Highly accurate protein structure prediction with AlphaFold

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What are Proteins?

- Proteins are long chains of amino acids
- They perform many functions within organisms



Fig. 1: Primary structure of a protein

Protein Folding



Fig. 25: DNA replication by DNA polymerase

Fig. 2: Kinesin protein walking on microtubule

Protein Folding



Why prediction?

- Finding the structure of proteins is time intensive
- The process can span years for complex proteins



CASP

- Participants are asked to predict the structure of Proteins
- Predictions are made on Proteins with previously unknown structure
- Problem of structure prediction considered solved at a GDT-TD of 90





Median Free-Modelling Accuracy

Fig. 6: Performance of AlphaFold2 compared to previous competitions

The Network



The Network





Input

- MSA maps evolutionary relationships between residues
- Conserved regions are likely to be functionally important



Fig. 8: Illustration of MSA

Input

- The model uses the structures of similar proteins as a basis
- These templates are used to bias the pair representation



Fig. 9: Examples of myoglobin from different animals

Evoformer



Evoformer



Row/Column-wise Attention



Fig. 11: Row attention block

Evoformer



Triangle updates

- Basic idea: adhere to triangle inequality
- Update an edge based on all the triangles it is involved in

- C Triangle multiplicative update using 'outgoing' edges
- Triangle multiplicative update using 'incoming' edges



Structure module



Structure module

- Protein backbone modeled as 3D gas of triangles
- No enforcement of chain





Fig. 13: Illustration of the ridged body gas

Structure module



Invariant Point Attention

- 3D equivariant transformer updates triangles (position/rotation)
- Coordinates in local frame are invariant in respect to global frame





Residue prediction

 The atom location of the side chains can be parametrized using only four additional angles



The Model



Physical correction



Recycling iteration 0, block 01 Secondary structure assigned from the final prediction

Vid. 16: AlphaFold folding a complex protein

Model Confidence

- Alpha fold produces confidence measures for each residue
- This is essential for the interpretability of the produced structures





https://alphafold.com/entry/Q9Y223

Fig. 18: Alpha Fold prediction overlayed with experimental structure (green)

Fig. 17: Confidence Matrix

Results





Recycling iteration 0, block 01 Secondary structure assigned from the final prediction

Fig. 20: GDT against number of Evoformer blocks

Results



Fig. 22: Structure of SARS-CoV-2 ORF3a

Sources

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Fig. 6: https://www.deepmind.com/blog/alphafold-a-solution-to-a-50-year-old-grand-challenge-in-biology 11.03

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Fig. 8: https://jgi.doe.gov/seeking-structure-metagenome-sequences/cartoon-coevolution-sergey-o/

Fig. 9: https://i0.wp.com/www.blopig.com/blog/wp-content/uploads/2021/07/myoglobin-1.png 14.03

Fig. 10: Jumper, J., Evans, R., Pritzel, A. et al. Highly accurate protein structure prediction with AlphaFold (fig. 3)

Fig. 11: Jumper, J., Evans, R., Pritzel, A. et al. Highly accurate protein structure prediction with AlphaFold (sup. fig. 3 (Additional Material))

Fig. 12: Jumper, J., Evans, R., Pritzel, A. et al. Highly accurate protein structure prediction with AlphaFold (fig. 3)

Fig. 13: https://github.com/bayesgroup/bayesgroup.github.io/blob/master/bmml_sem/2021/Figurnov_Alphafold.pdf 11.03

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Fig. 22: Jumper, J., Evans, R., Pritzel, A. et al. Highly accurate protein structure prediction with AlphaFold (fig. 4)

Fig. 23: Jumper, J., Evans, R., Pritzel, A. et al. Highly accurate protein structure prediction with AlphaFold (fig. 1)

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Triangle update



Jumper, J., Evans, R., Pritzel, A. *et al.* Highly accurate protein structure prediction with AlphaFold (Additional Material)

Triangle attention



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IPA



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Feature embedding



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Alpha Fold 1

a Sequence and MSA features

Deep neural network

Distance and torsion distribution predictions

Gradient descent on protein-specific potential

AlphaFold: Improved protein structure prediction using potentials from deep learning