

# Protein Structure Prediction

BCH394P/364C Systems Biology/Bioinformatics

April 1, 2025

Daryl Barth



# Today's Goals

- Motivation and bit of history on protein structure prediction
- CASP
- AlphaFold
- Metrics you need to know
- ColabFold
- Demo!



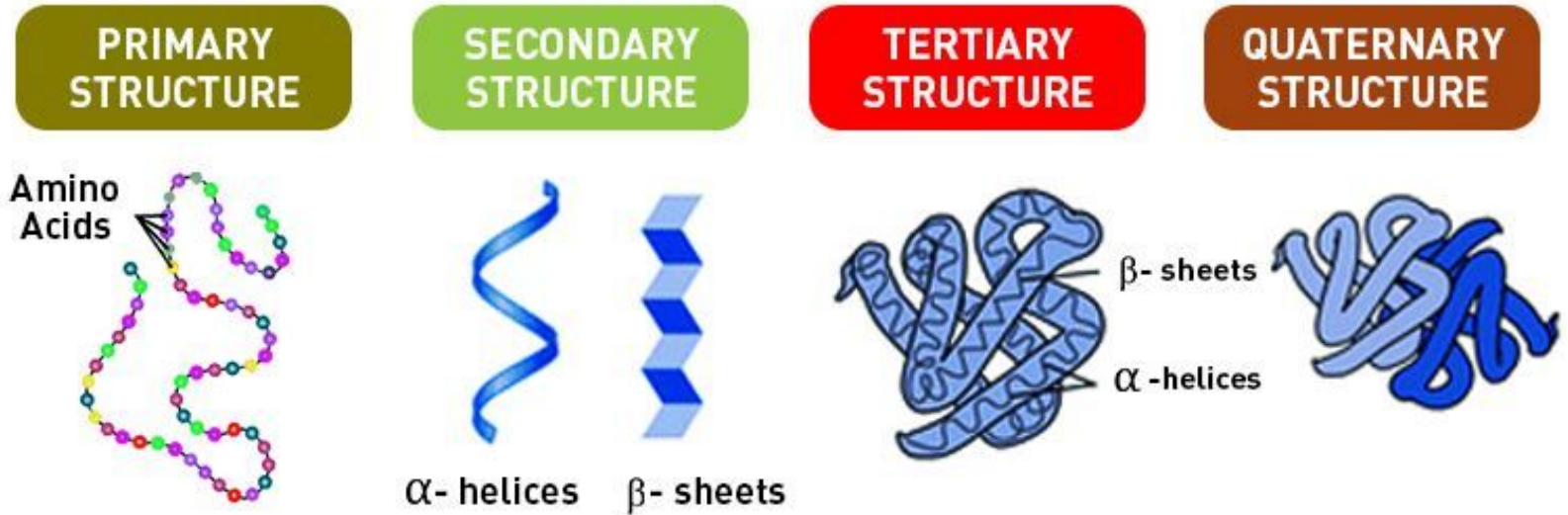
# Why?



[https://www.ted.com/talks/david\\_baker\\_5\\_challenges\\_we\\_could\\_solve\\_by\\_designing\\_new\\_proteins](https://www.ted.com/talks/david_baker_5_challenges_we_could_solve_by_designing_new_proteins)



# The four levels of protein structure





# Is an amino acid sequence all you need?

- Anfinsen's dogma:
- The Protein Folding Problem
  - What is the folding code?
  - What is the folding mechanism?
  - Can we predict a native protein structure from its primary, amino acid sequence?

A protein's native structure stands for a free energy minimum determined by its amino acid sequence...

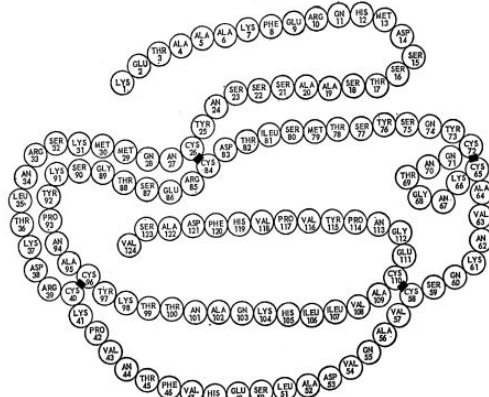
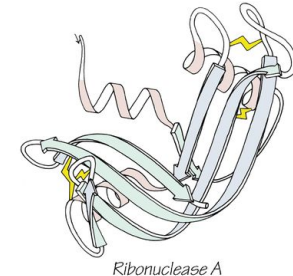


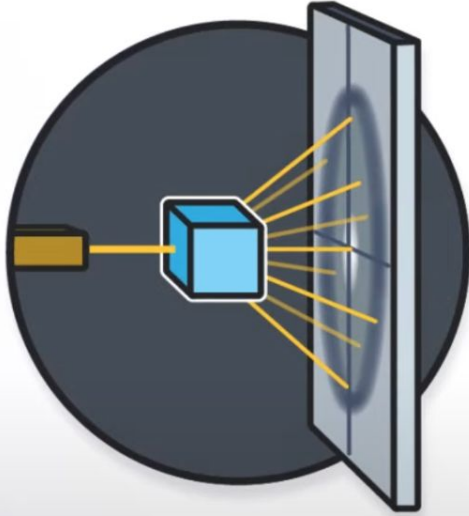
Fig. 1. The amino acid sequence of bovine pancreatic ribonuclease (50).





# Experimental Methods for Determining Protein Structure

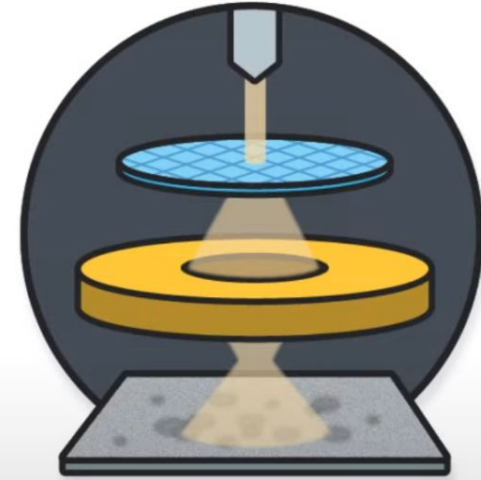
X-Ray  
crystallography



Nuclear magnetic  
resonance spectroscopy

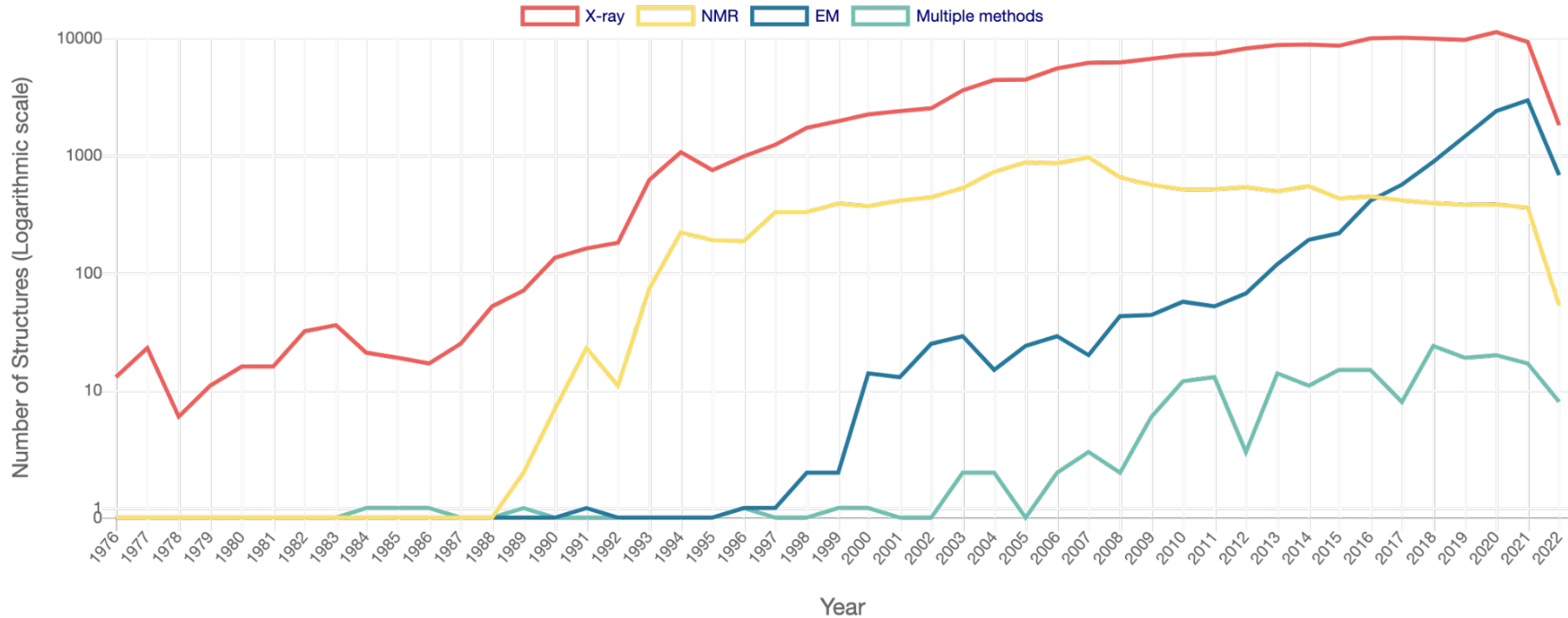


Cryoelectron  
microscopy



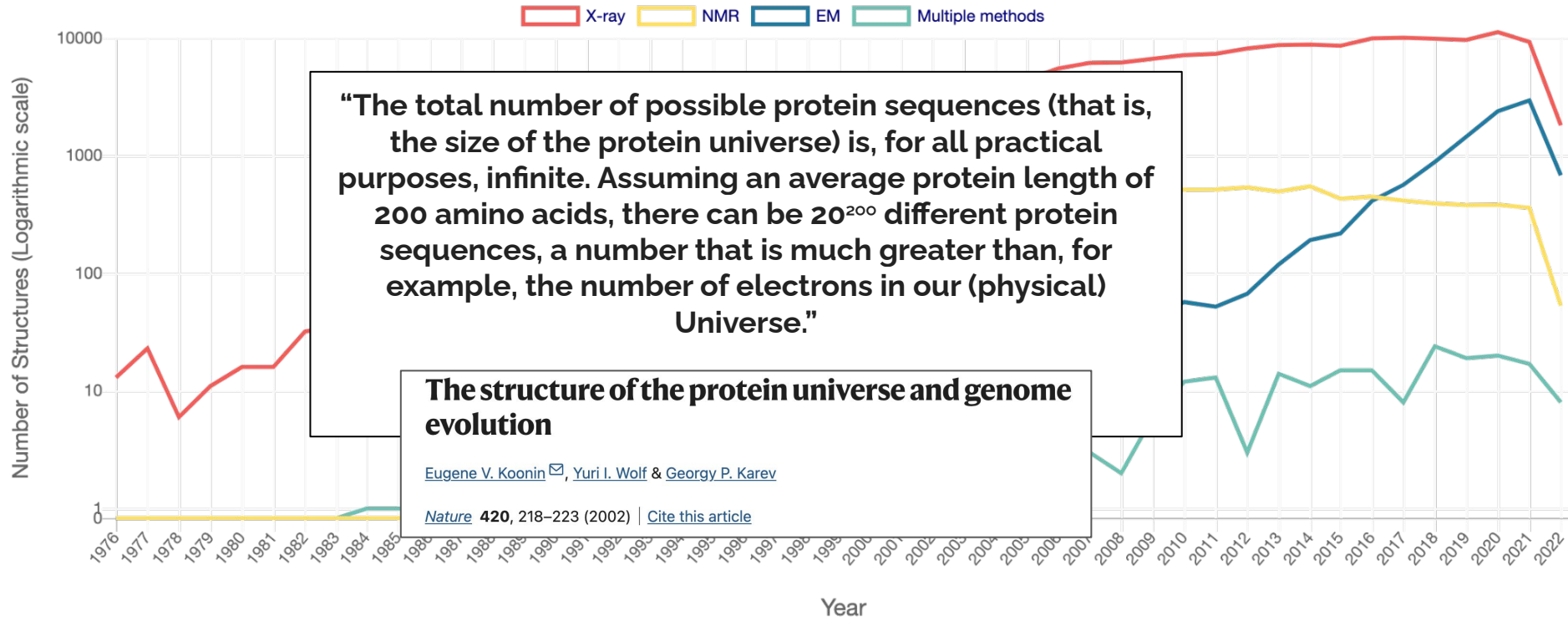


# How many structures have been determined?



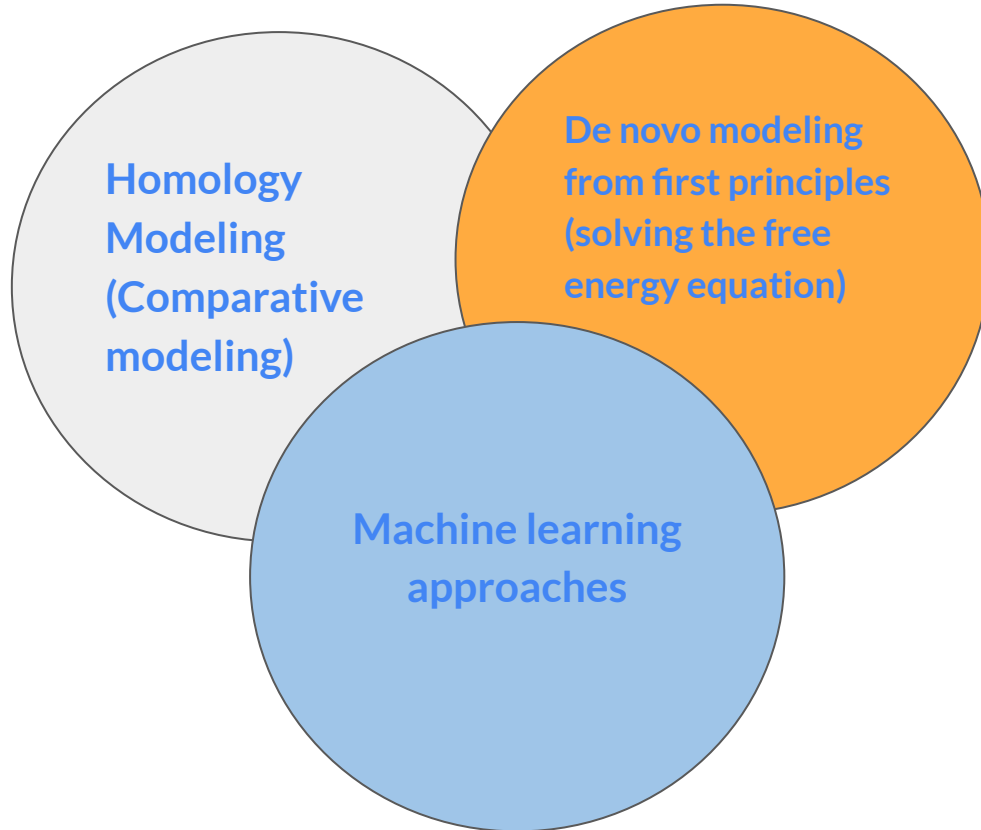


# How many structures have been determined?





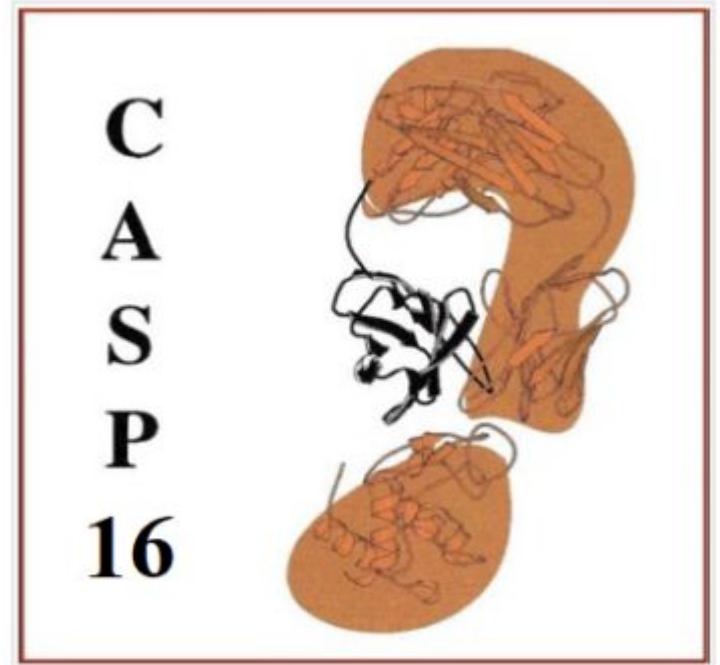
# Computational Methods





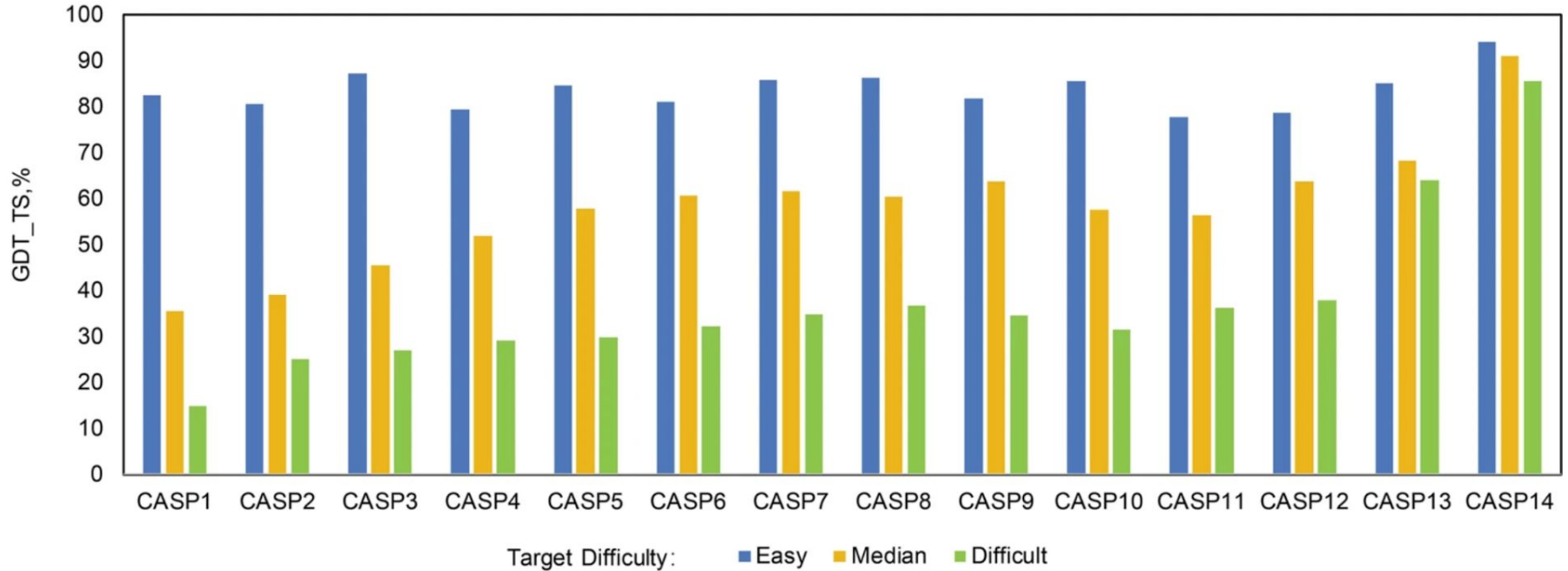
# CASP Competition drives protein structure prediction

- Critical Assessment of Structure Prediction (CASP)
- Founded in 1994
- ‘Olympics’ of protein structure prediction
- Supported by Lawrence Livermore National Labs, DOE, etc.
- Check it out:  
<https://www.predictioncenter.org>



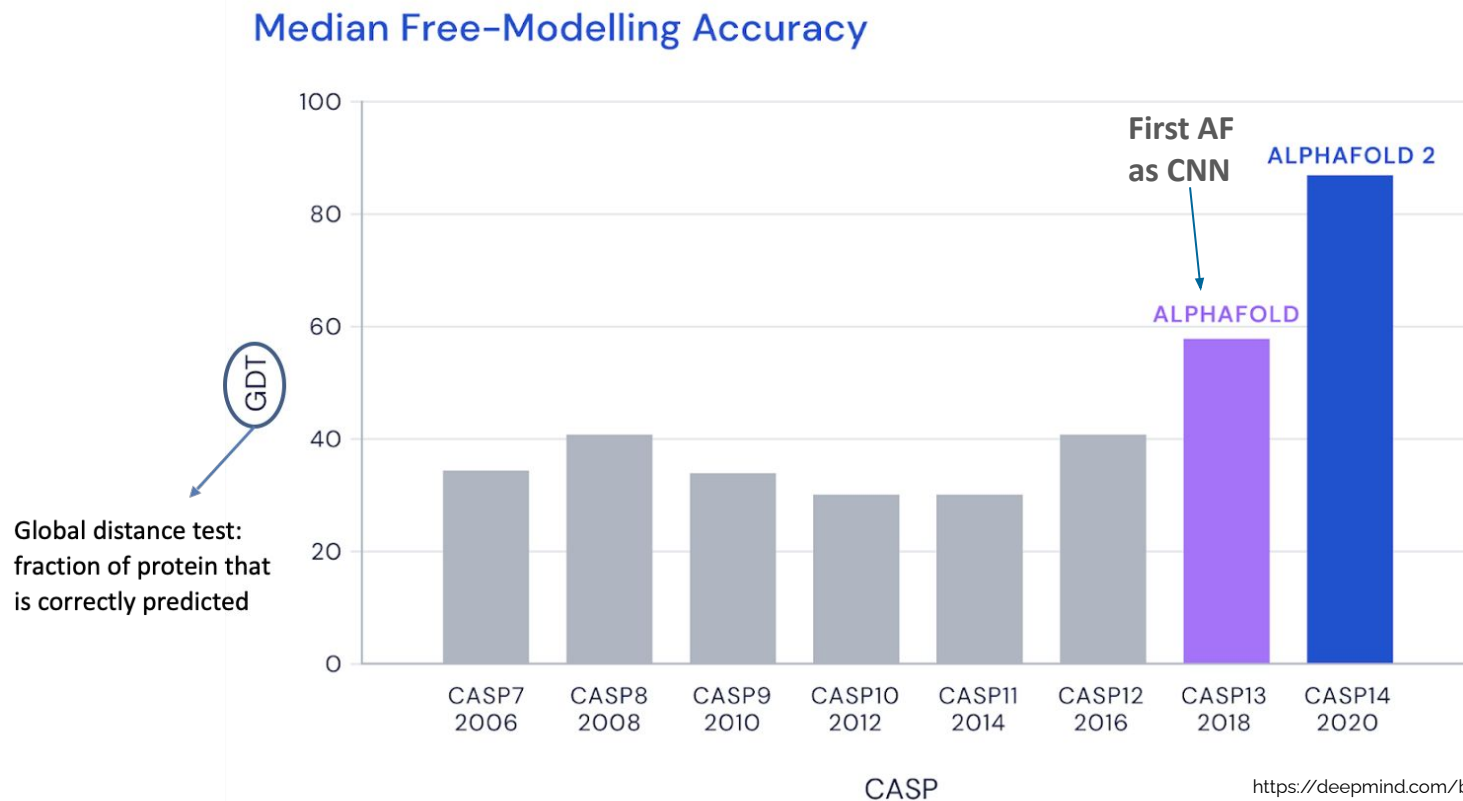


# CASP Competition drives protein structure prediction



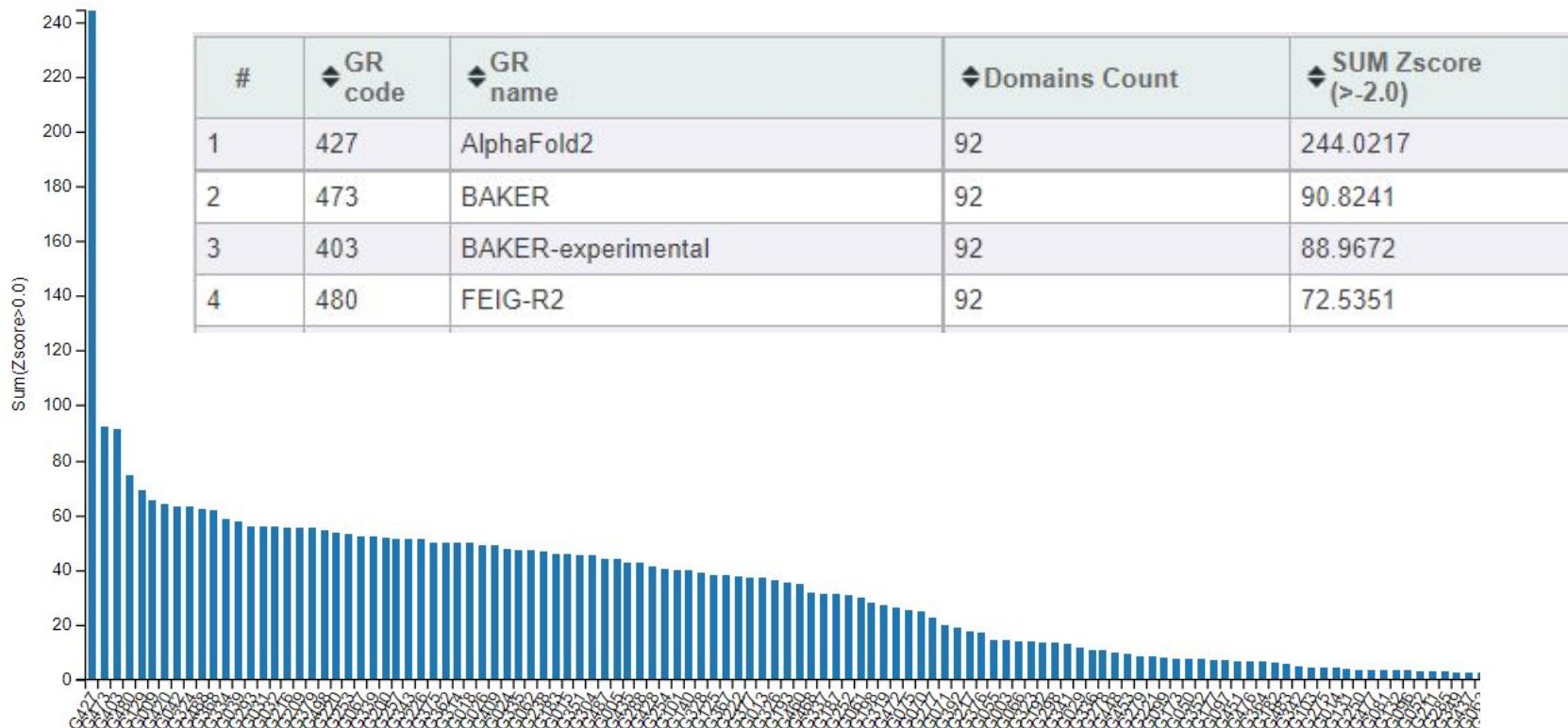


# AF2 compared to CASP winners over the years



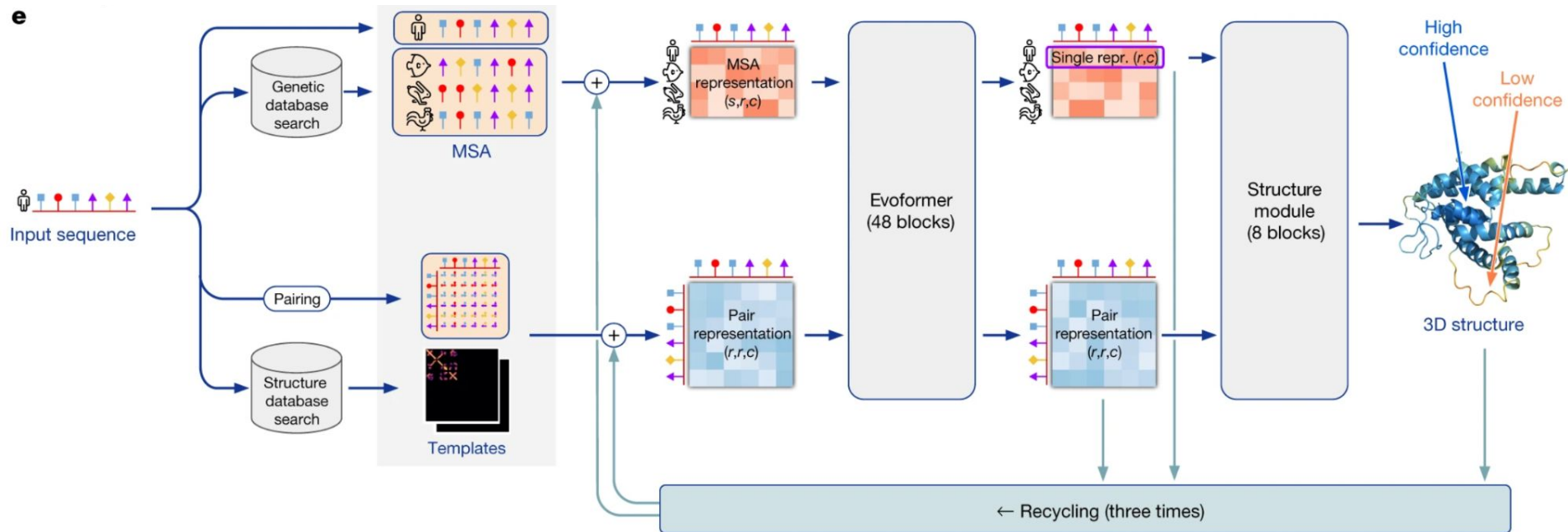


# CASP14 (2020) Results: Entrance of AlphaFold2





# AlphaFold2: the dawn of a new age



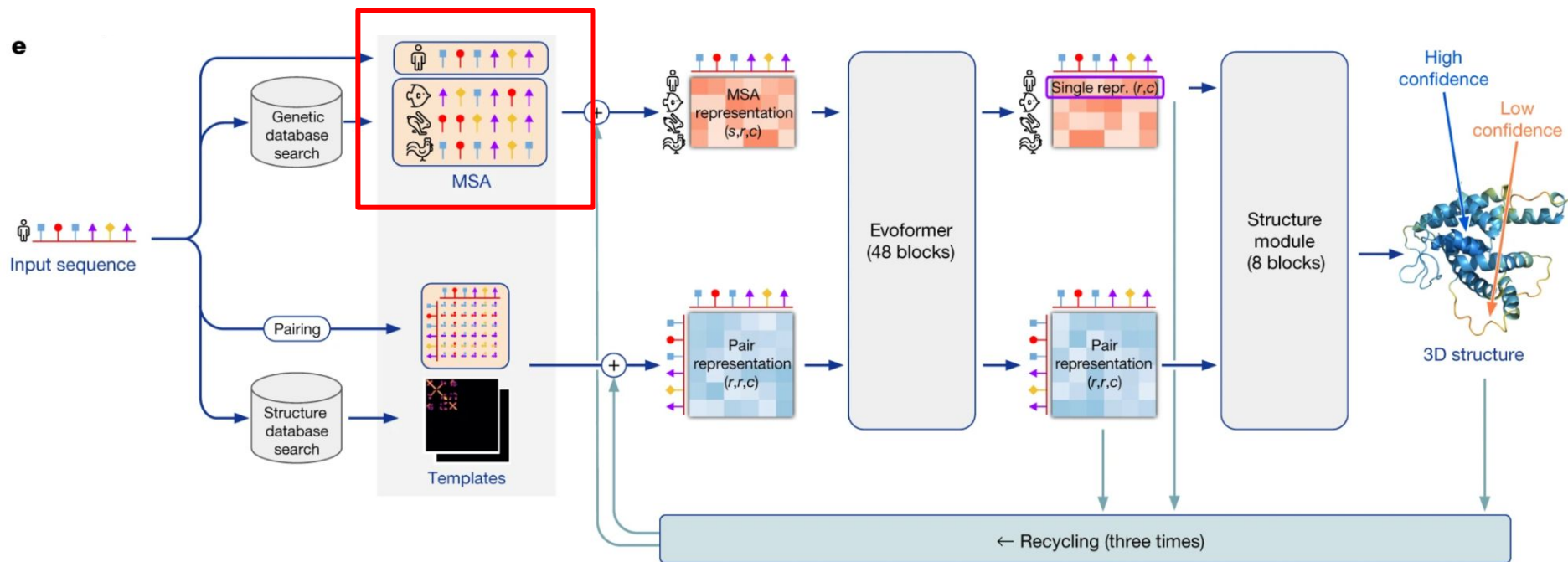


# Start Demo!

- [ColabFold Server](#)
- >2E8I\_1|Chain A|6-aminohexanoate-dimer hydrolase|Flavobacterium sp. (37931)
- MNARSTGQHPARYPGAAAGEPTLDSWQEPPHNRWAF AHLGEMVPSAAVSRRPVNAPGHA  
LARLG AIAAQLPDLEQRLEQTYTDAFLVLRGTEVVAEYYRAGFAPDDRHLLMSVSKSLCGTVV  
GALVDEGRIDPAQPVTEYVPELAGSVYDGPSVLQVLDMQISIDYNEDYVDPASEVQTHDRSA  
GWRTRRHGDPADTYEFLTTLRGDGSTGEFQYCSANTDVLAWIVERVTGLRYVEALSTYLWA  
KLDADRDATITVDTTGFGFAHGGV SCTARDLARVGRMMLDGGVAPGGRVVSEDWVRRVLA  
GGSHEAMTDKGFTNTFPDGSYTRQWWCTGNERGNVSGIGIHGQNLWLDPLTDSVIVKLSS  
WPDPDTEHWHRLQNGILLDVSRLDAV
- 392 Amino Acids, Nylon Hydrolase

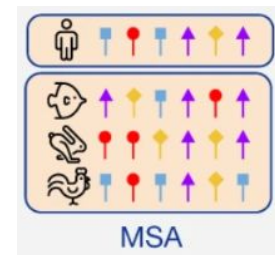


# What is an MSA?





# MSA: Multiple Sequence Alignment



			cov	pid	1	80
1	UniProt/Swiss-Prot	P26898	IL2RA_SHEEP	100.0%	100.0%	MEPSLLMWRFFVFIVVPGCVTEACHDDPPSLRNA-----MFKVIRYE---VGIMINCDCKAGFRRVS---AVMR
2	UniProt/Swiss-Prot	P01590	IL2RA_MOUSE	94.4%	45.1%	MEPRLLMLGFLSLTIVPSCRACLCLYDPPEVPNA-----TFKALSYK---NGATILNCECKRGFRRLKE-LVYMR
3	UniProt/Swiss-Prot	P41690	IL2RA_FELCA	98.9%	54.3%	MEPSLLLMGILTFVVVHGHVTELCDENPPDIQHA-----TFKALTYK---TGTMLNCECKKGFRRISNGSAFML
4	UniProt/Swiss-Prot	P01589	IL2RA_HUMAN	98.9%	47.8%	MDSYLLMWGLLTFIMVPGCQAELEDDDDPEIPHA-----TFKAMAYK---EGTMLNCECKRGFRRIKSGSLYML
5	UniProt/Swiss-Prot	Q5MNY4	IL2RA_MACMU	98.9%	48.9%	MDEYLLMWGLLTFITVPGCQAELEDDDDPKITHA-----TFKAVAYK---EGTMLNCECKRGFRRIKSGSPYML
6	UniProt/Swiss-Prot	Q95118	IL2RG_BOVIN	96.6%	11.0%	MLKPPLPLRSLLELQLLLGVGLNPKFLTPSGNEDIGGKPGTGGDFFLTSTPAGTLDVSTLPLPKVQC---FVFNVEYMN
7	UniProt/Swiss-Prot	P40321	IL2RG_CANFA	95.5%	10.7%	MLKPPLPLRSLLELQLSLLGVGLNSTVEMPNGNEDIT-----PDFFLTATPSETLSVSSSLPLPEVQC---FVFNVEYMN
8	UniProt/Swiss-Prot	P26896	IL2RB_RAT	73.0%	9.3%	MATVDLSWRPLPLYILLLLLAT-----WVSAAVNDCSHLKC---FYNSRANVS
9	UniProt/Swiss-Prot	Q8BZM1	GLMN_MOUSE	27.0%	5.7%	-----
10	UniProt/Swiss-Prot	P36835	IL2_CAPHI	0.0%	0.0%	-----
11	UniProt/Swiss-Prot	Q7JFM4	IL2_AOTVO	0.0%	0.0%	-----
12	UniProt/Swiss-Prot	Q29416	IL2_CANFA	0.0%	0.0%	-----
	consensus/100%					.....
	consensus/90%					.....
	consensus/80%					.....
	consensus/70%					.....

Reference sequence (1): UniProt/Swiss-Prot|P26898|IL2RA\_SHEEP  
Identities normalised by aligned length.  
Colored by: identity



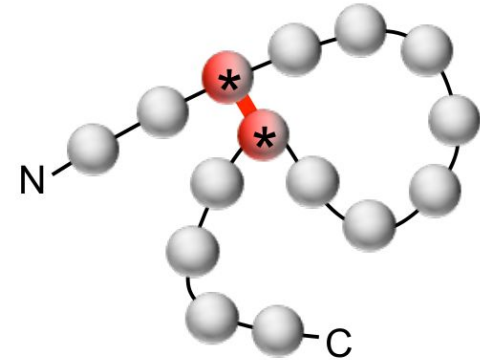
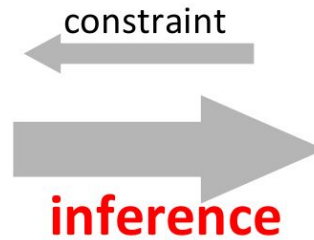
# Direct-coupling analysis of residue coevolution captures native contacts across many protein families



A	T	R	L	T	L	T	A	K	K	D	G	P	C	D
A	T	R	L	T	L	T	A	K	K	D	G	P	C	D
A	T	R	L	T	L	T	A	K	K	D	G	P	C	D
A	T	K	L	C	L	T	A	K	K	E	G	P	K	D
A	T	K	L	T	L	T	A	K	K	E	G	P	K	D
A	T	K	L	T	L	G	A	K	K	E	G	G	C	D
A	T	W	L	T	L	T	A	K	K	V	G	P	C	D
A	T	W	L	T	L	T	A	K	K	V	G	P	C	D



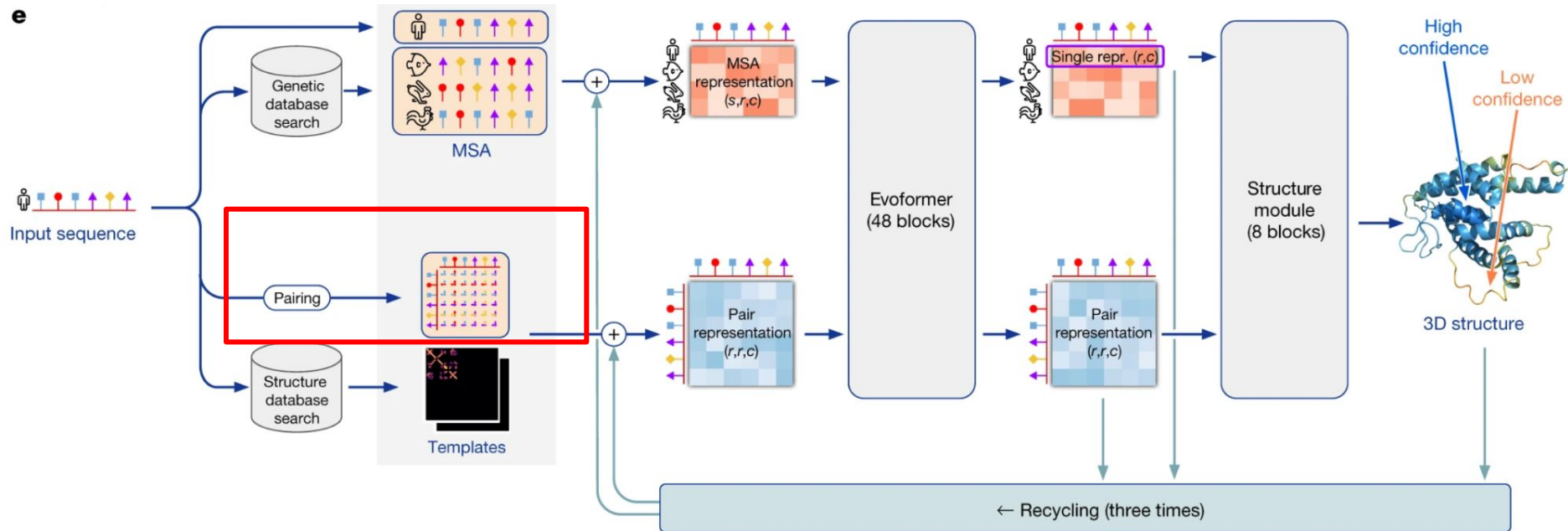
correlated



contact in 3D

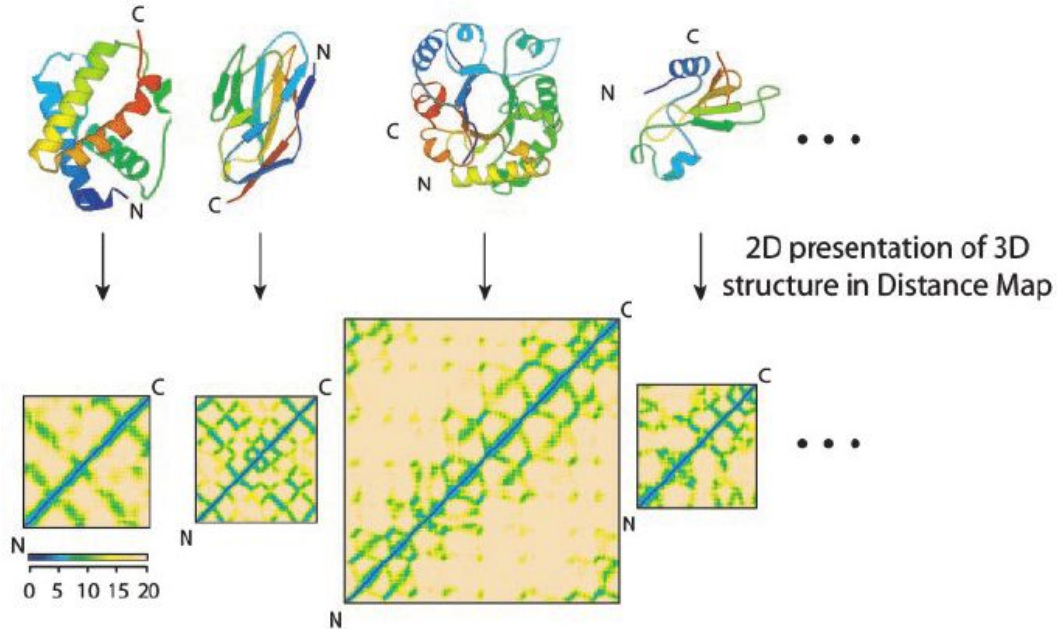


# Initializing the Pair Representation



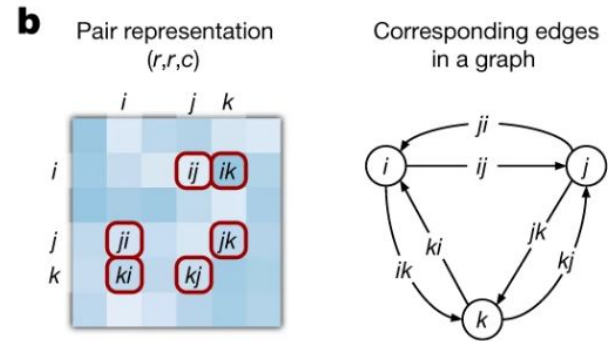


# What is a pair representation?



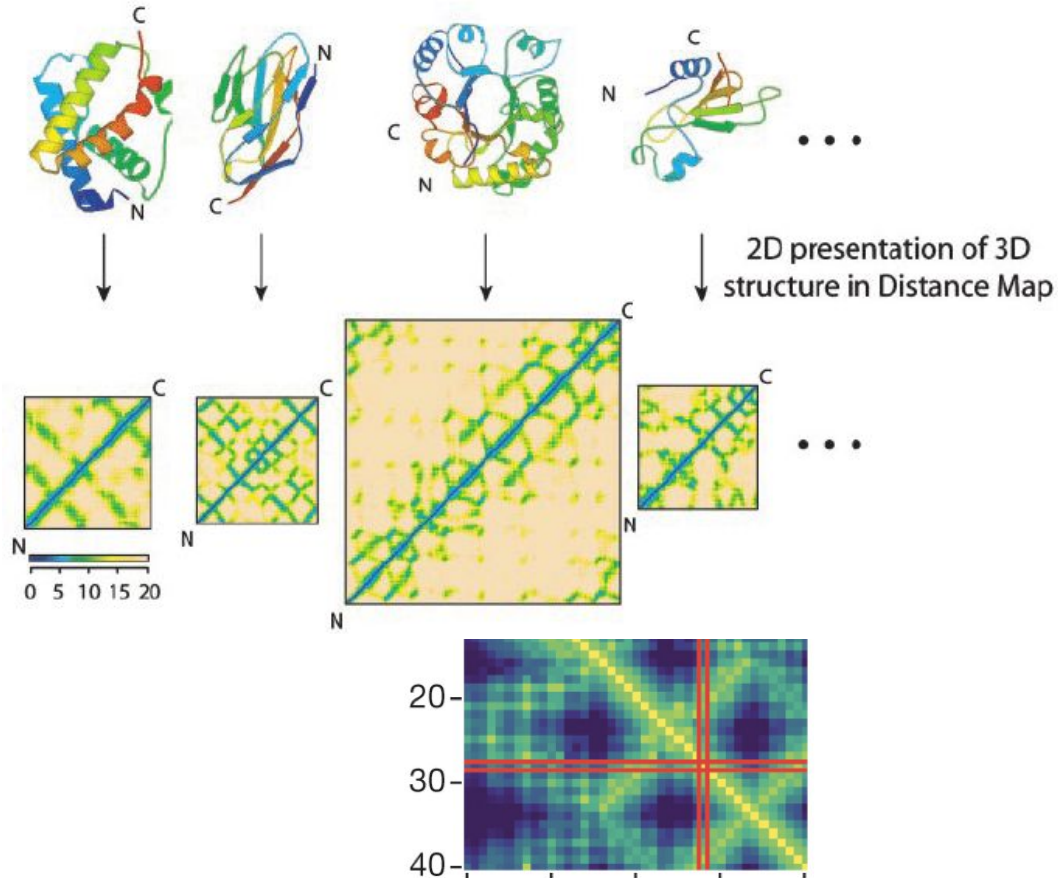
Distogram is used to map 2D pairwise distances

Distogram are independent of translations and rotations, so no need to align structures (much faster)



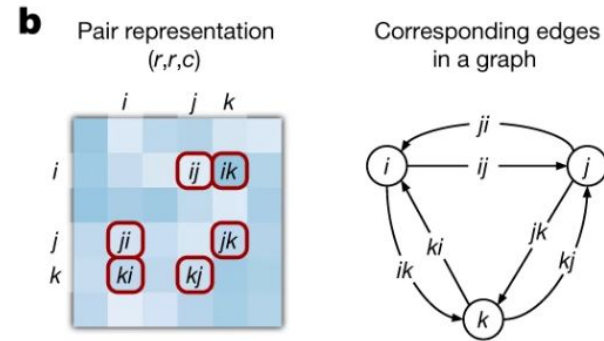


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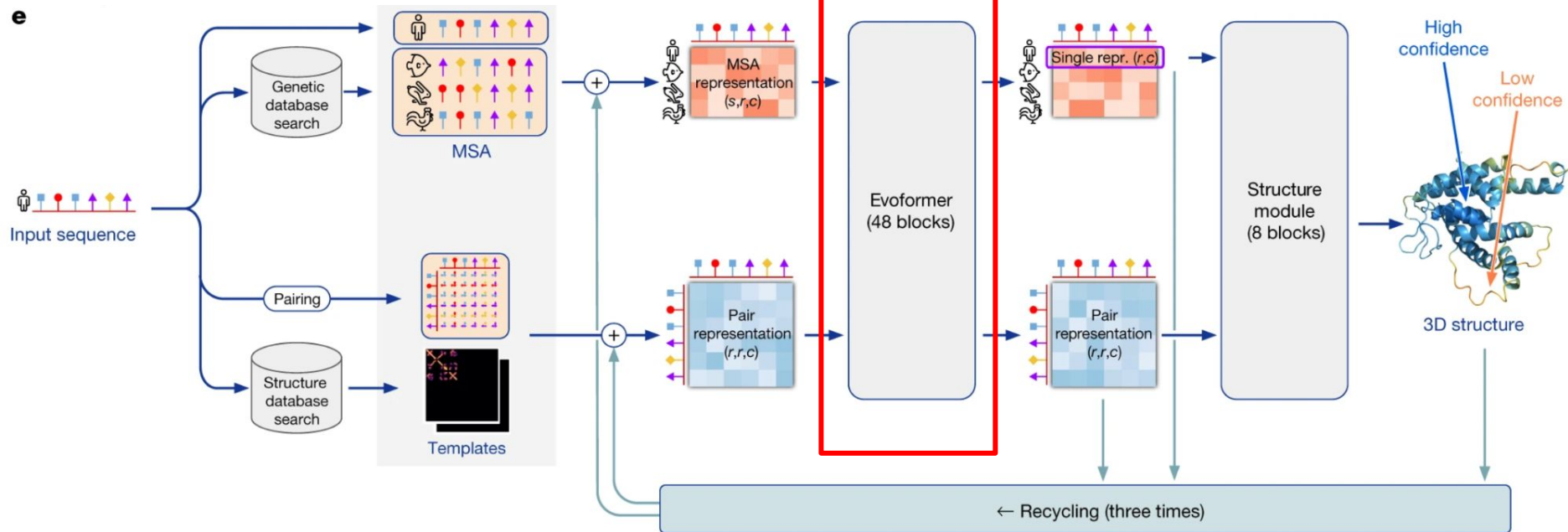
Distogram is used to map 2D pairwise distances

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# Transformer-like module updates representations





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# Attention Is All You Need

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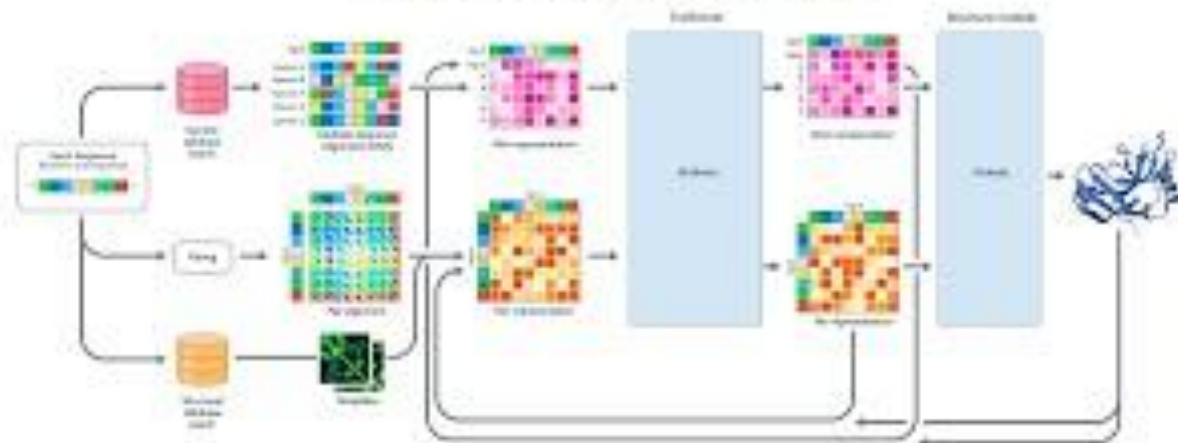
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**Illia Polosukhin\* ‡**  
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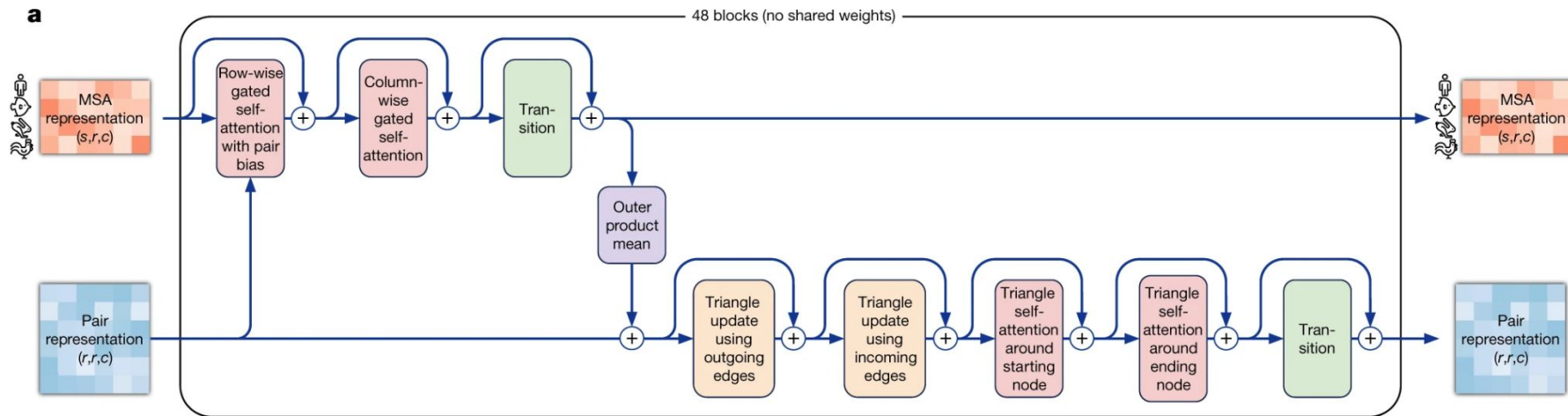


# What Is AlphaFold?



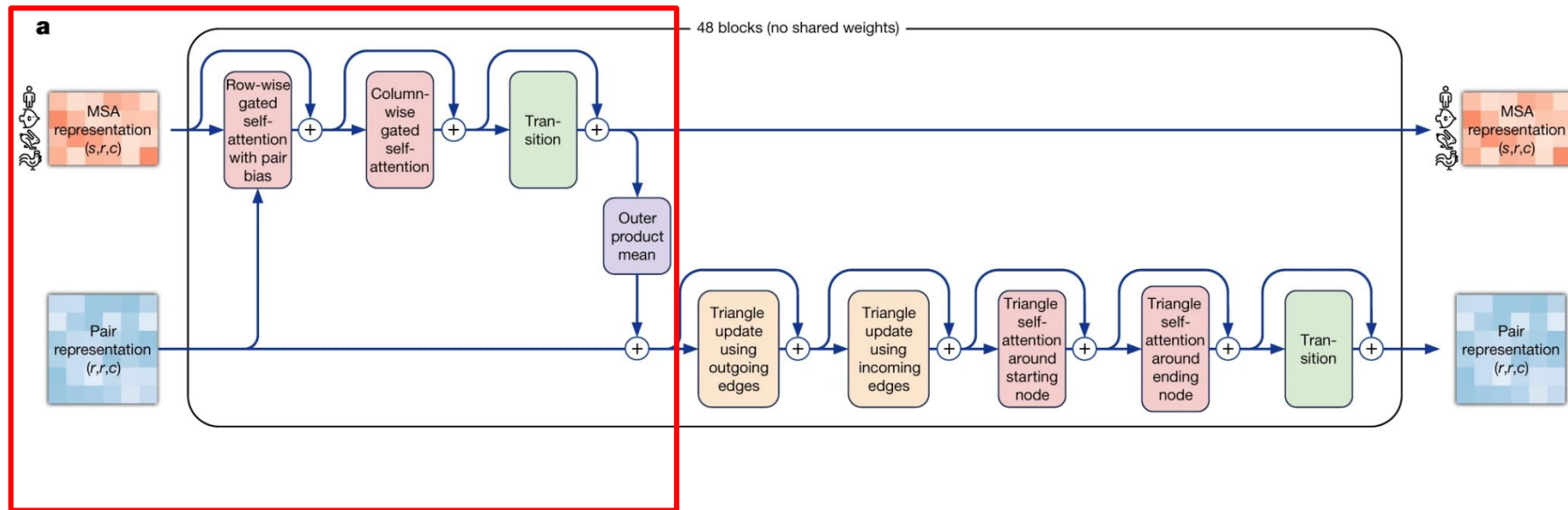


# The Evoformer: 48 blocks of back and forth between evolutionary and spatial reasoning.



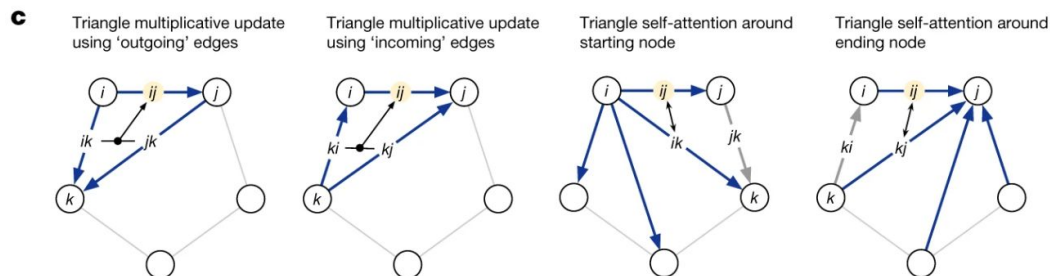
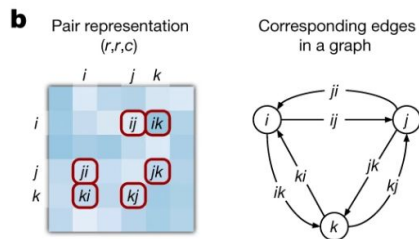
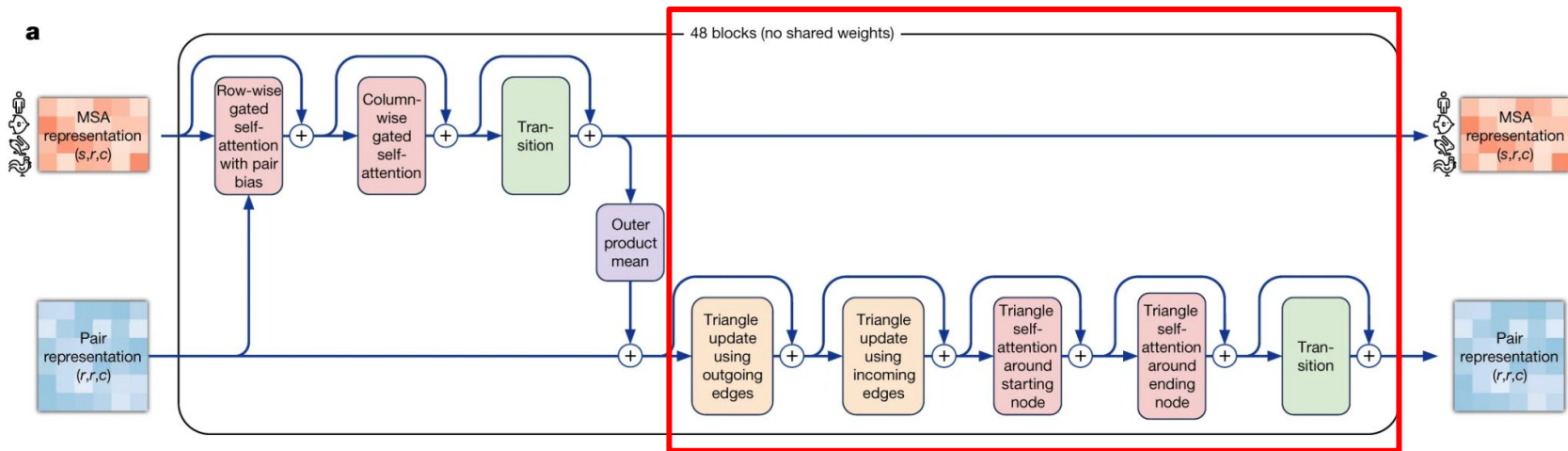


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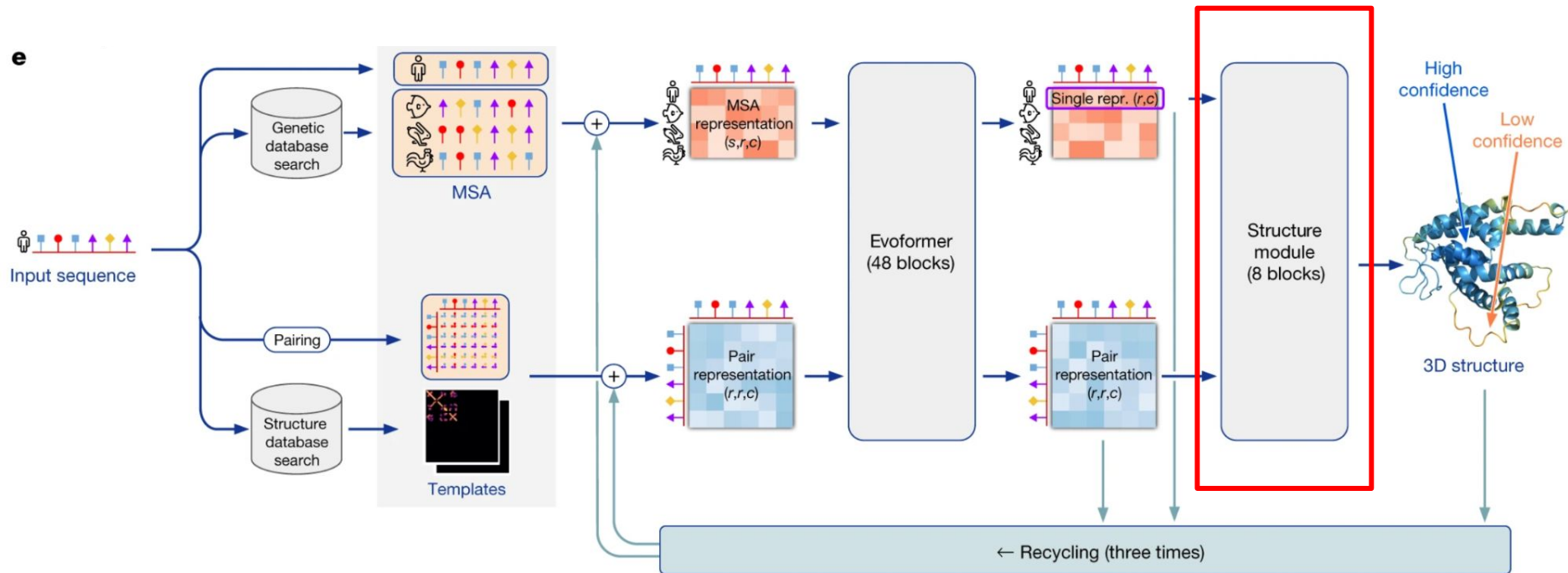


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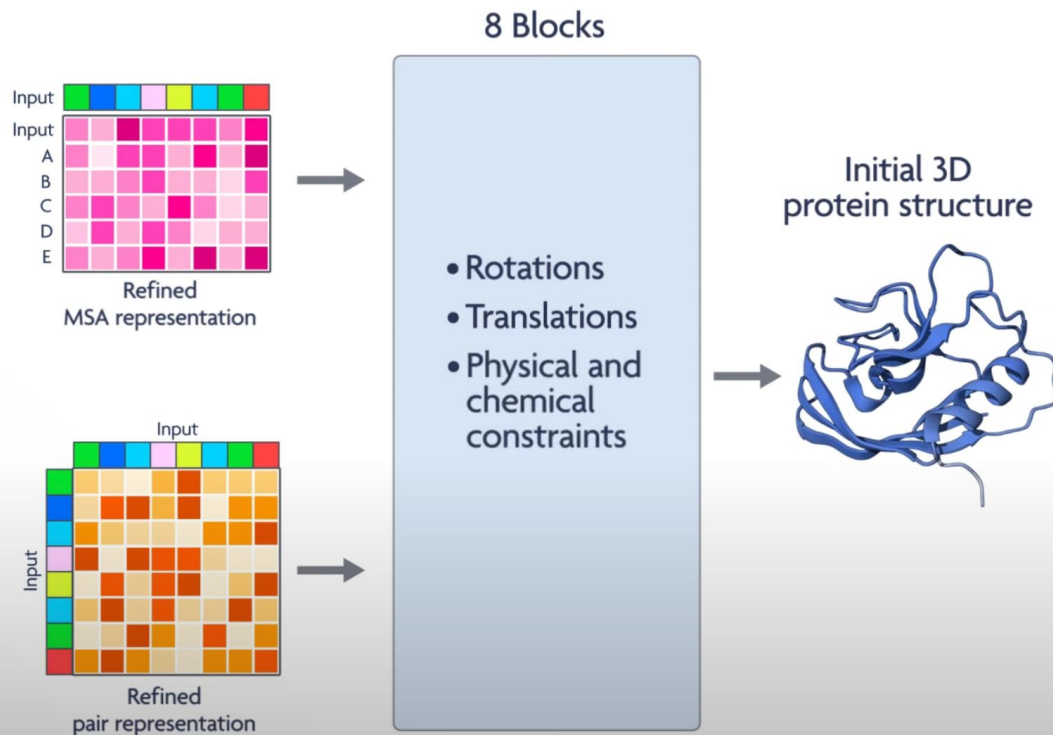


# Structure module converts representation to structure



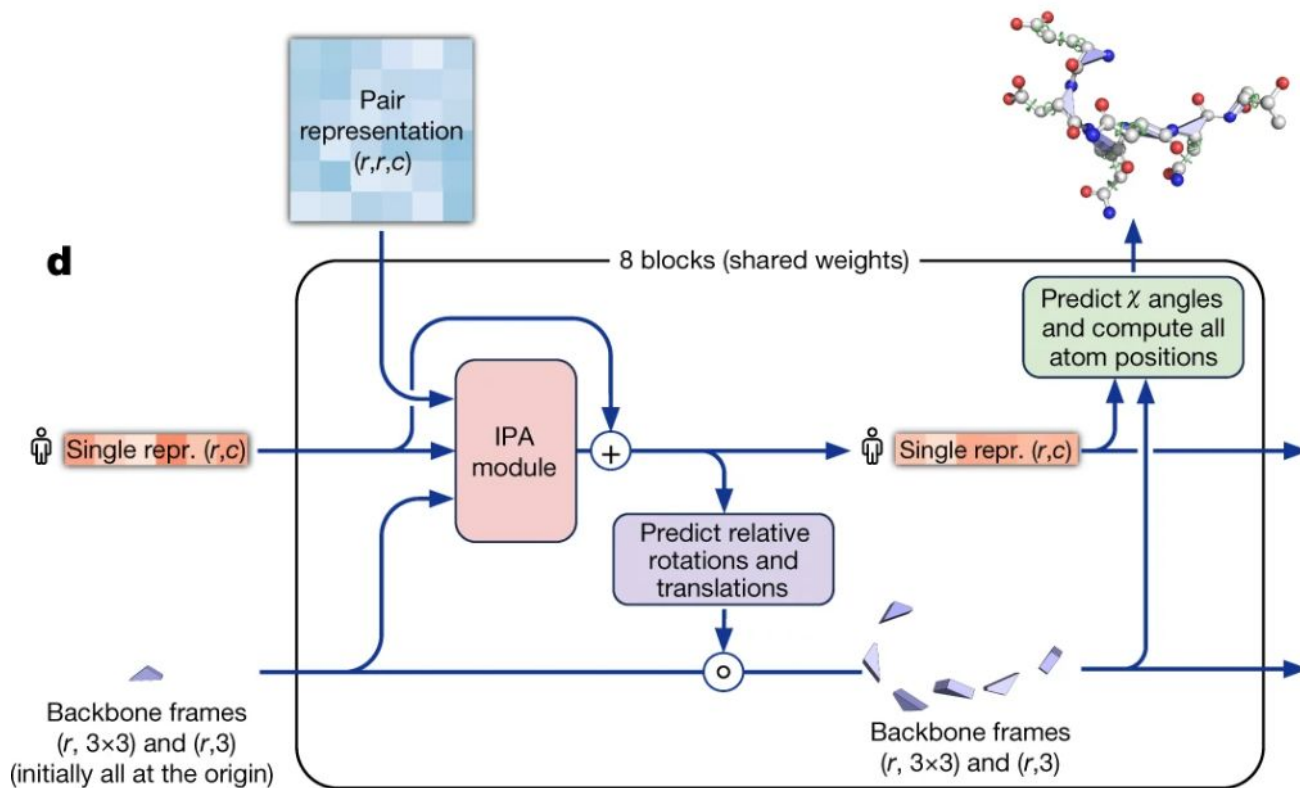


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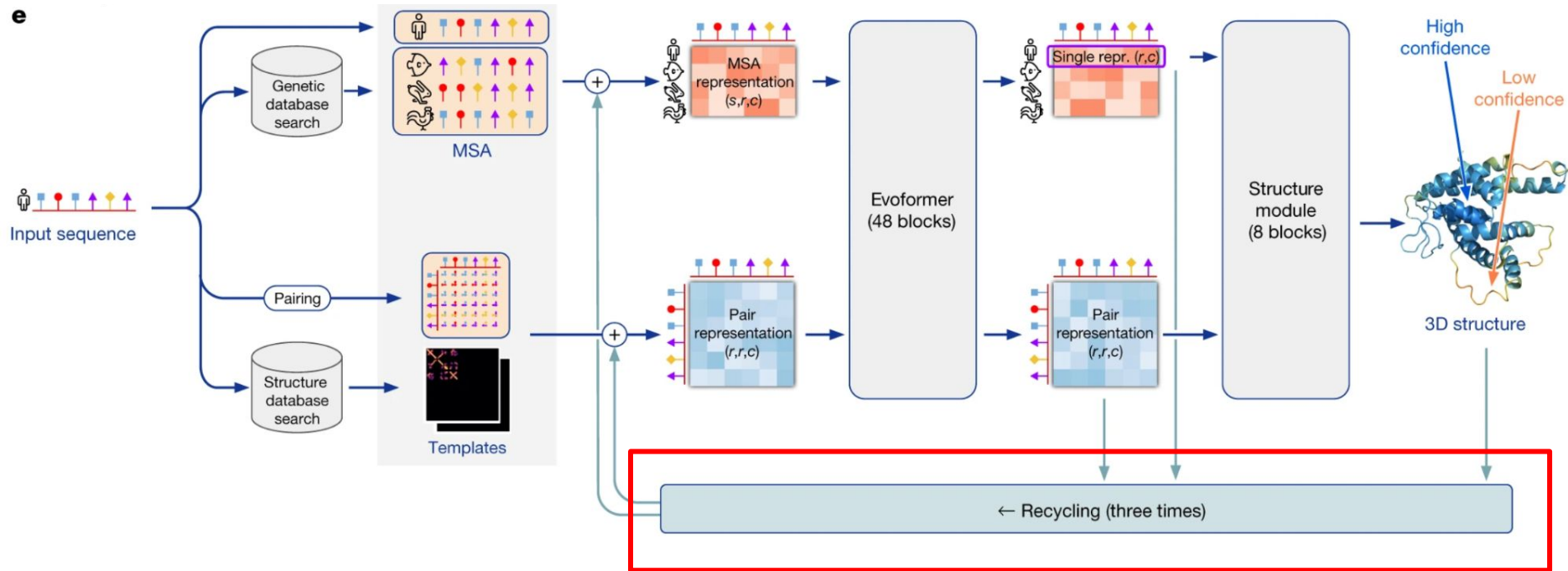


# The Structure Module: iteratively updating a residue gas



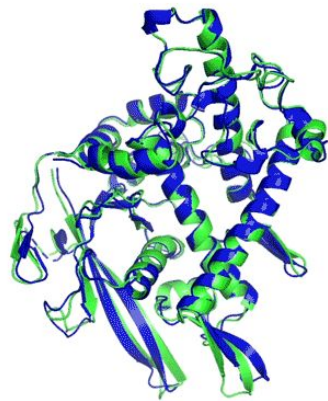


# Recycling iteratively refines structure

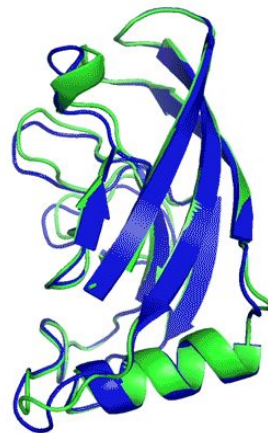




# How do we know it works? – ask the model!



**T1037 / 6vr4**  
90.7 GDT  
(RNA polymerase domain)



**T1049 / 6y4f**  
93.3 GDT  
(adhesin tip)

- Experimental result
- Computational prediction



Check in on Demo



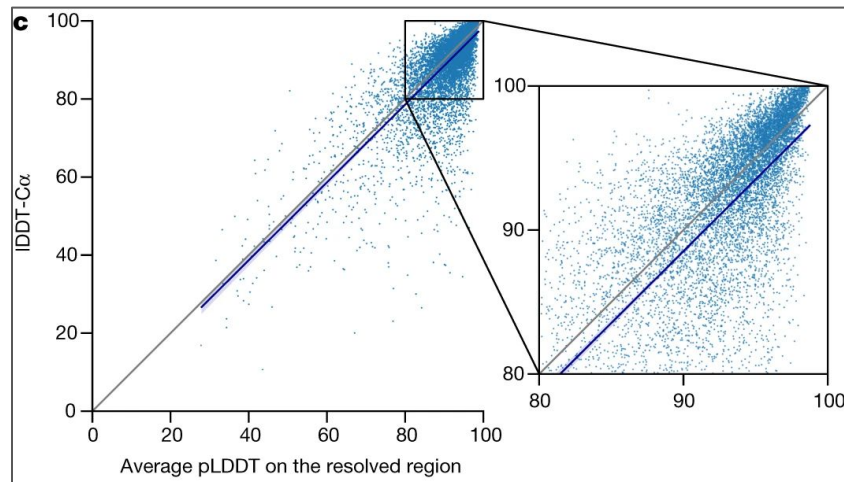
# The 3Ps: pLDDT, PAE, and pTM

- LDDT, AE, TM require ground truth. What can we do?
- pLDDT: predicted **L**ocal **D**istance **D**ifference **T**est
- PAE: **P**redicted **A**ligned **E**rror
- pTM: predicted **T**emplate **M**odeling score
  - Global comparison of similarity between two structures
  - Measure of 0 to 1



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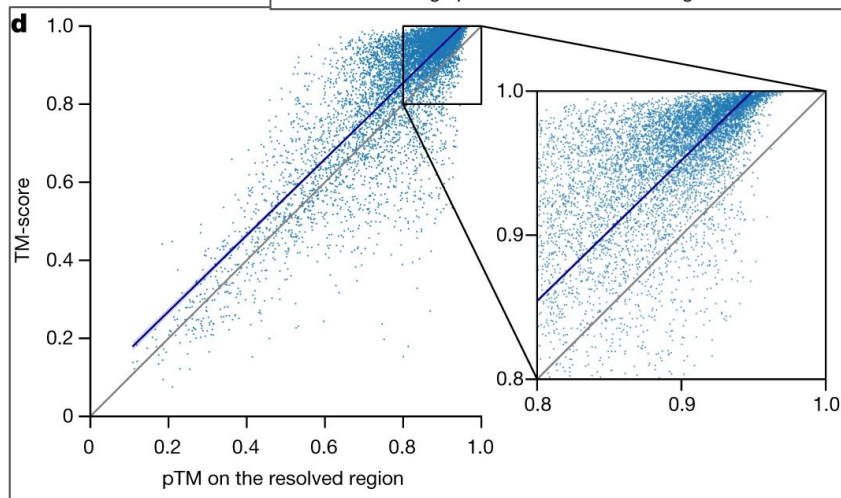
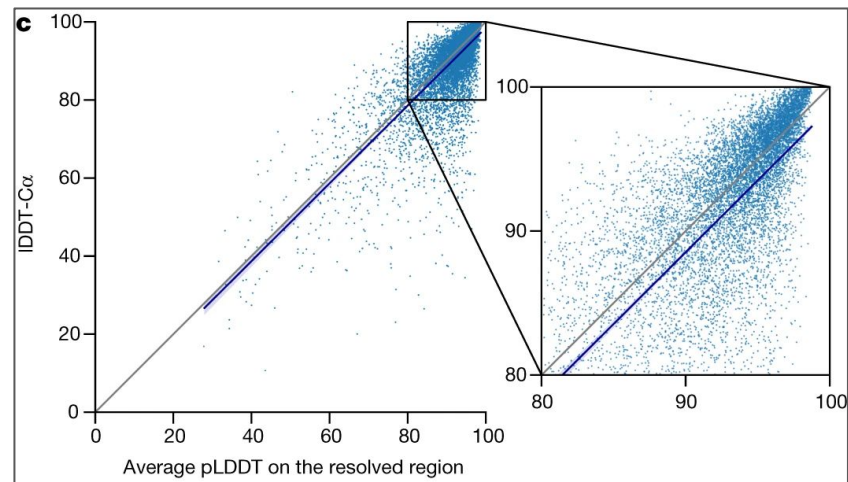
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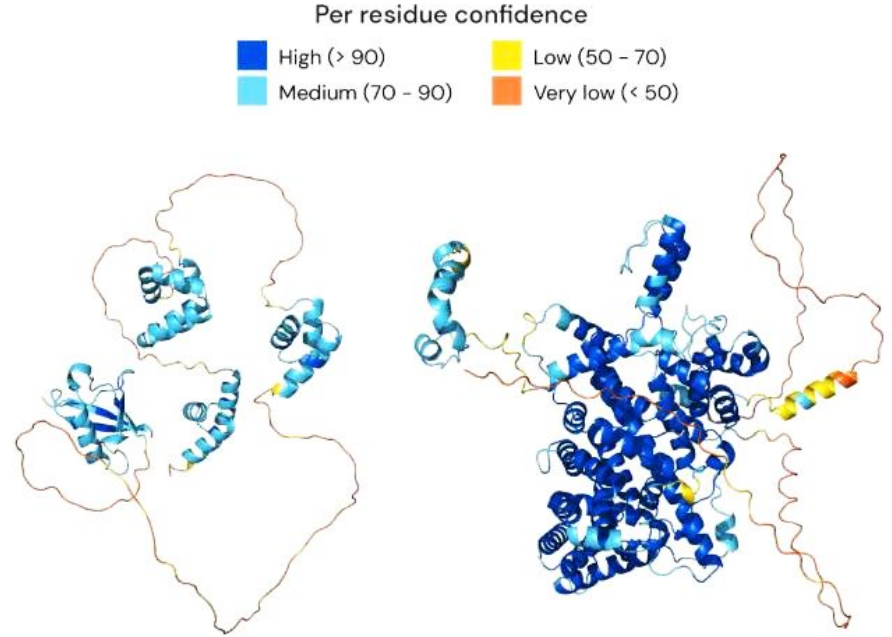
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# Predicted Local Distance Difference Test

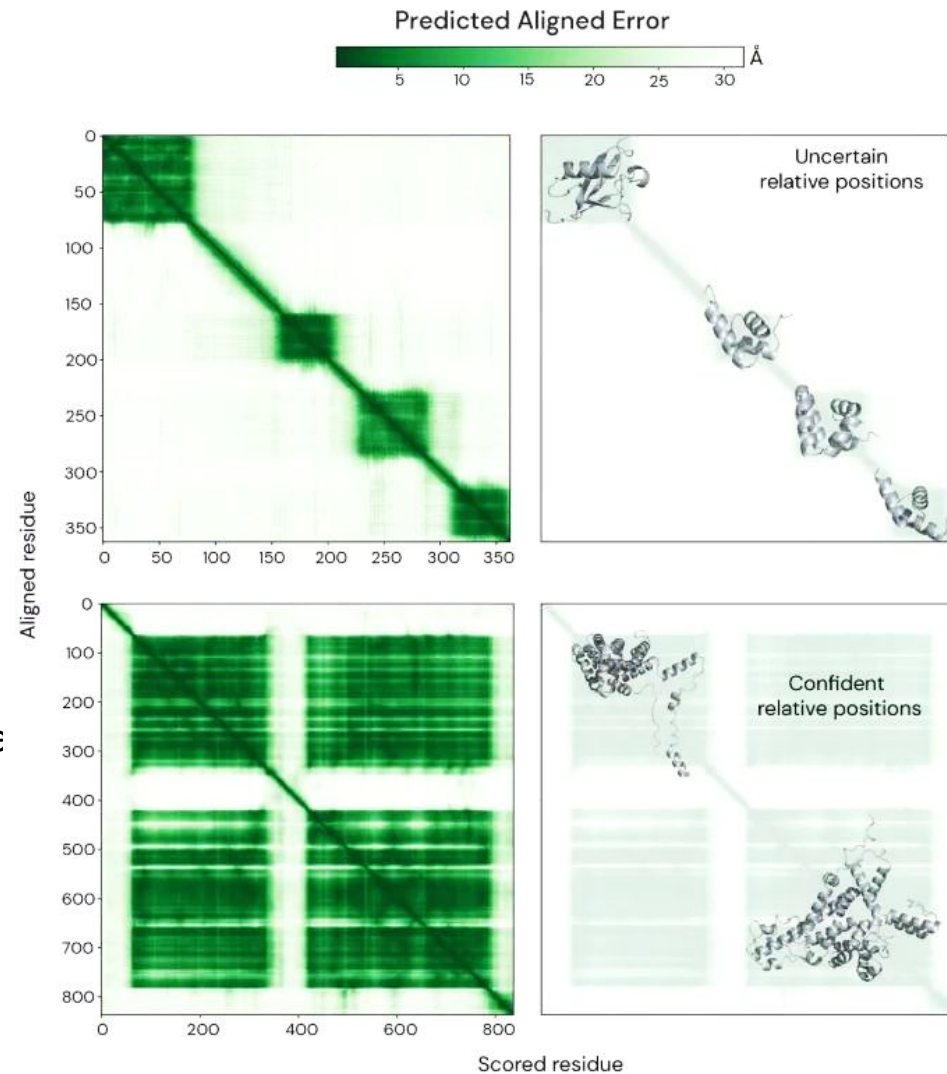
- AlphaFold's per-residue prediction of its IDDT-Ca score
- Low IDDT commonly associated with disorder
- High pLDDT on each domain doesn't imply confidence in relative positions!





# Predicted Aligned Error (PAE)

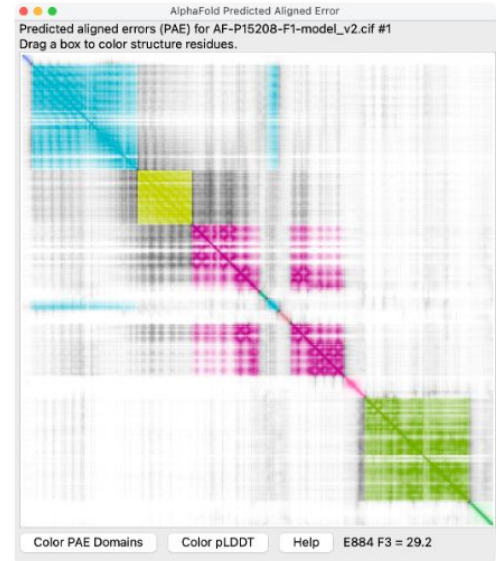
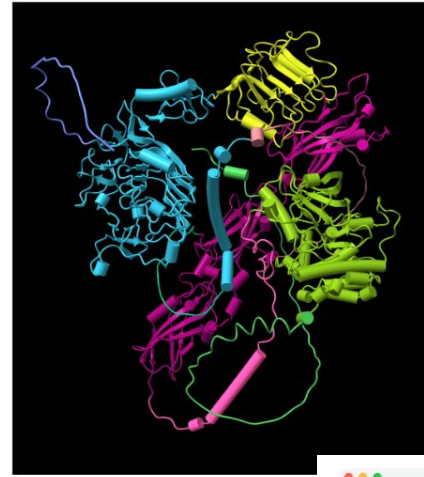
- Prediction of position error at residue x if the predicted and the true structures were aligned on y
- PAE aims to measure confidence in the relative positions of pairs of residues
- Use where pairwise confidence is relevant – interpreting domain distances in a multi domain protein
- Suppose residue y were aligned to the true structure and we measured the position error at residue x. The color at (x,y) is AF's prediction of that error





# Predicted Aligned Error (PAE)

- Prediction of position error at residue x if the predicted and the true structures were aligned on y
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# 214,683,839 Predicted Protein Structures on AFDB!

## AlphaFold Protein Structure Database

Developed by Google DeepMind and EMBL-EBI

Search for protein, gene, UniProt accession or organism or sequence search

BETA

Search

Examples: [MENFQKVEKIGEGTYGV...](#) [Free fatty acid receptor 2](#) [At1g58602](#) [Q5VSL9](#) [E. coli](#)

[See search help](#) [Go to online course](#)

PROTEIN DATA BANK



RCSB **PDB**  
PROTEIN DATA BANK



**217,705** Structures from the PDB



**1,068,577** Computed Structure  
Models (CSM)

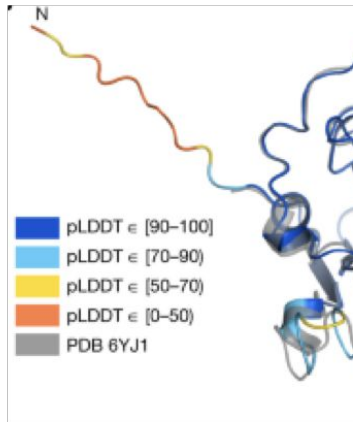
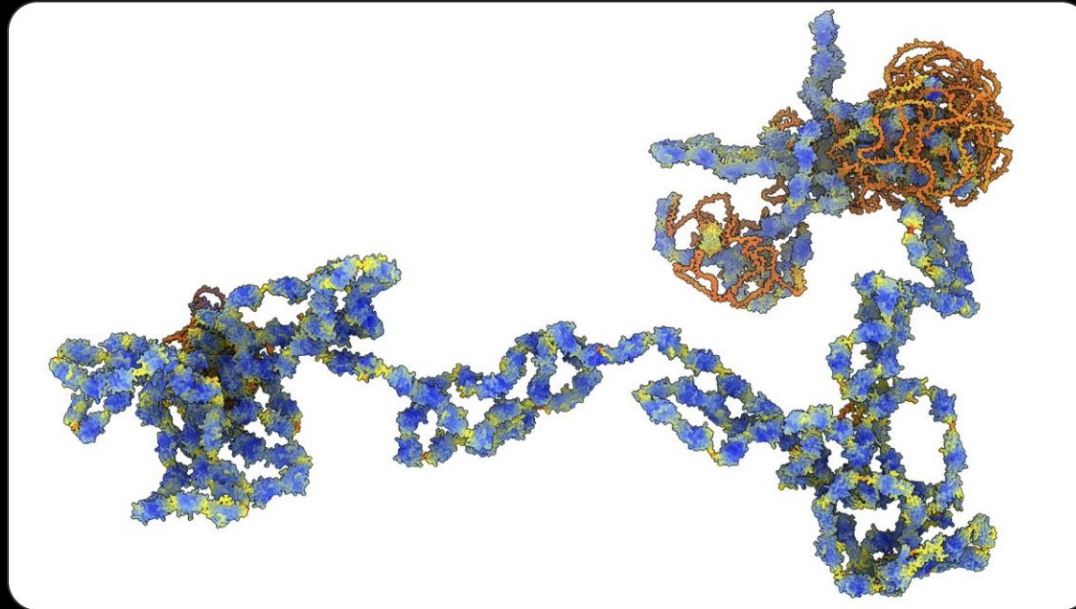


# Proteome



ChimeraX  
@UCSFChimeraX

The human muscle protein titin predicted by AlphaFold, 34350 residues.



ically  
tered

20%  
y

95%  
y

Dark proteome

Intrinsically disordered

PFAM – No PDB / AF

Gained AF – 70 > pLDDT ≥ 50

Gained AF – 90 > pLDDT ≥ 70

Gained AF – pLDDT ≥ 90

PDB 20% to 50% - pLDDT < 90

PDB 50% to 95%

PDB ≥ 95%

7:23 PM · Sep 8, 2021 · Twitter Web App

, Valentini, S., & Valencia, A.  
biology.



# 2024 Nobel Prize in Chemistry

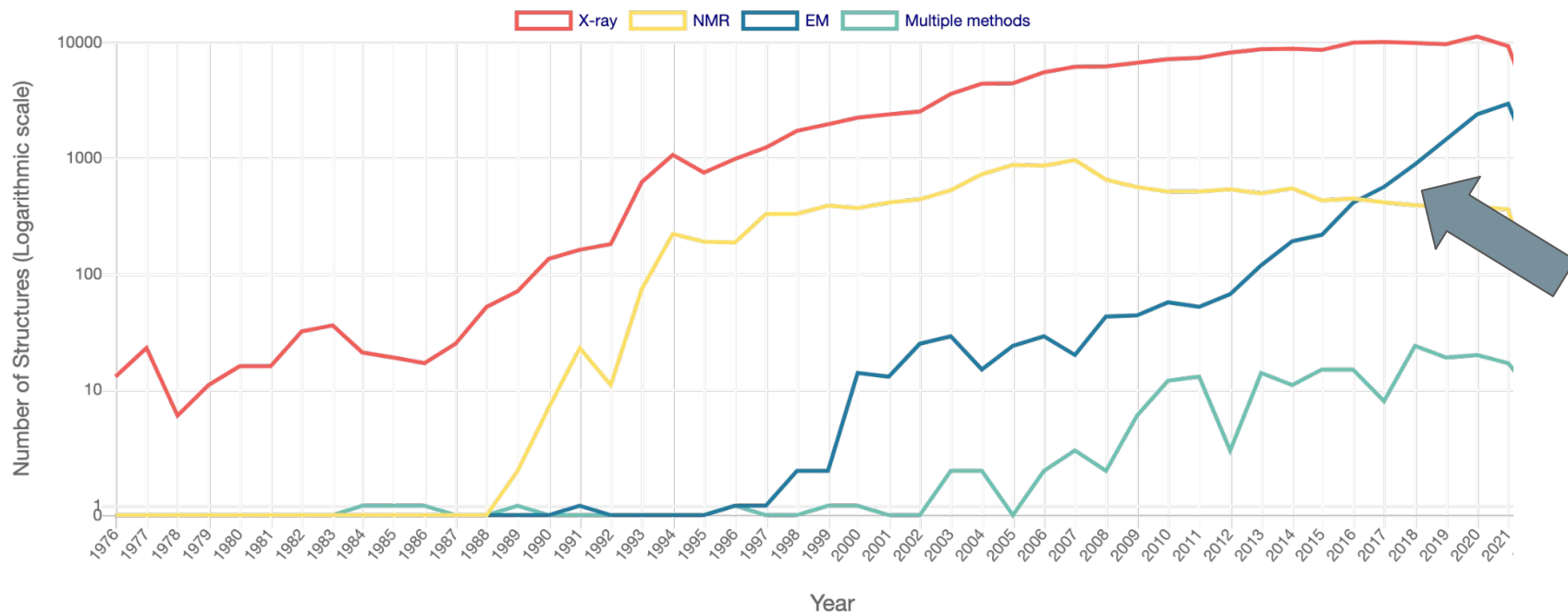
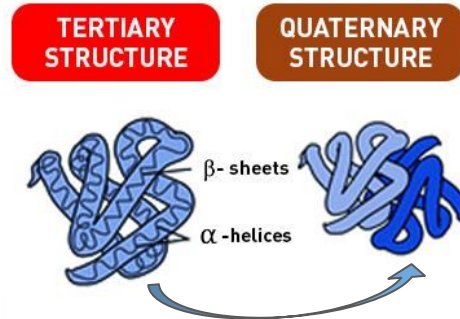


David Baker, Demis Hassabis and John Jumper (left to right) won the chemistry Nobel for developing computational tools that can predict and design protein structures. Credit: BBVA Foundation

<https://www-nature-com.ezproxy.lib.utexas.edu/articles/d41586-024-03214-7>

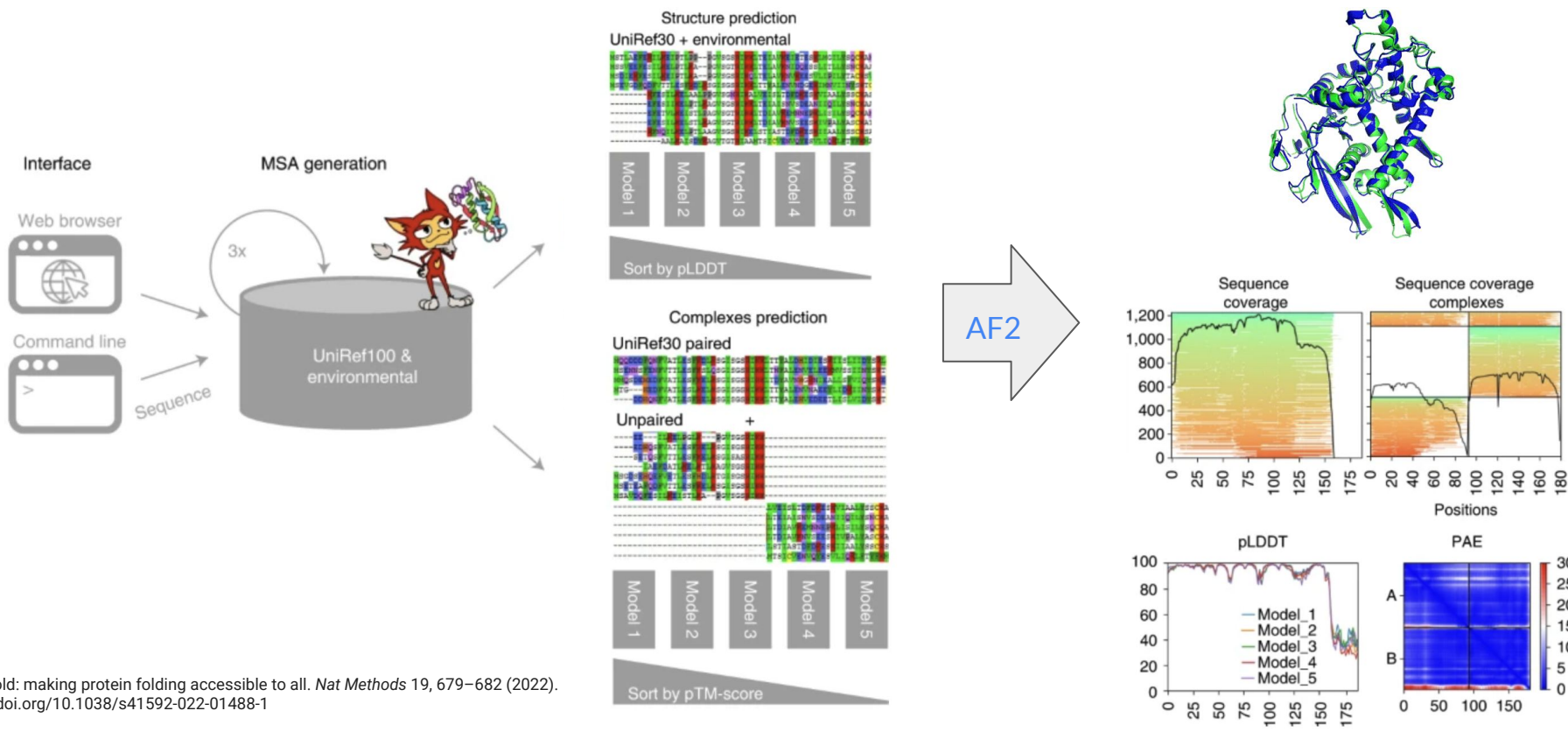


# AlphaFold Multimer!



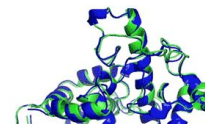
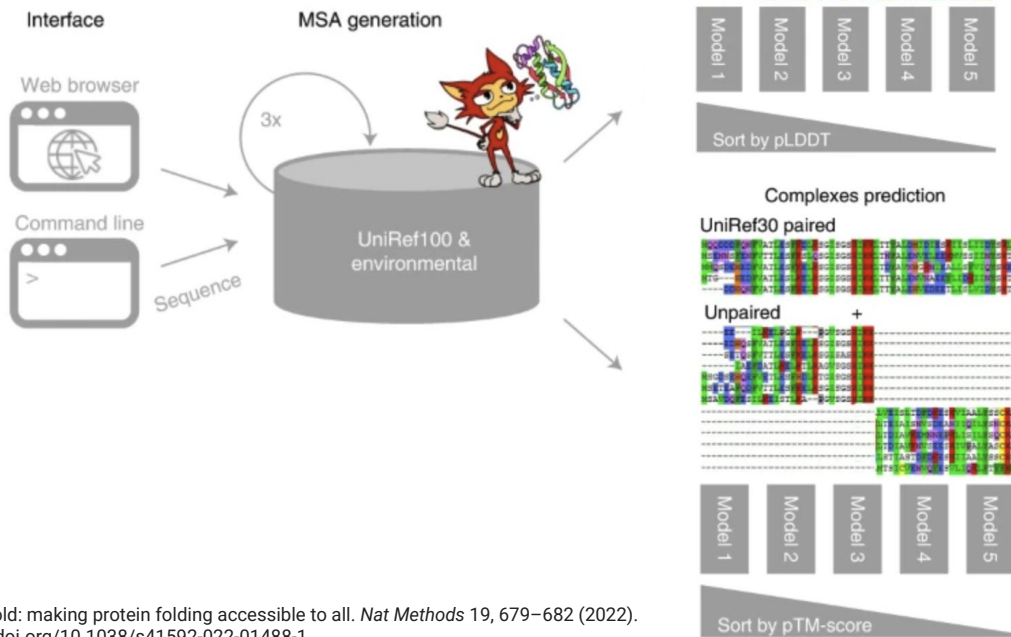


# ColabFold speeds up AlphaFold by 40-60 fold!

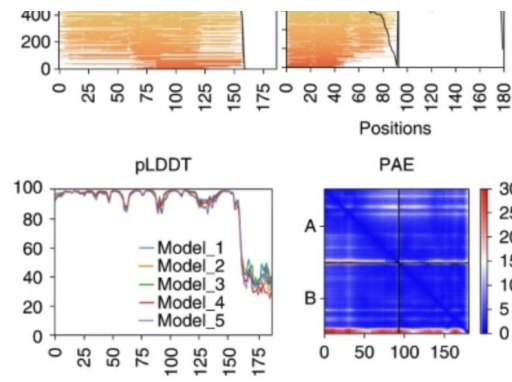




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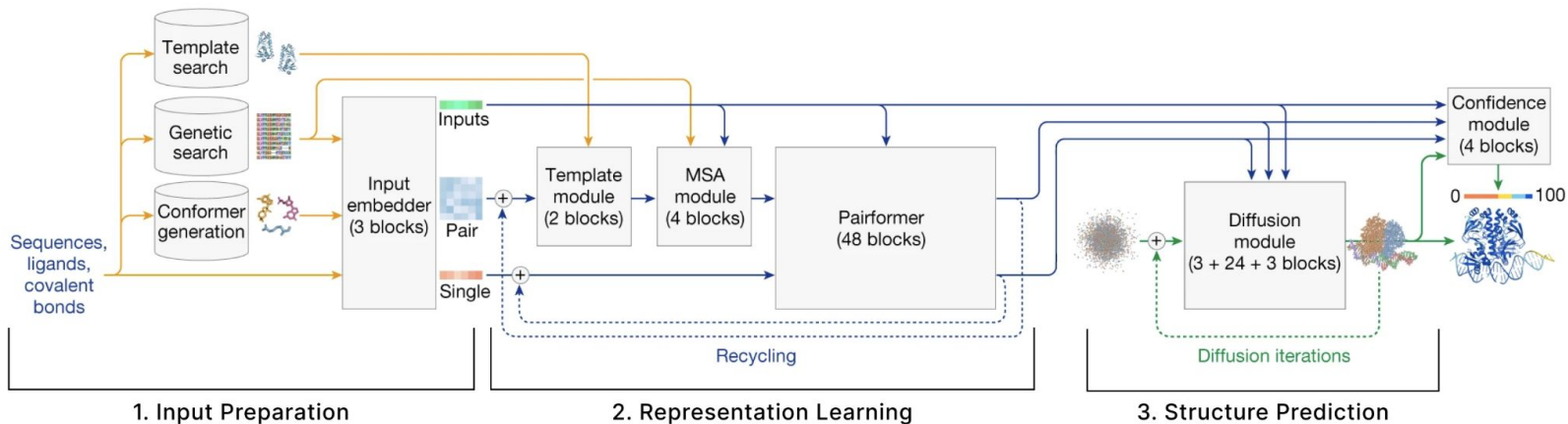


“ColabFold’s 40–60-fold faster search and optimized model utilization enables prediction of close to 1,000 structures per day on a server with one graphics processing unit.”






# A new era: All Atom Models (AF3)



Article | [Open access](#) | Published: 08 May 2024

## Accurate structure prediction of biomolecular interactions with AlphaFold 3

[Josh Abramson](#), [Jonas Adler](#), [Jack Dunger](#), [Richard Evans](#), [Tim Green](#), [Alexander Pritzel](#), [Olaf Ronneberger](#), [Lindsay Willmore](#), [Andrew J. Ballard](#), [Joshua Bambrick](#), [Sebastian W. Bodenstein](#), [David A. Evans](#), [Chia-Chun Hung](#), [Michael O'Neill](#), [David Reiman](#), [Kathryn Tunyasuvunakool](#), [Zachary Wu](#), [Akvilė Žemgulytė](#), [Eirini Arvaniti](#), [Charles Beattie](#), [Ottavia Bertolli](#), [Alex Bridgland](#), [Alexey Cherepanov](#), [Miles Congreve](#), ... [John M. Jumper](#)  [+ Show authors](#)

[Nature](#) **630**, 493–500 (2024) | [Cite this article](#)

**800k** Accesses | **2091** Altmetric | [Metrics](#)

## The Illustrated AlphaFold

A visual walkthrough of the AlphaFold3 architecture, with more details and diagrams than you were probably looking for.

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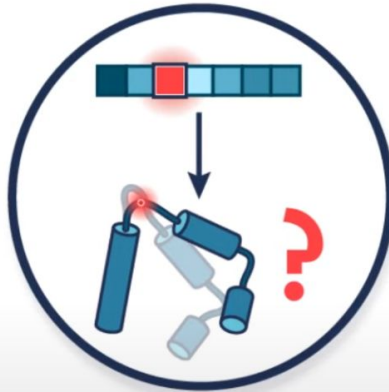


# Applications and Frontiers!

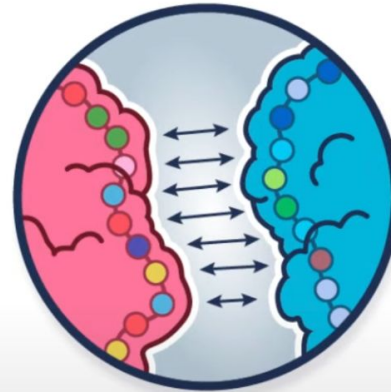
Discovering  
drugs



Effect of  
genetic variants



Modeling  
protein–protein  
interactions



Engineering  
artificial  
proteins





# Has protein structure prediction been solved?

- Sort of.
- The Protein Folding Problem (de novo)
  - What is the folding code?
  - What is the folding mechanism?
  - Can we predict a native protein structure from its primary, amino acid sequence?
    - No for a sequence in isolation...
    - Yes when informed by like sequences and their structures



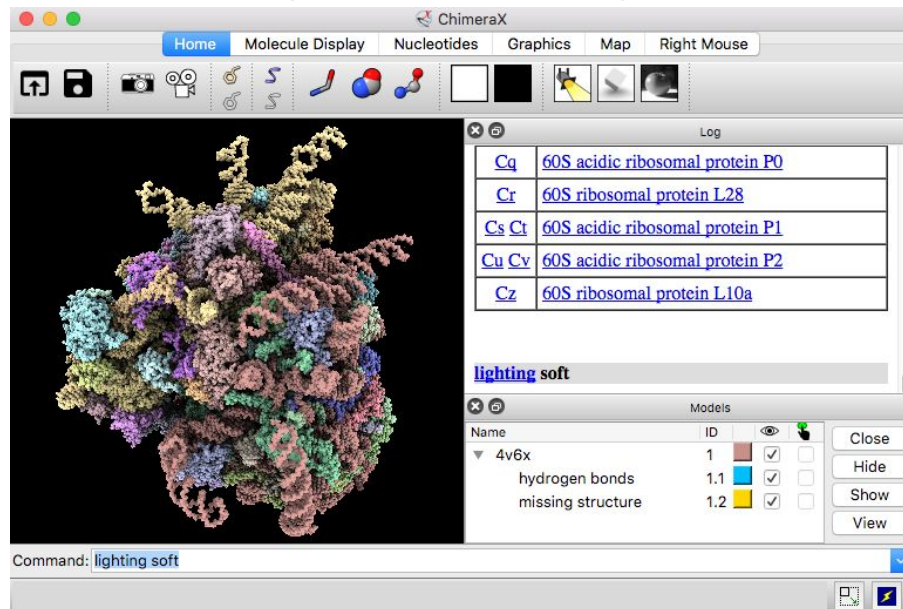
# Resources & Useful Links

- Prediction Servers/Colabs:
  - [AF3 Server](#)
  - [ColabFold](#) (AF2 w/MMSeqs2)
  - Maintained list of Google Colabs: <https://github.com/sokrypton/ColabFold>
- AlphaFold Resources:
  - Github page: <https://github.com/google-deepmind/alphafold>
  - AFDB Protein Structure Database and links: <https://alphafold.com>
  - [The Illustrated AlphaFold](#)
  - [Lovely & more in depth lecture from John Jumper at Vanderbilt University](#)
  - Running AlphaFold on the BRCF Servers:
    - [AMD GPUs](#)
    - [NVIDIA GPUs](#)
  - [Running AlphaFold on TACC](#)
  - [AlphaFold2 & Equivariance](#)



# ChimeraX

- Download ChimeraX: <https://www.cgl.ucsf.edu/chimerax/download.html>
- Quick Start: <https://www.cgl.ucsf.edu/chimerax/docs/quickstart/index.html>
- Very comprehensive user guide: <https://www.cgl.ucsf.edu/chimerax/docs/user/index.html>





# Thank you! Questions?



Looking for people to work with/learn more about Machine Learning applied to Biology?

<https://www.biomlsociety.org>

We meet every other Thursday from 11am to noon, in MBB 3.204

**TACOS** and **COFFEE** provided!

Watch the Commander Complex assemble!

