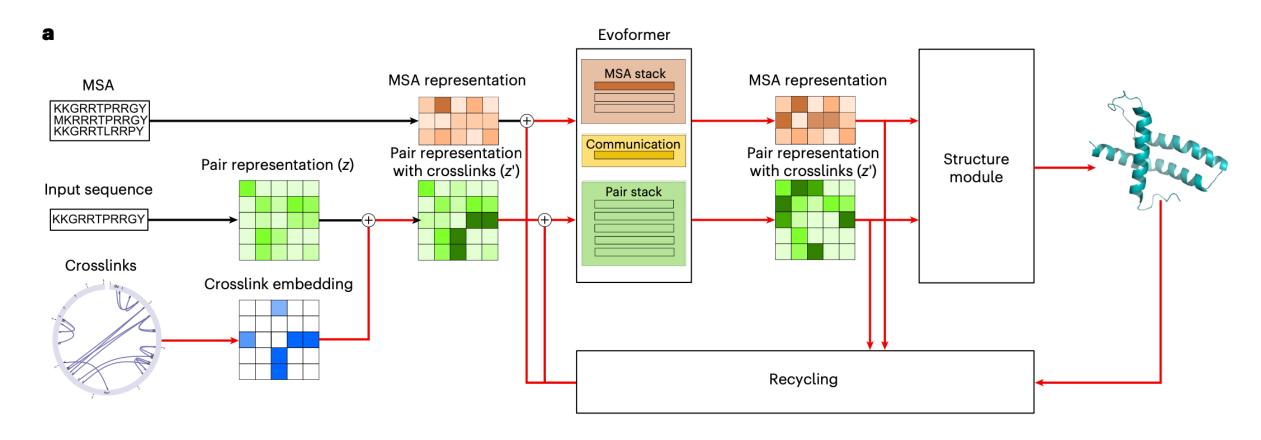
- AlphaFold2 predicts static models based on static input data, which is trained on two information sources, the protein structures in the Protein Data Bank (PDB) and multiple sequence alignments (MSAs).
- Challenged by targets with <u>insufficient evolutionary information</u>
 - viral proteins, proteins from understudied organisms, antibodies, and synthetic proteins have <u>misleading information</u>
 - structures underlying the model <u>poorly reflect structural flexibility</u>, multiple conformations, and dynamic interactions

Motivation

- Structural restraints observed on proteins in solution steer the prediction towards structural states occurring in situ under specific conditions.
- Crosslinking mass spectrometry (MS): distance restraints
 - photo amino acids (photo-AA) in prokaryotic and eukaryotic cells
 - photo-AA crosslinks: tight distance restraints that align well with coevolutionary contacts.
 - Photo-leucine (photo-L): mapping conformations & binders.

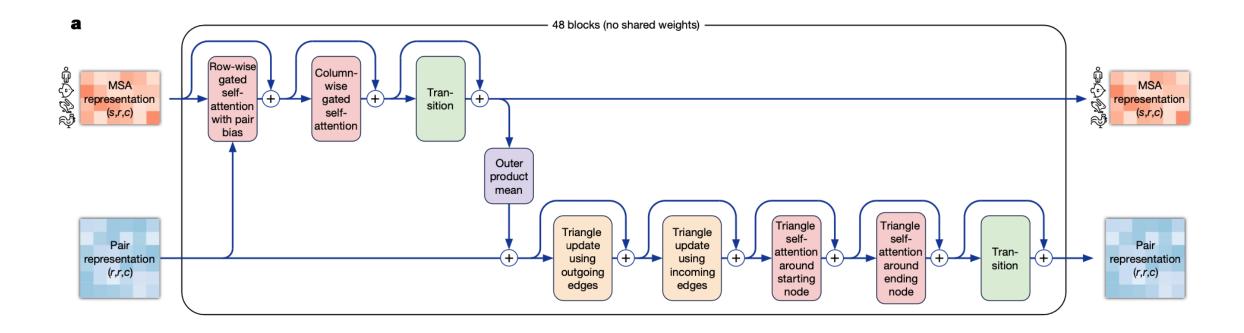
$$N=N$$
 NH_2
OH

AlphaLink: integrating crosslinks into AlphaFold2 via OpenFold



• Crosslinks (blue) are embedded and added onto the pair representation (green)

Recall Evoformer

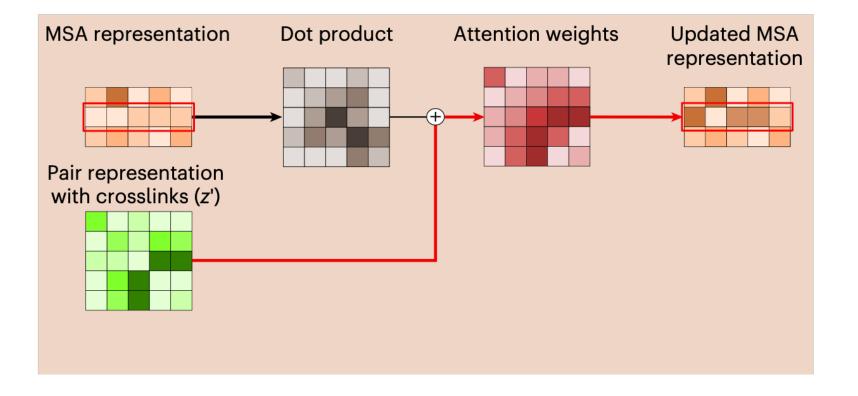


How to encode crosslink

- Two representations to encode crosslinking information
 - **soft labels**: each contact is weighted by the link-level false discovery rate (FDR) of the dataset (1-FDR)
 - distance distributions (distograms): uniformly distributed distograms for the given cutoff
- Use the same binning for the first 64 bins in Evoformer and extend the distogram further to 128 bins, spanning from 2.3125 A° to 42 A°.

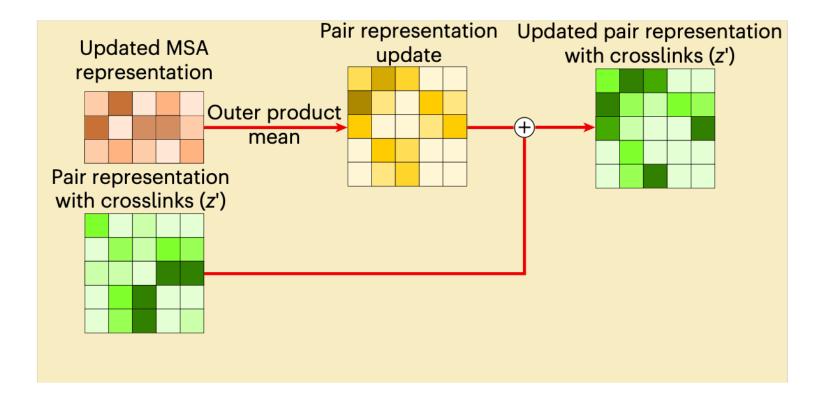
Crosslink as a bias in MSA transformer

Crosslinks influence the retrieval of co-evolutionary information.

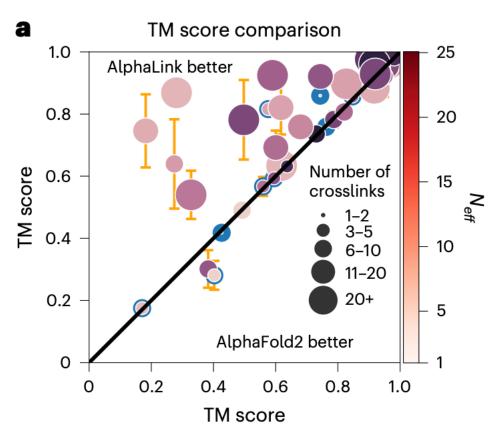


Crosslink as a bias in pair representations

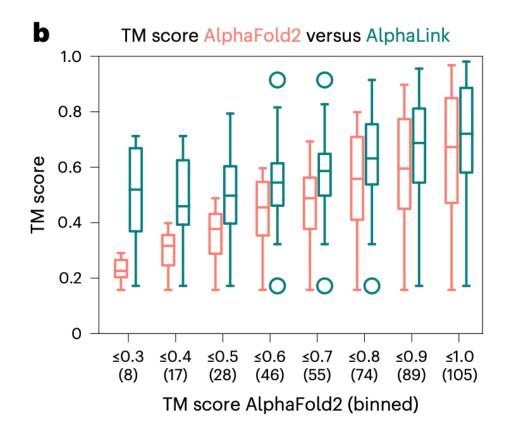
• The pair representation is updated with information from the MSAs that have been biased with the crosslinks.



Integrating photo-AA crosslinks enables noise-tolerant prediction of challenging targets



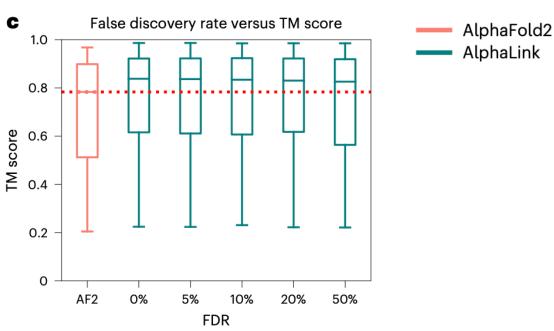
• TM score comparison on 49 CAMEO targets. TM score improves on average by 19.2%.



• TM score performance on 60 CASP14 and 45 CAMEO targets (15.2% improvement)

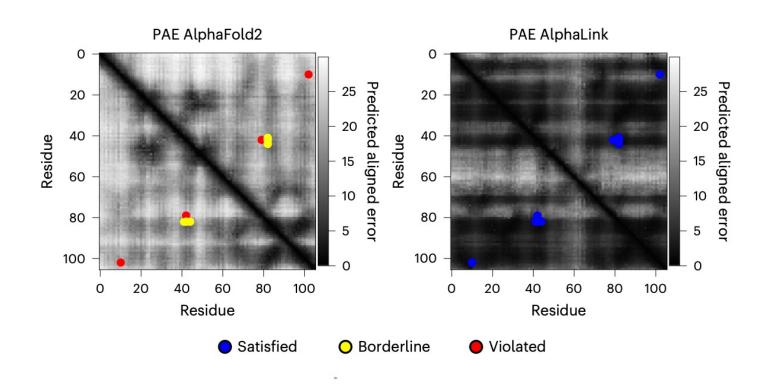
Performance on different noise levels

- Performance on 60 CASP14 targets with different noise levels (FDR 0%, 5%, 10%, 20% and 50%)
- AlphaLink improves in the median for all noise levels. Performance shows robust noise rejection.

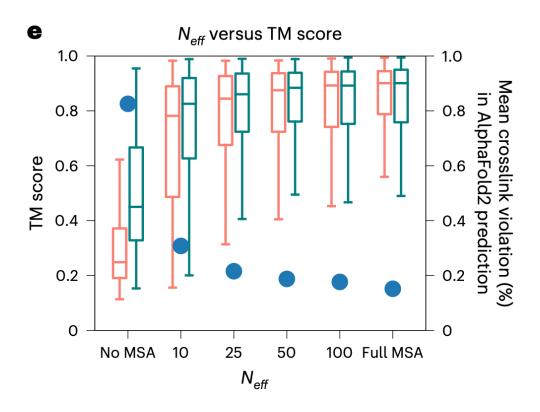


Sparse restraints decrease uncertainty across the whole protein

- Predicted aligned error of AlphaFold2 (left) and AlphaLink (right) on T1064.
- Satisfied crosslinks <10 A° C α –C α , borderline crosslinks (10–15 A° C α –C α), violated crosslinks >15 A° C α –C α



Crosslink diminishes with increasing MSA size



- Performance on 60 CASP14 targets
 as a function of MSA size (N = 100,
 10 MSAs and 10 crosslink sets)
- Blue dots represent the mean percentage of nonsatisfied crosslinks (>10 A° Cα–Cα) in the AlphaFold2 prediction.

Performance without MSAs

- Performance without MSAs on 60
 CASP14 and 45 CAMEO targets.
- AlphaLink predicts the correct fold (TM score >0.5) for 43/105 (13/105 for AlphaFold2).
- Error bars represent the 95% confidence interval (N = 10).

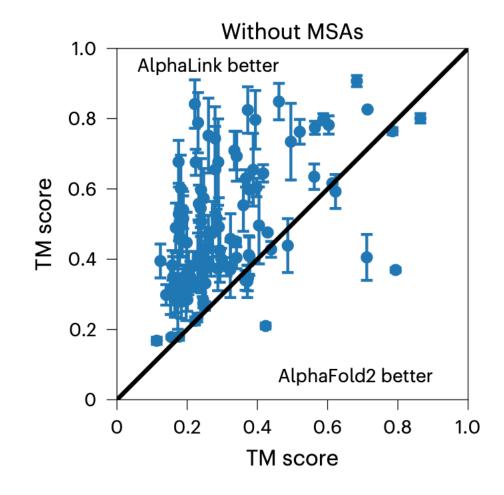
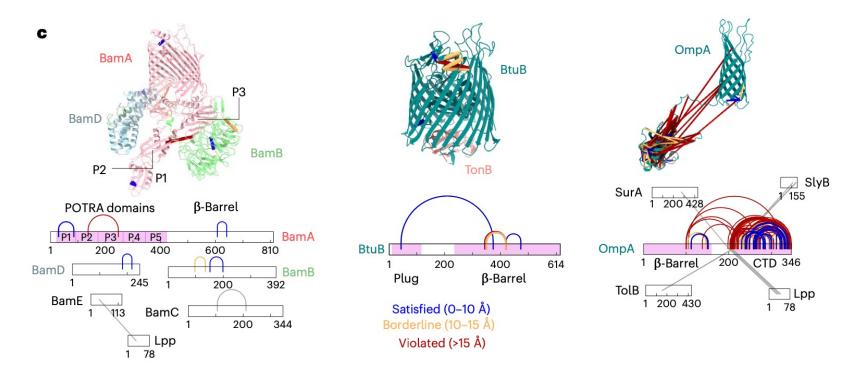


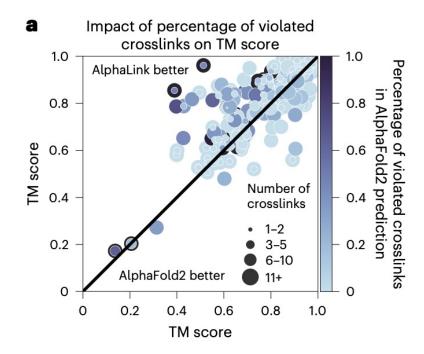
Photo-Leucine as an in situ structural probe

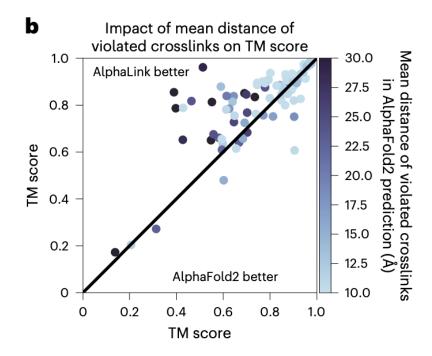
- large-scale experimental photo-AA dataset with in situ structural restraints on the Escherichia coli membrane fraction
- 615 residue pairs involving 112 proteins at 5% link-level FDR
- Photo-Leucine provides in situ conformation of multiprotein complexes



Structure prediction with Photo-Leucine data

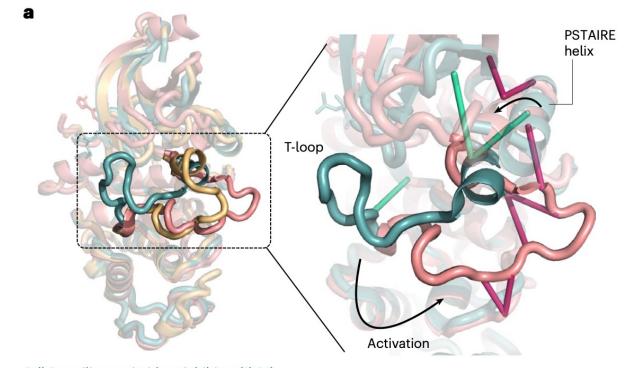
- 31 targets with high-resolution structures
- **a**: performance improvement is bigger for targets with a higher percentage of nonsatisfied crosslinks in the base prediction
- **b**: predictions that improved the most have unsatisfied crosslinks with large distances





Probing conformational dynamics in situ

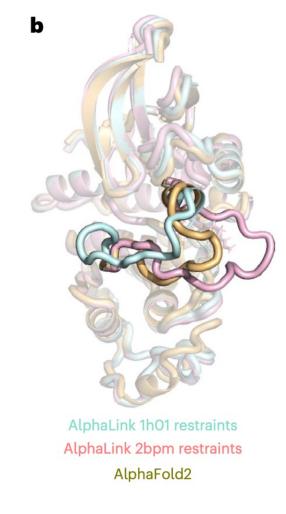
- a <u>proof-of-concept</u> experiment on the human cyclin-dependent protein kinase
 Cdk2, a drug target in cancer therapy.
- Activation of Cdk2 in the S phase
 proceeds via a conformational change
 in the <u>T-loop</u> (residues 145–165) and
 the <u>PSTAIRE helix</u> (residues 45–55)
 triggered by binding of cyclin A.



Cdk2—anilino pyrimidase inhibitor (1h01)
Cdk2—cyclin A—PHA-630529 (2 bpm)
AlphaFold2 prediction

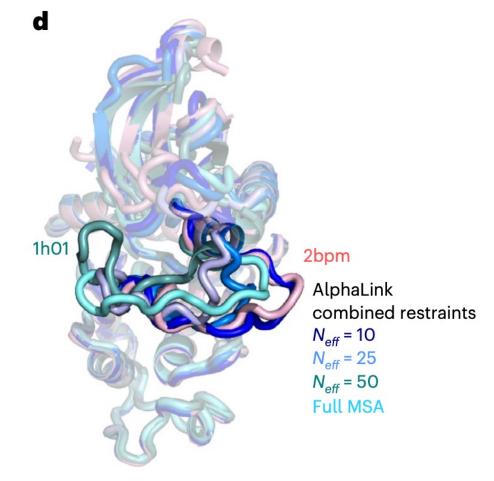
Comparison on active/inactive conformation

- photo-crosslinking: protein was acquired in either its inhibited or in its cyclin A-bound states.
- Cdk2 structure using AlphaLink with these restraints, showing that the <u>loop structure</u> is driven towards the appropriate conformation.
- all AlphaFold2 predictions converge to the cyclin A-bound state, <u>failing to predict the inactive</u> conformation.



Outcome with a combined set of restraints

- At low $N_{\rm eff}$ (number of effective sequences) values, the crosslinks drive the prediction towards the cyclin E-bound state.
- As the MSA information increases, the prediction is steered more towards the inhibited state and closer to the AlphaFold2 prediction.
- Crosslinking is weighted against the MSA depending on the information <u>content</u> and <u>size</u> of both strands of information.



Conclusion

- AlphaLink integrates experimental data from photo-AA crosslinking directly into the AlphaFold2 architecture to merge co-evolutionary relationships and crosslinking data in distance space, exploiting the complementary nature of the data.
- AlphaLink can leverage noisy experimental contacts to improve predictions.
- AlphaLink performs a large-scale crosslinking MS study with photo-Leucine,
 identifying 615 in situ residue-residue contacts in *Escherichia coli* membrane fractions.
- Even sparse crosslinking MS data can anchor predictions to particular conformational states, opening up the possibility of hybrid experimental/deep learning approaches.

Questions