

how artificial intelligence is reshaping protein structure prediction and therapeutic design

Jens Meiler

Humboldt Professor and Director, Institute for Drug Discovery, University Leipzig, Leipzig, Saxony, Germany
Distinguished Professor, Department of Chemistry and Pharmacology, Vanderbilt University, Nashville, TN, USA



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Nobelpreis Chemie 2024



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The Nobel Prize in Chemistry 2024 was divided, one half awarded to David Baker “for computational protein design”, the other half jointly to Demis Hassabis and John M. Jumper “for protein structure prediction”.



Ill. Niklas Elmehed © Nobel Prize
Outreach
David Baker
Prize share: 1/2



Ill. Niklas Elmehed © Nobel Prize
Outreach
Demis Hassabis
Prize share: 1/4



Ill. Niklas Elmehed © Nobel Prize
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John M. Jumper
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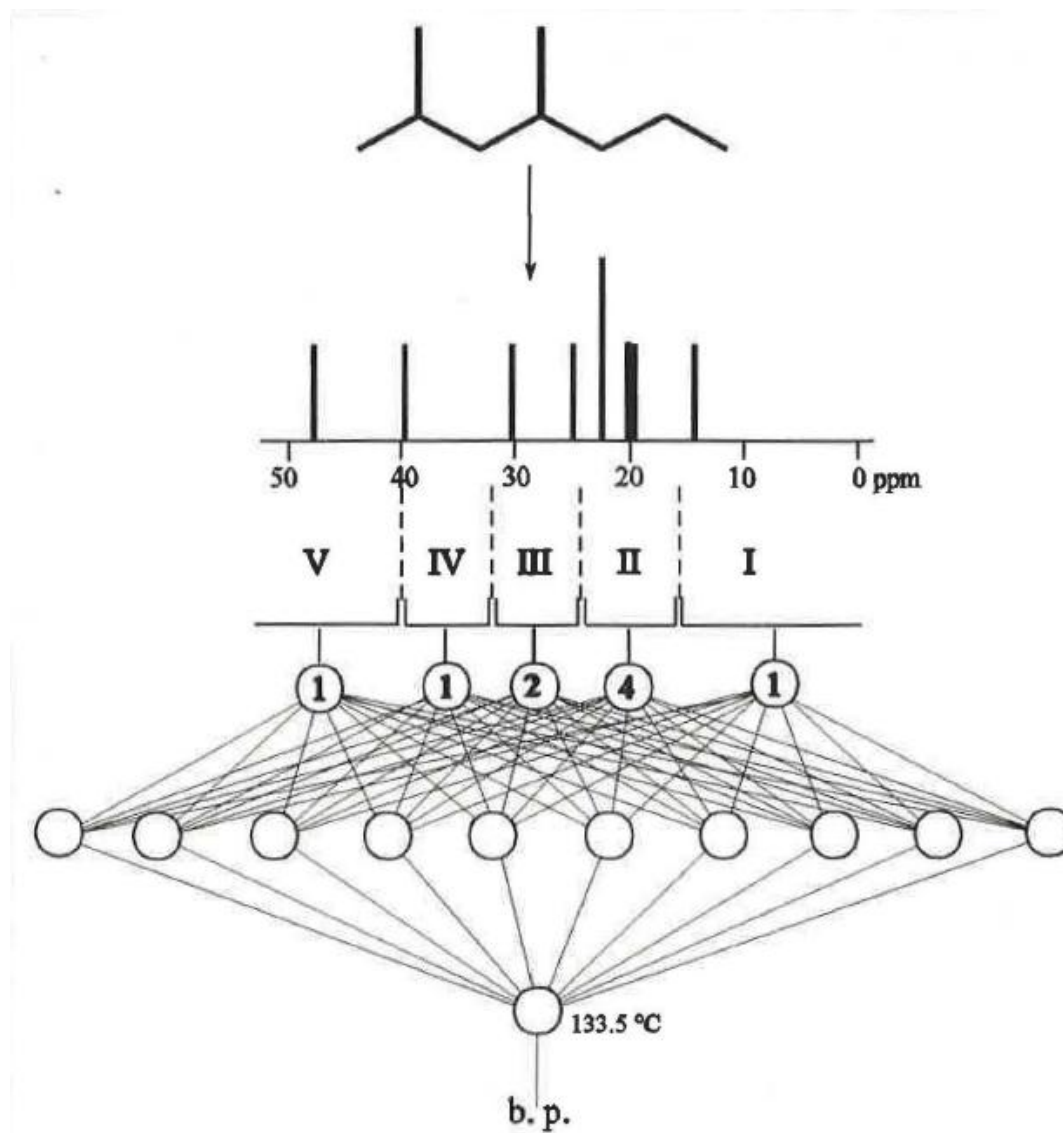
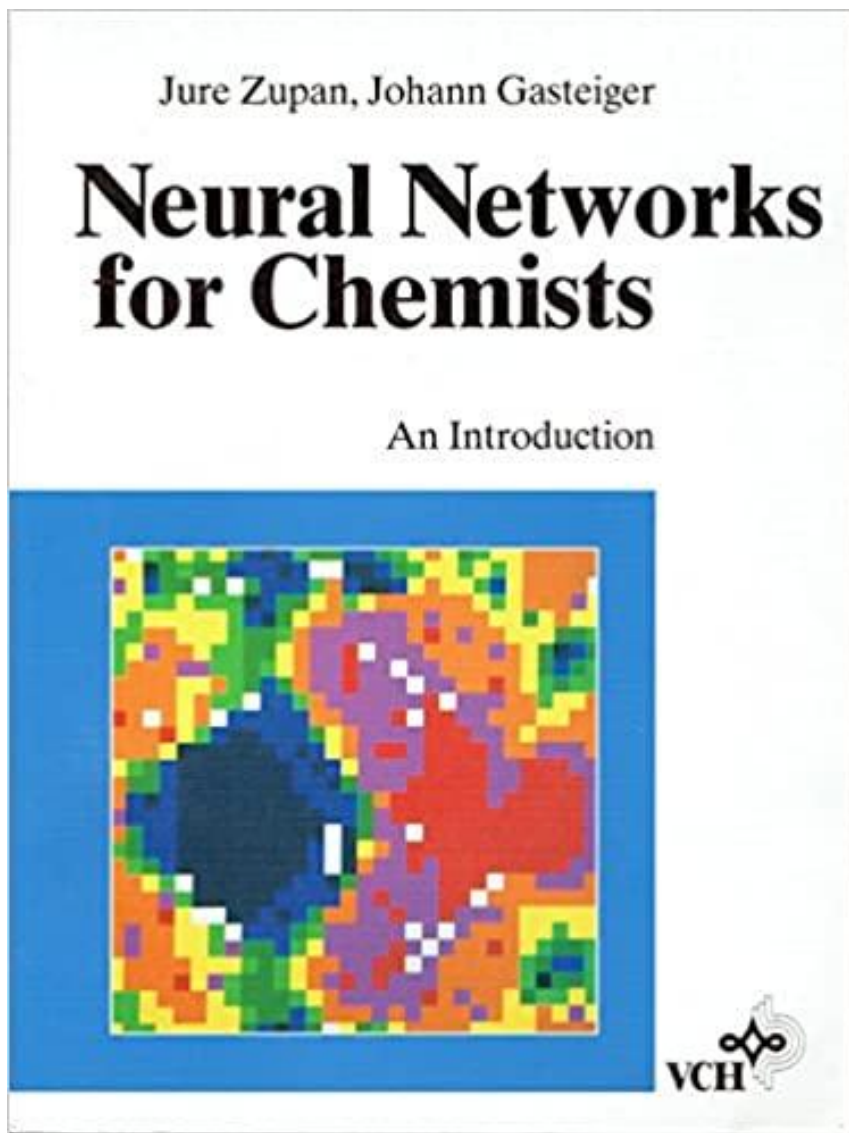
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Since 1993(!) – Neural Networks for Chemists



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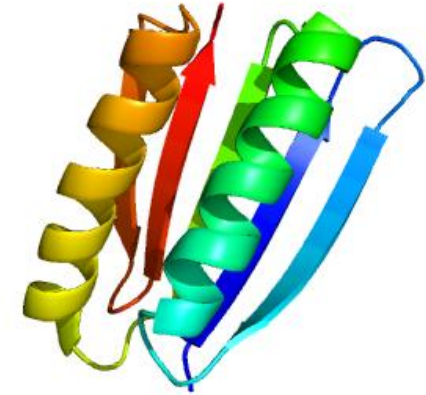
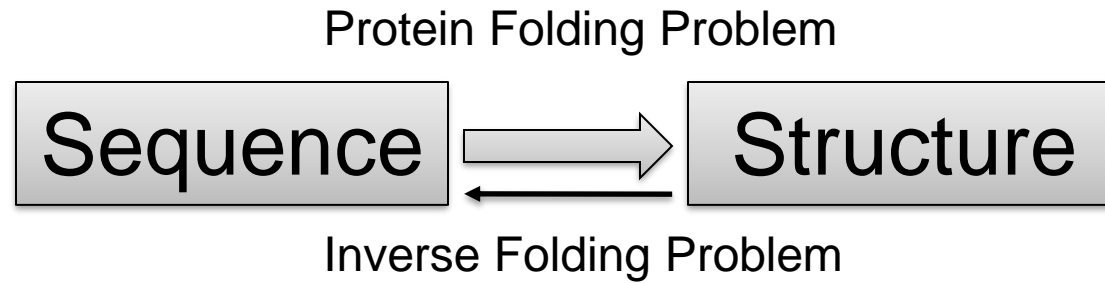
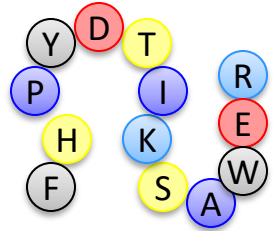


J. Meiler and R. Meusinger "Use of Neural Networks to Determine Properties of Alkanes from their ^{13}C -NMR Spectra" in *Software - Entwicklung in der Chemie*; Gasteiger, J., Ed. Gesellschaft Deutscher Chemiker: Frankfurt am Main; **1995**; Vol. 10: p. 259-263.



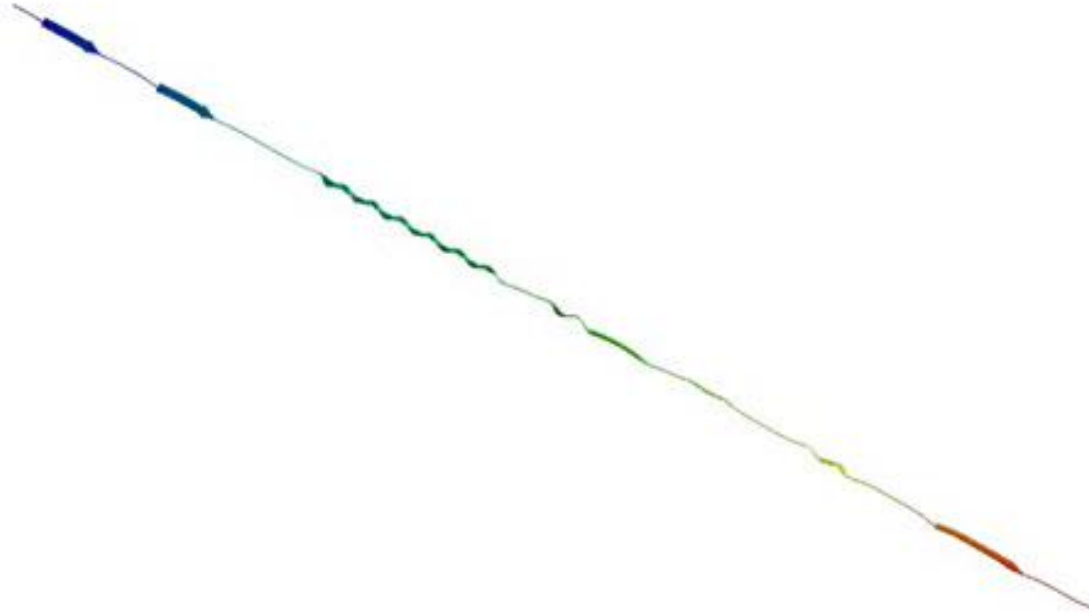
The (Inverse) Protein Folding Problem


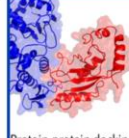


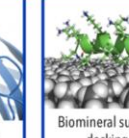
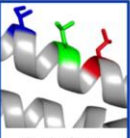
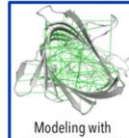

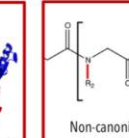
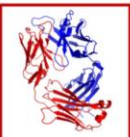

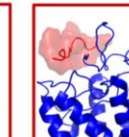
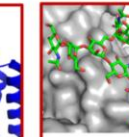





Holy Grail of Comp. Structural Biology



- Given a protein's AA sequence, what is its 3-dimensional fold, and how does it get there?
- Assume 100 conformations for each amino acid in a 100 amino acid protein $\Rightarrow 10^{200}$ possible conformations!
- Cyrus Levinthal's paradox of protein folding, 1968.

Rosetta: A Unified Framework for Protein Structure Prediction and Design



 Structure prediction	 Protein-protein docking	 Ligand docking	 Loop modeling	 Biomaterial surface docking
 Protein design	 Modeling with experimental data	Tasks	 Symmetric assemblies	 Non-canonical chemistries
 Antibodies	 Membrane proteins	Systems	 RNA / DNA	 Peptides
 Carbohydrates	 Peptides	 RNA / DNA	 Peptides	 Carbohydrates


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A Team Approach

Close collaboration between the labs the norm, even within single code modules. This allows for rapid enhancements and promotes the values of team science.

 Rosetta Software: The premier suite for macromolecular modelling

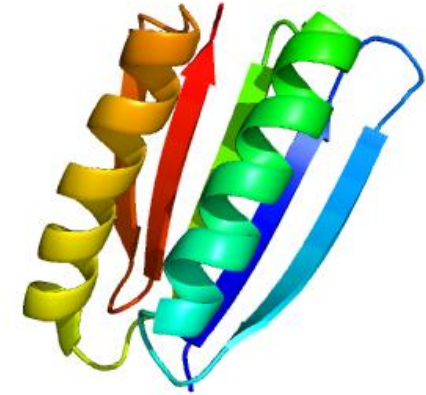
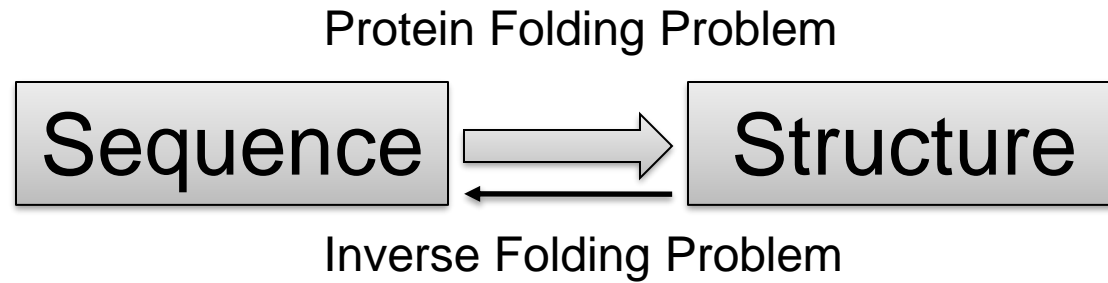
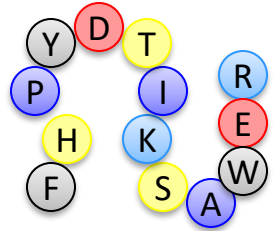
 Rosetta Commons: An Innovative Model for Collaboration

 Rosetta News



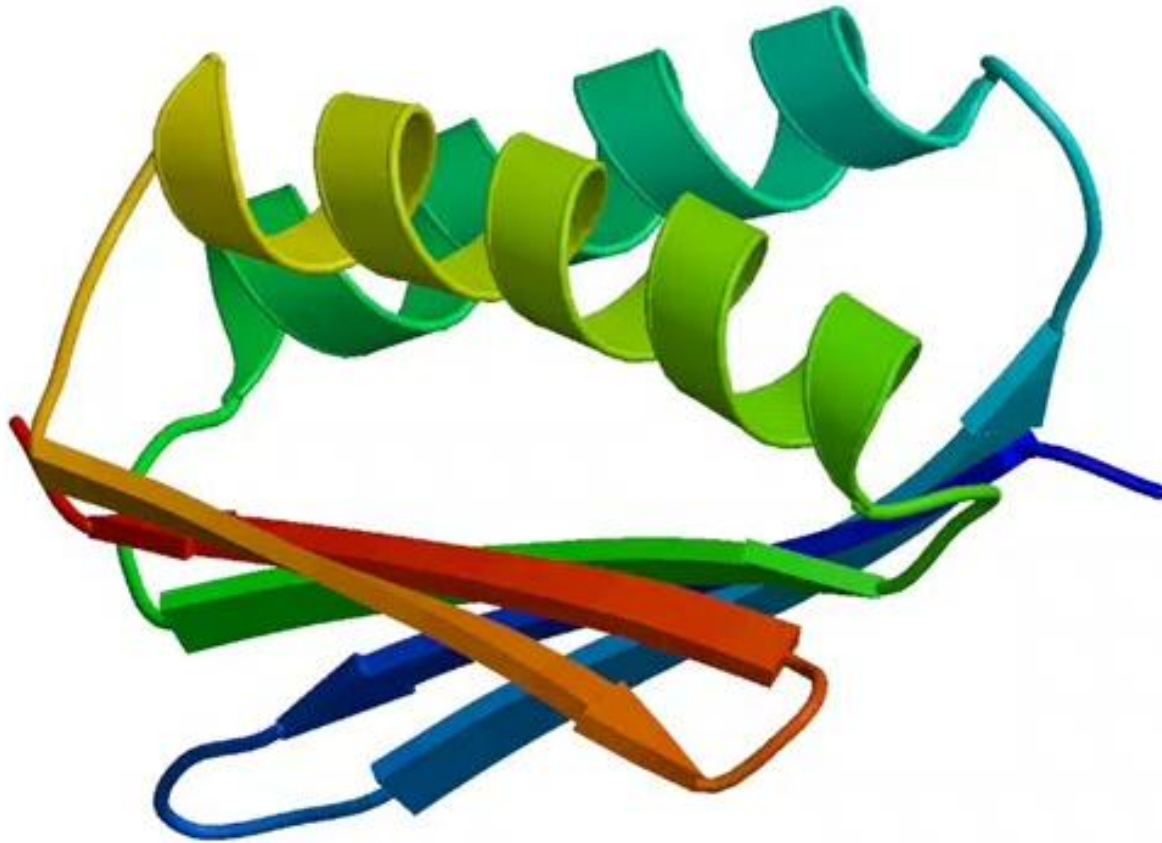
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- Assume 100 conformations for each amino acid in a 100 amino acid protein $\Rightarrow 10^{200}$ possible conformations!
- Cyrus Levinthal's paradox of protein folding, 1968.
- Given a protein fold, which primary sequence(s) fold into it?
- Assume a total of 100 conformations for all 20 natural occurring amino acids side chains in a 100 amino acid protein $\Rightarrow 10^{200}$ possible conformations!
- Earth is less than 10^{10} years old.

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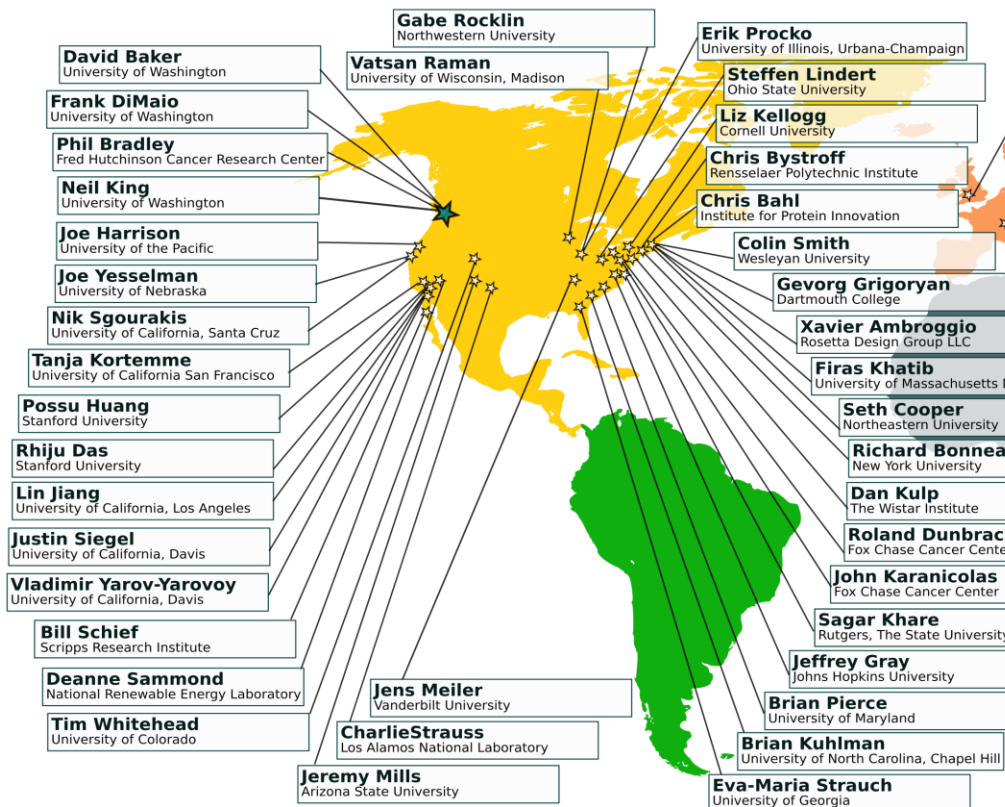
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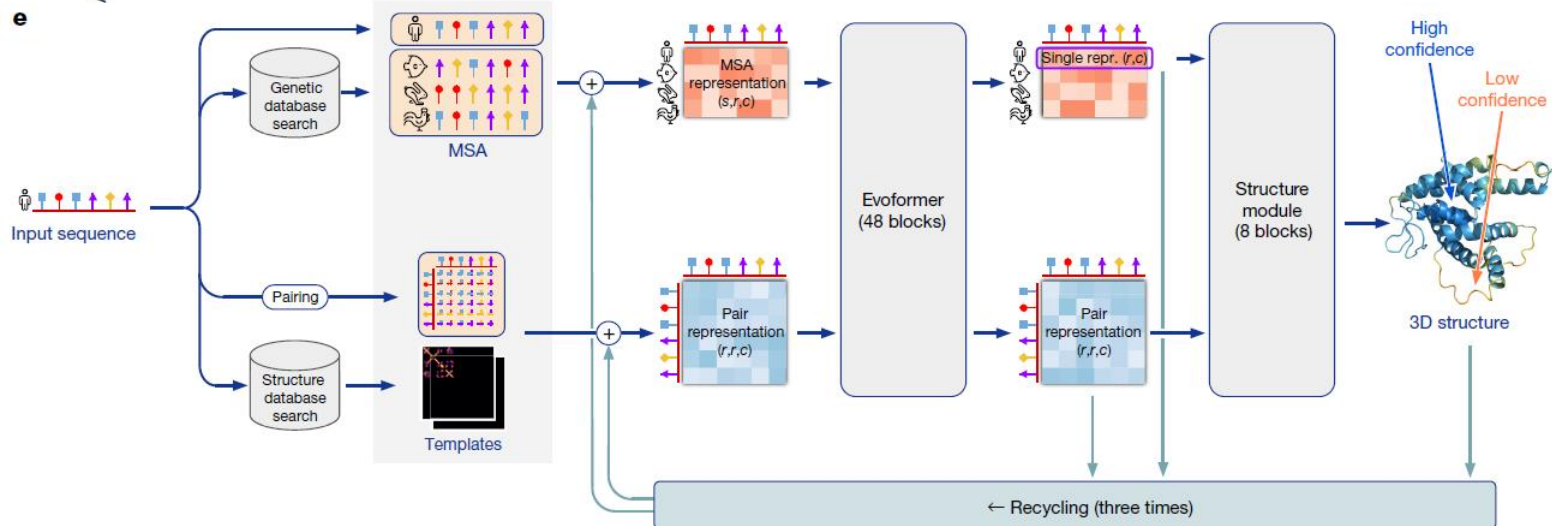
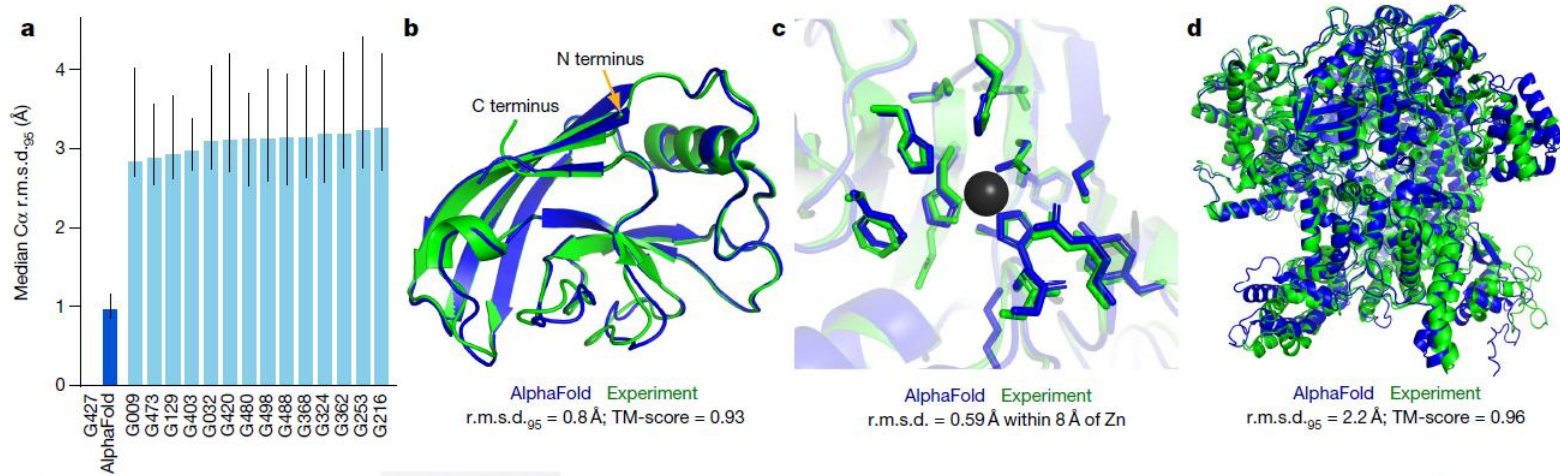


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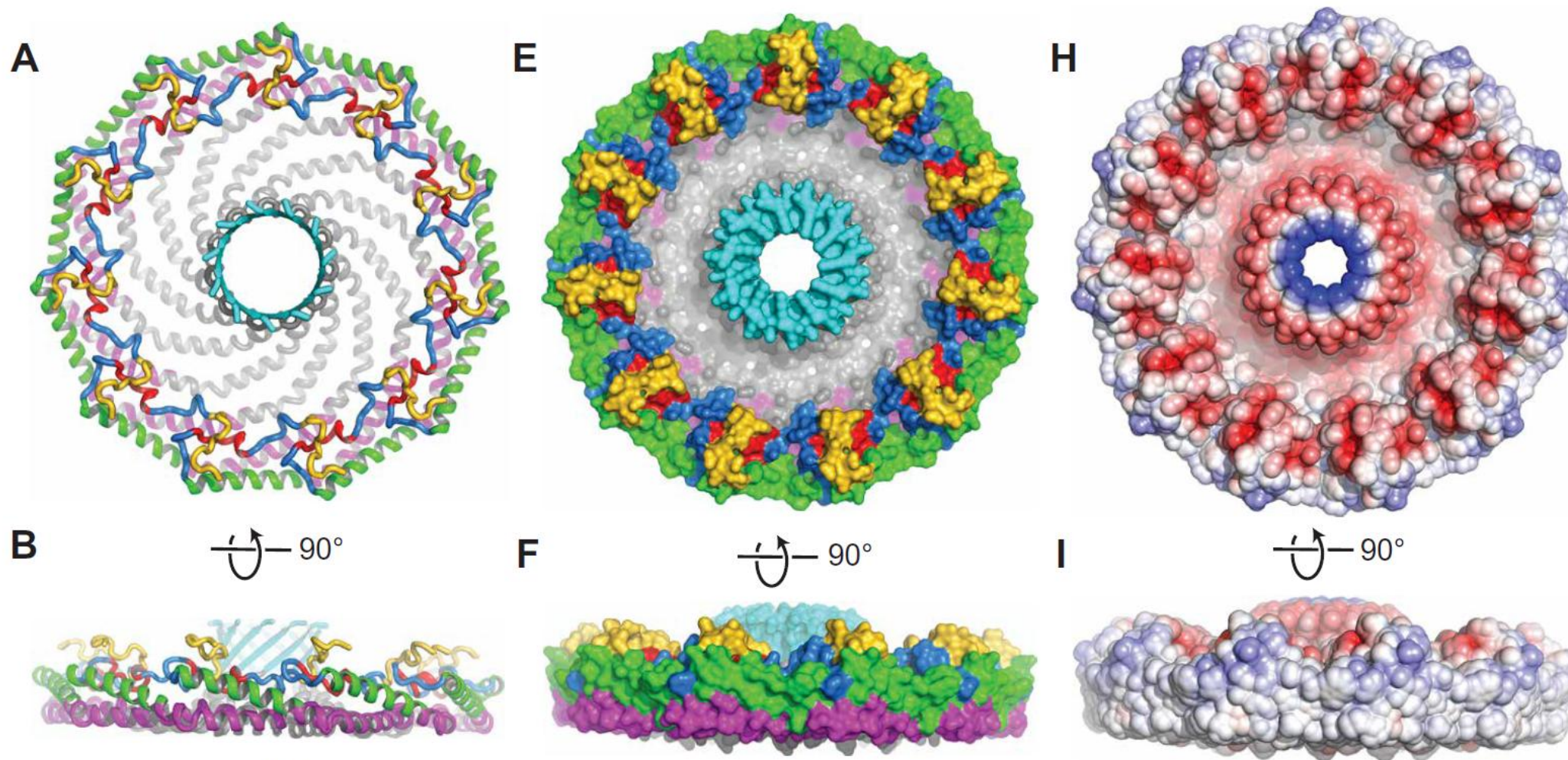


Highly accurate protein structure prediction with AlphaFold2



J. Jumper, et al.; "Highly accurate protein structure prediction with AlphaFold"; *Nature*; **2021**; Vol. 596 (7873): p. 583-589.

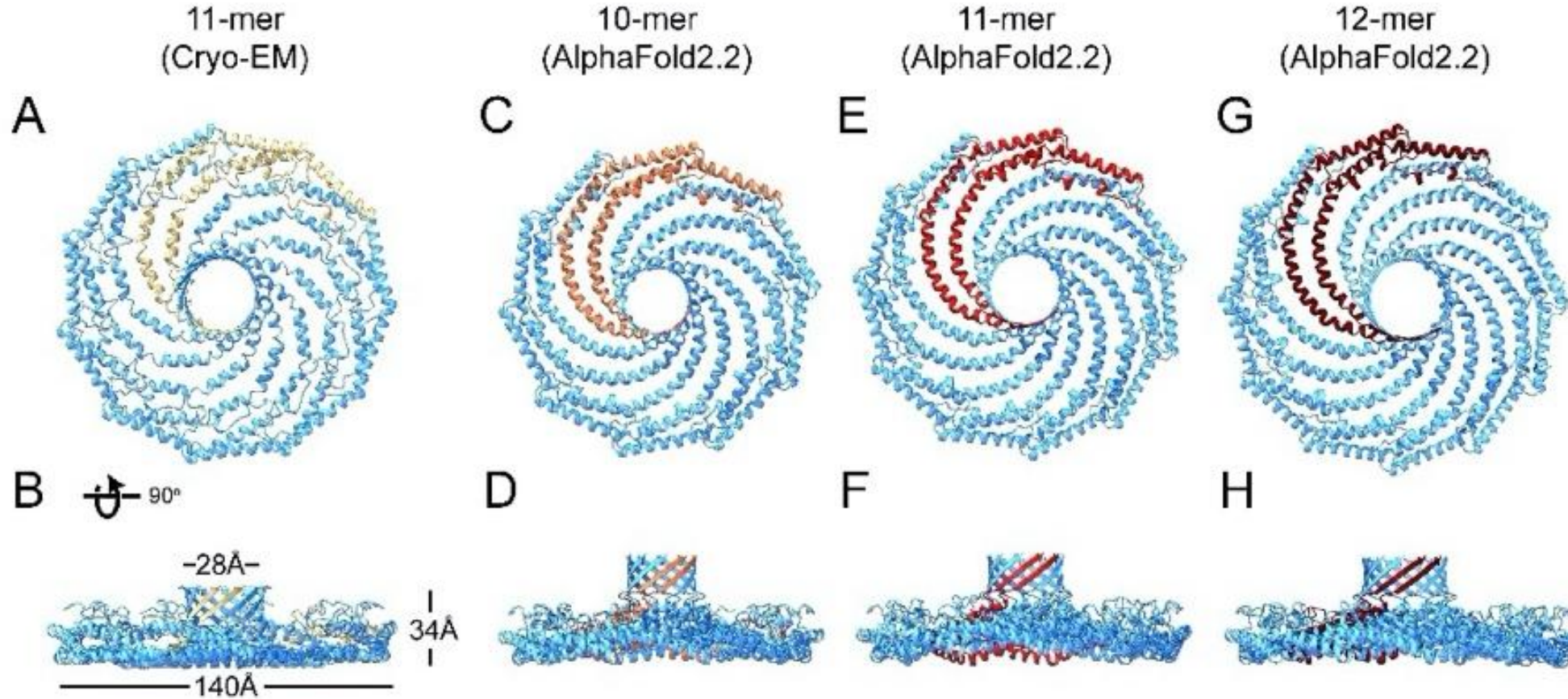
Molecular Architecture of the Human Caveolin-1 Complex



J. C. Porta, B. Han, A. Gulsevin, J. Chung, Y. Peskova, S. Connolly, H. S. Mchaourab, J. Meiler, E. Karakas, A. K. Kenworthy and M. D. Ohl; "Molecular architecture of the human caveolin-1 complex"; *Science Advances*; **2022**; Vol. p.



Molecular Architecture of the Human Caveolin-1 Complex with AlphaFold2

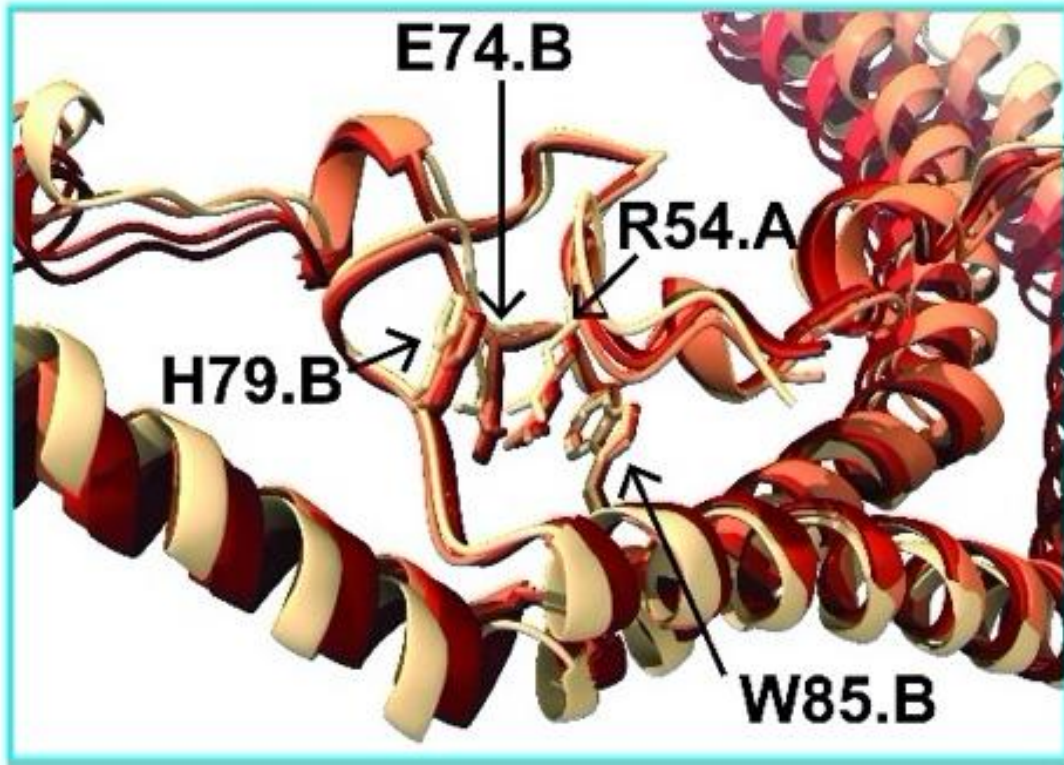


A. Gulsevin, B. Han, J. C. Porta, H. S. McHaourab, J. Meiler and A. K. Kenworthy; "Template-free prediction of a new monotopic membrane protein fold and assembly by AlphaFold2"; *Biophys J*; **2022**; Vol. p.

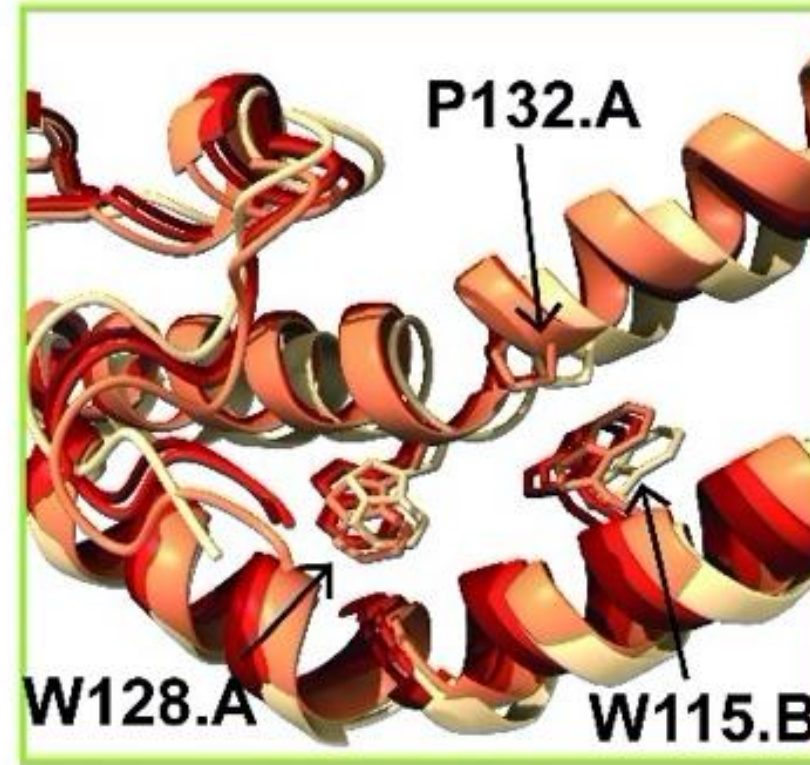
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What does the Nobel Prize mean for 2025+? Integrating Chemical and Structural Biology!

Technology

- Expand the chemical space of AI design
- Learn to design protein dynamics for function
- Control at 0.1Å instead of 1.0Å

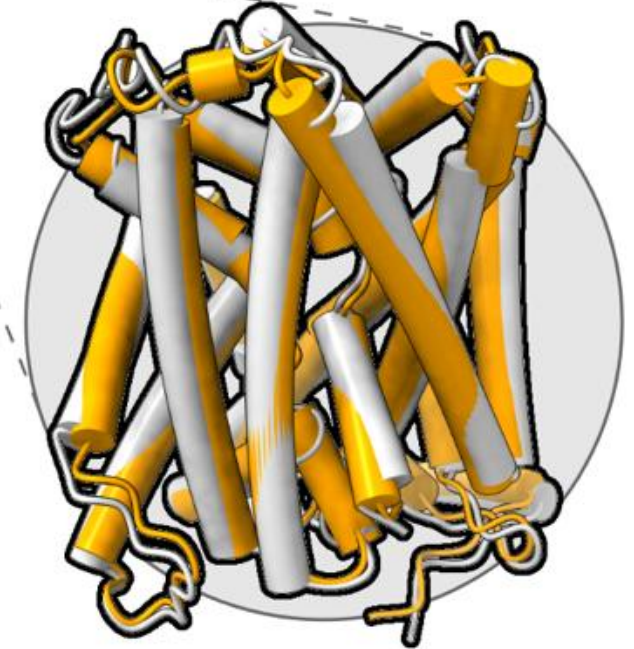
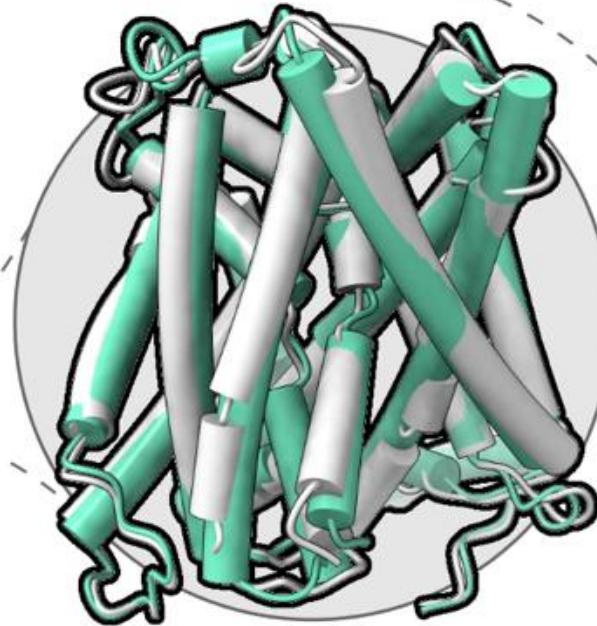
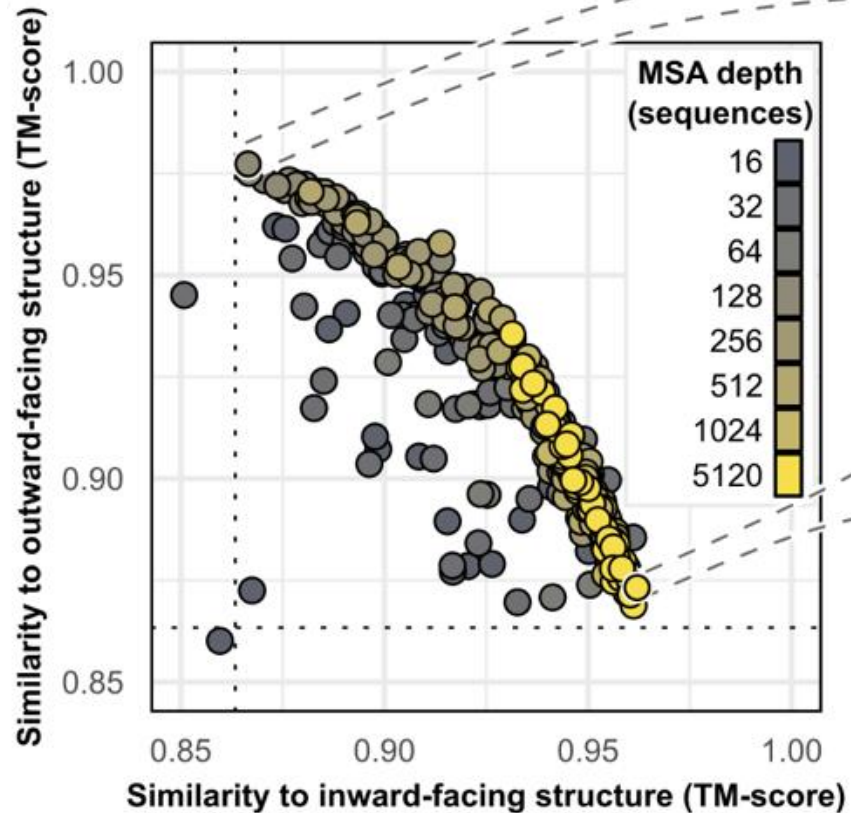


Application

- Medicine & Therapeutics
- Technology and Materials
- Sustainability

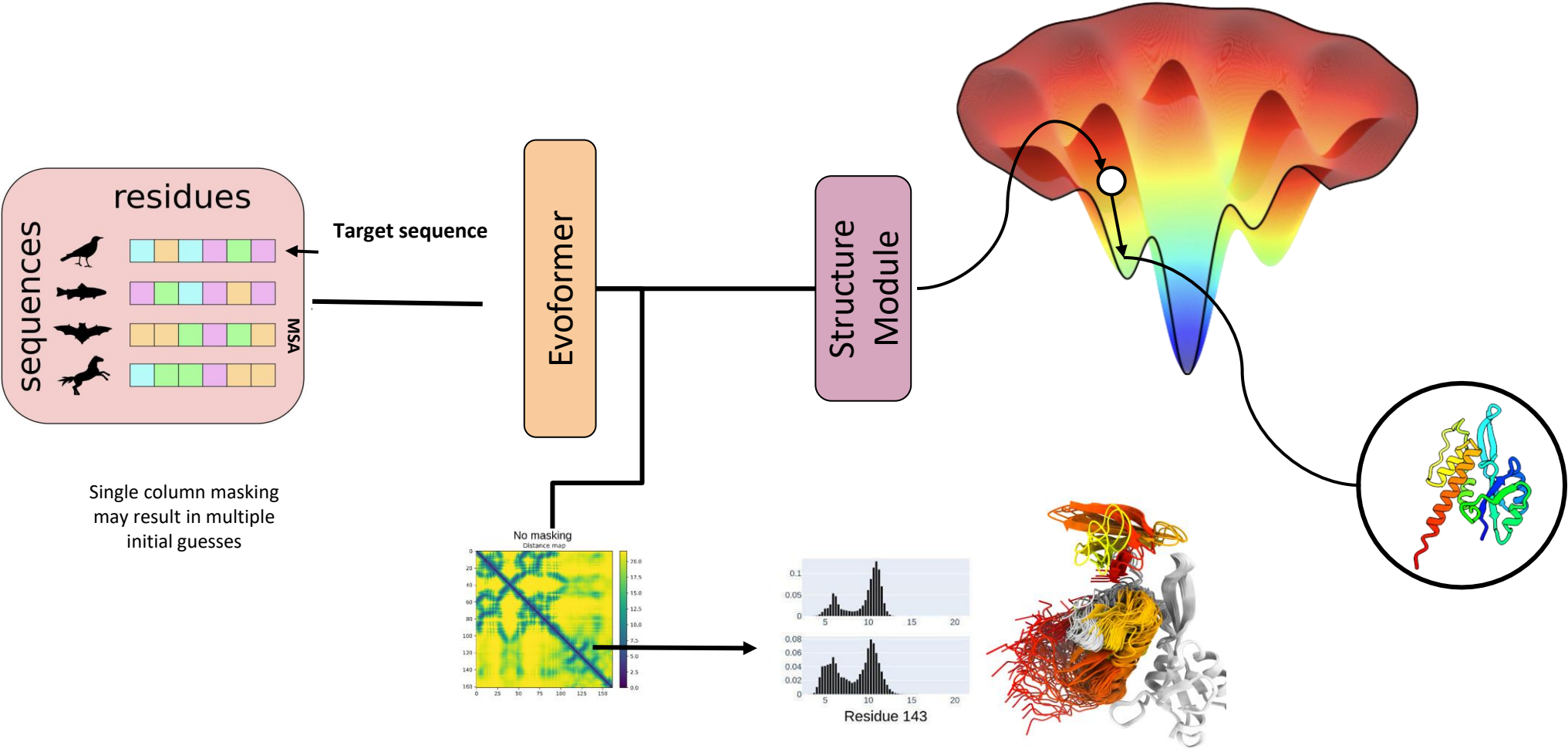


Sampling Alternative Conformational States with AlphaFold2

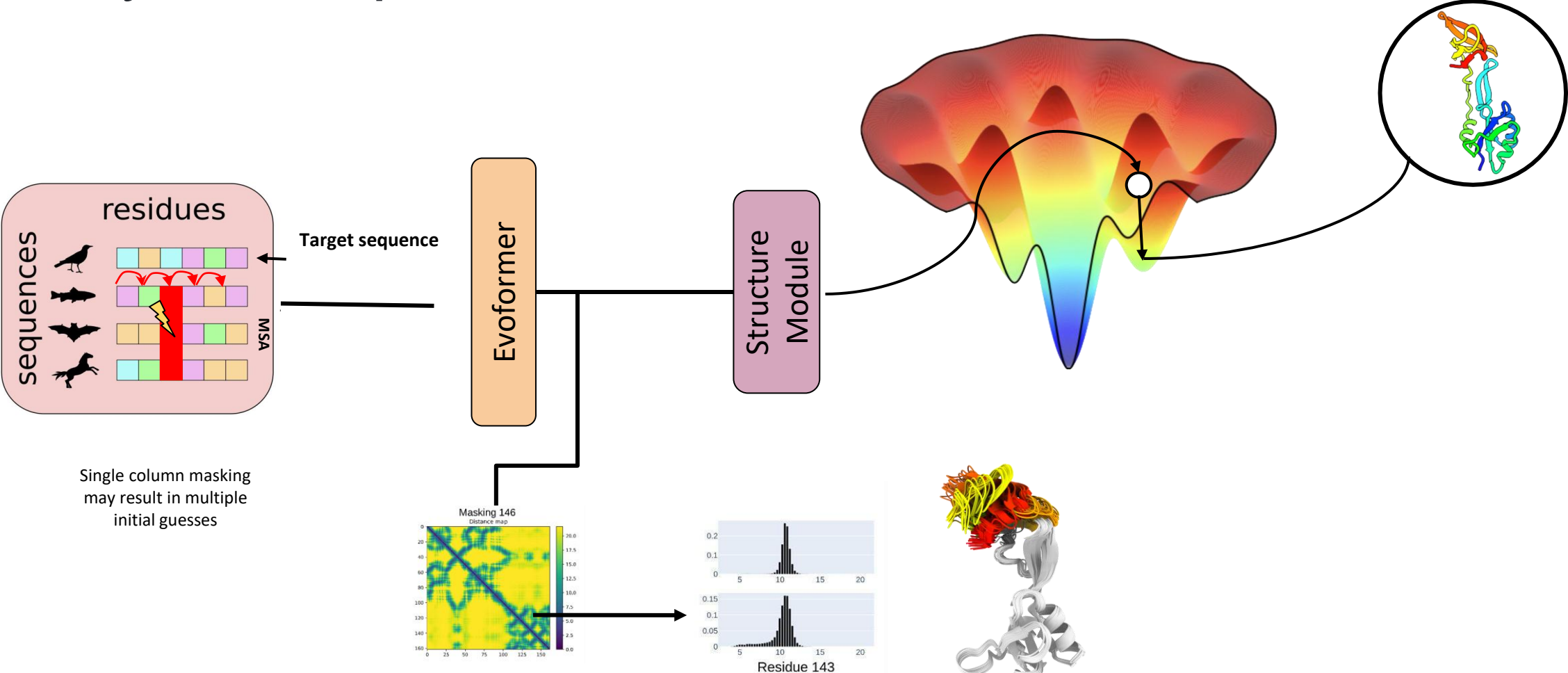


D. Del Alamo, D. Sala, H. S. Mchaourab and J. Meiler; "Sampling alternative conformational states of transporters and receptors with AlphaFold2"; *Elife*; **2022**; Vol. 11 p.

Explaining by removing: Demystifying AF2 ability to sample conformations – rfaH



Explaining by removing: Demystifying AF2 ability to sample conformations – rfaH

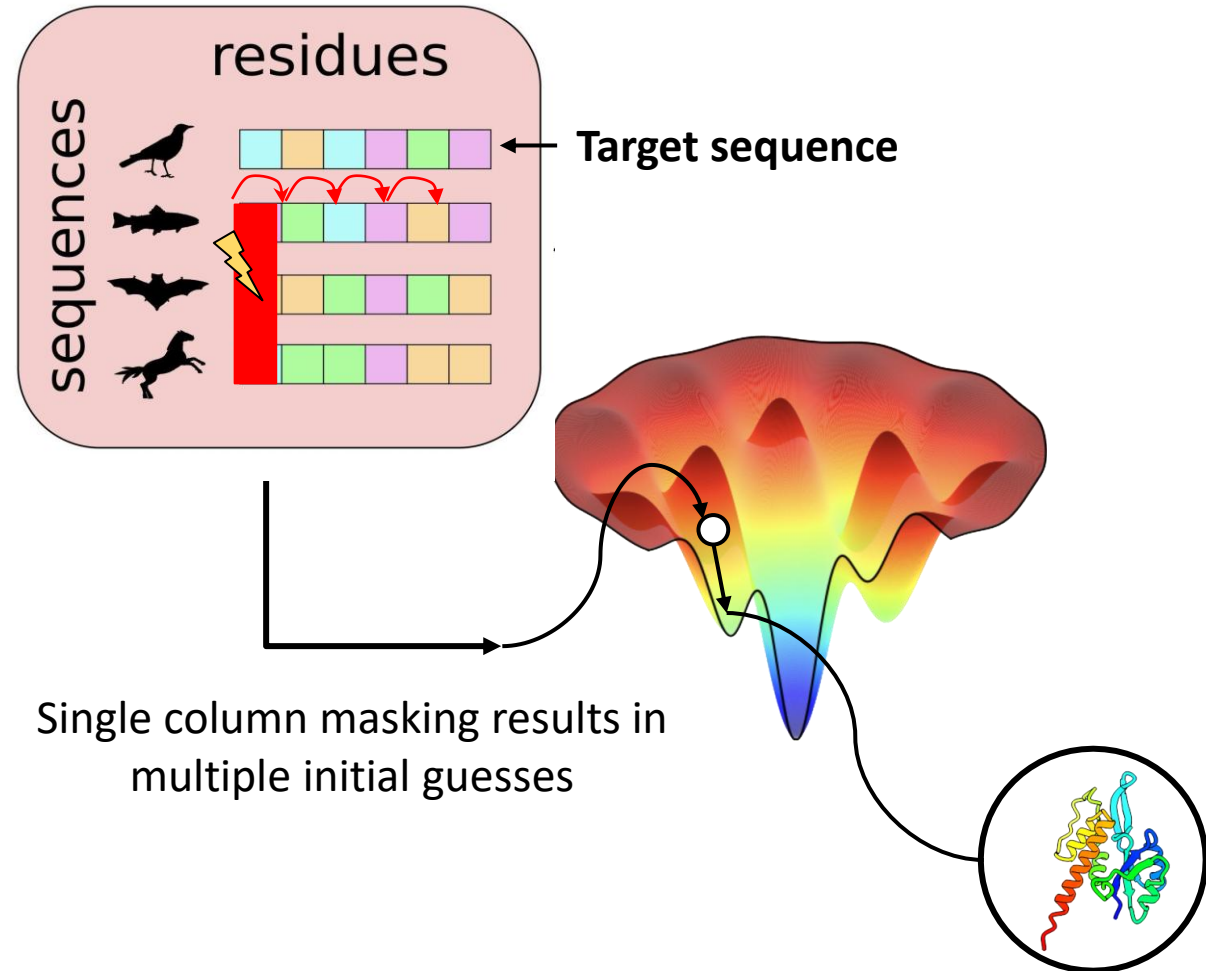
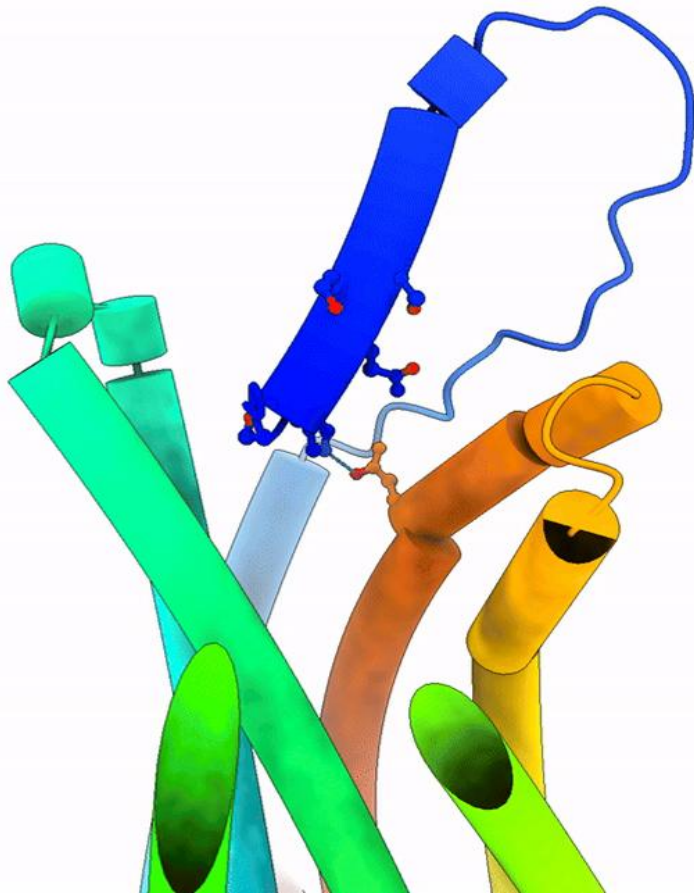


Masking Position 146 rfaH makes AF2 predict only the fold switch state



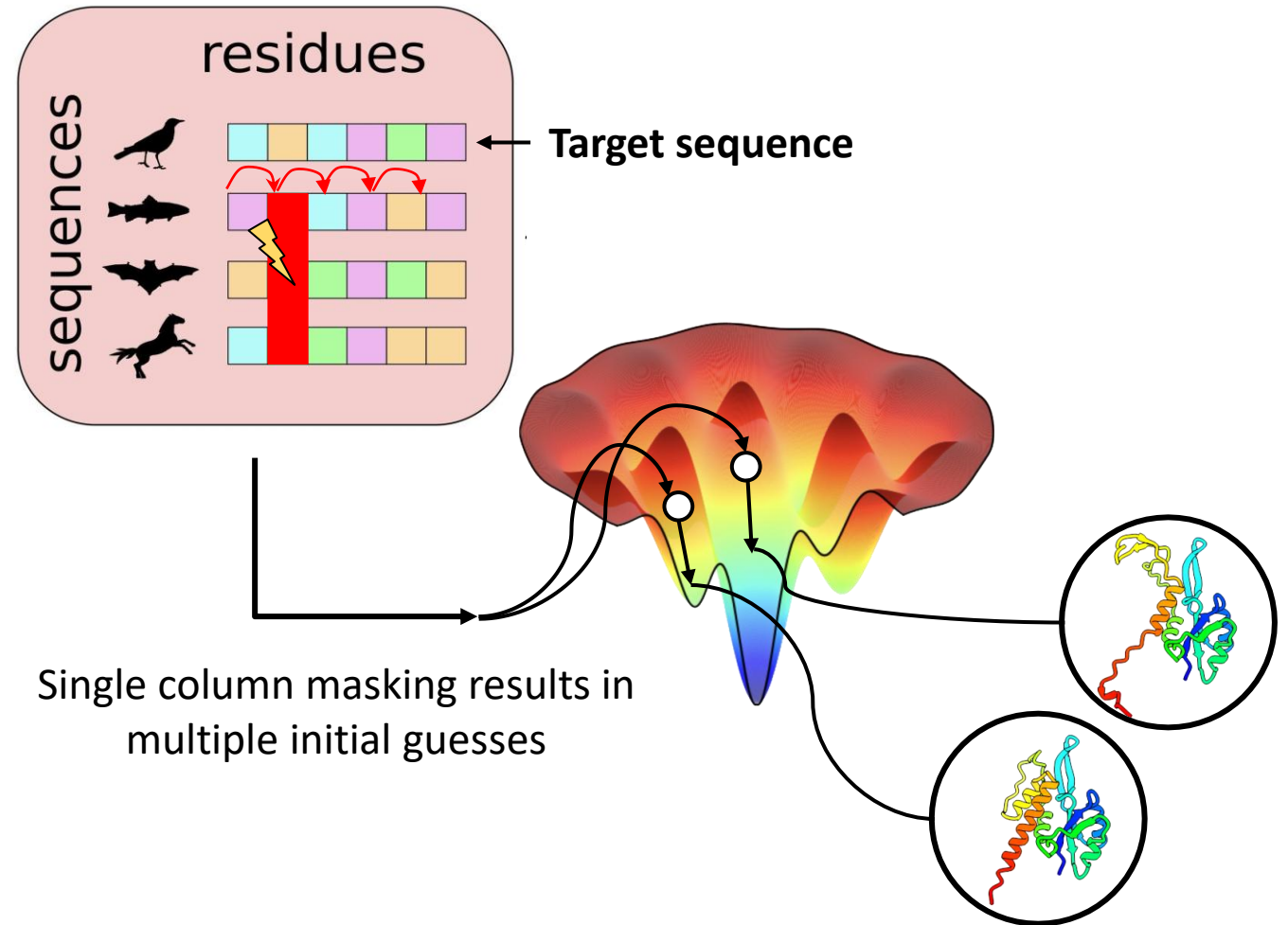
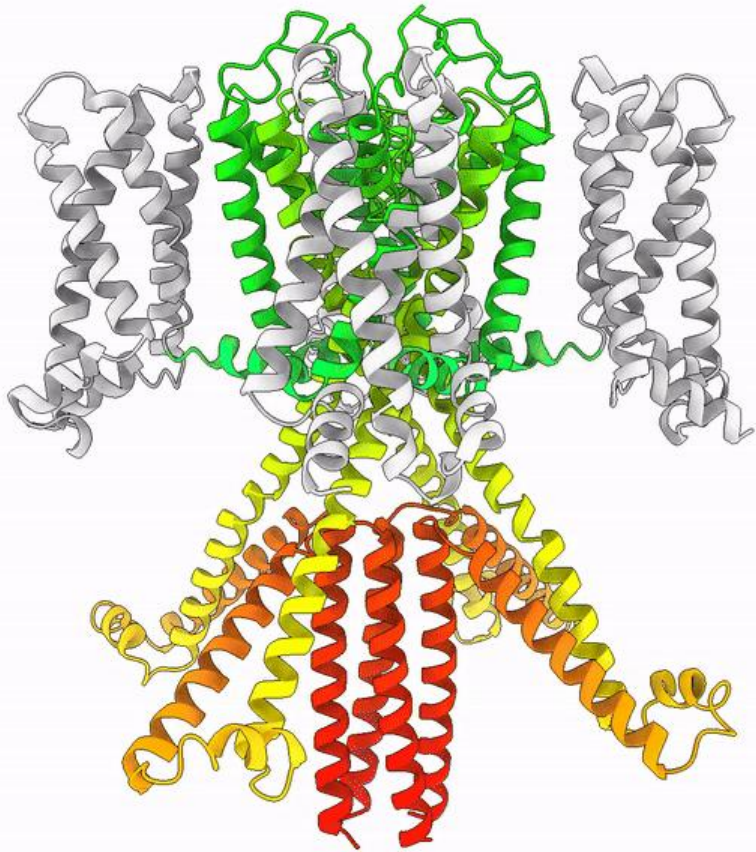
Demystifying AF2 ability to sample conformations

MC4R auto inhibiting peptide



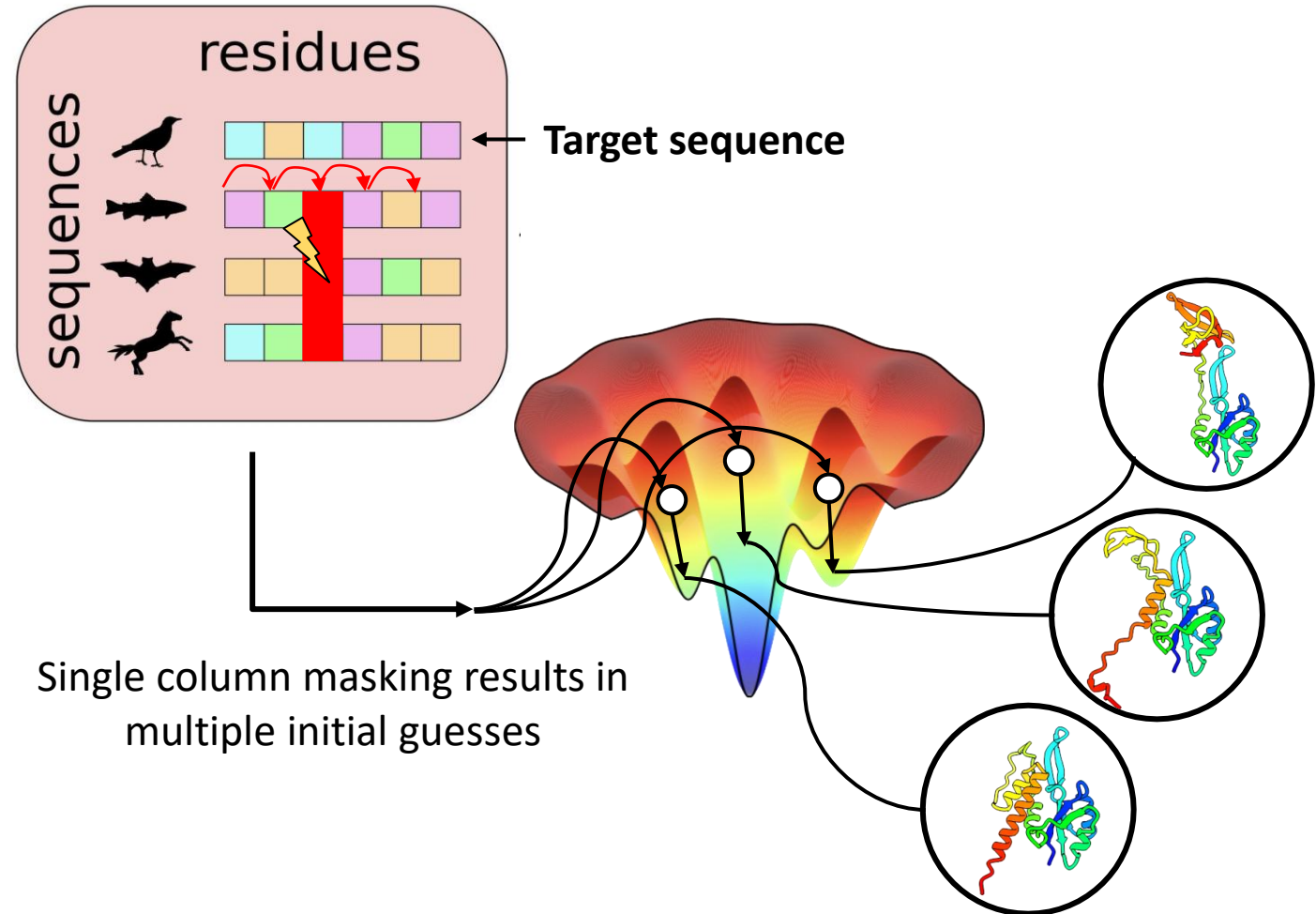
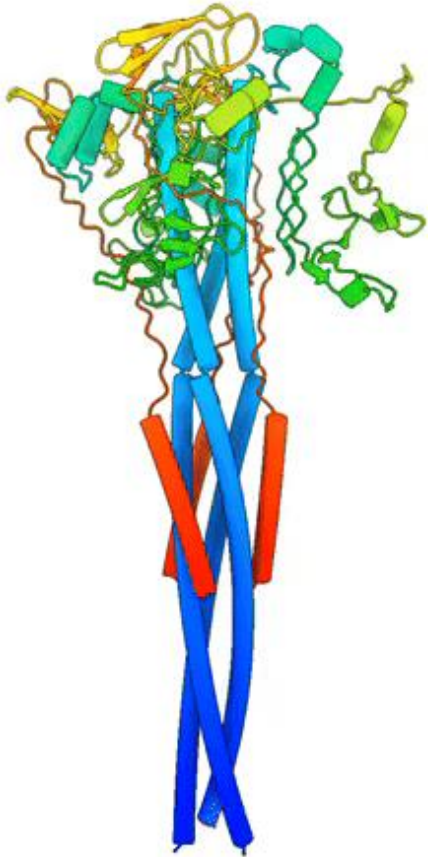
Demystifying AF2 ability to sample conformations

KCNQ1 potassium channel

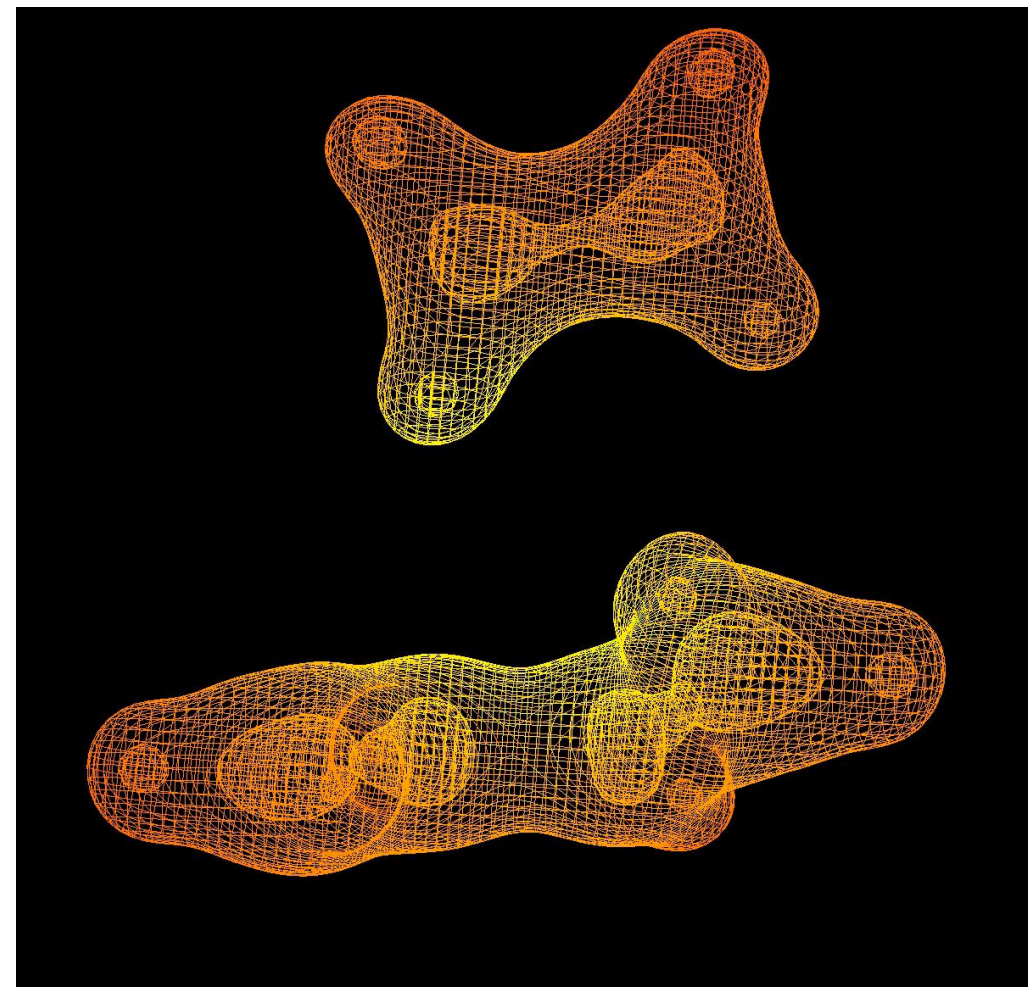
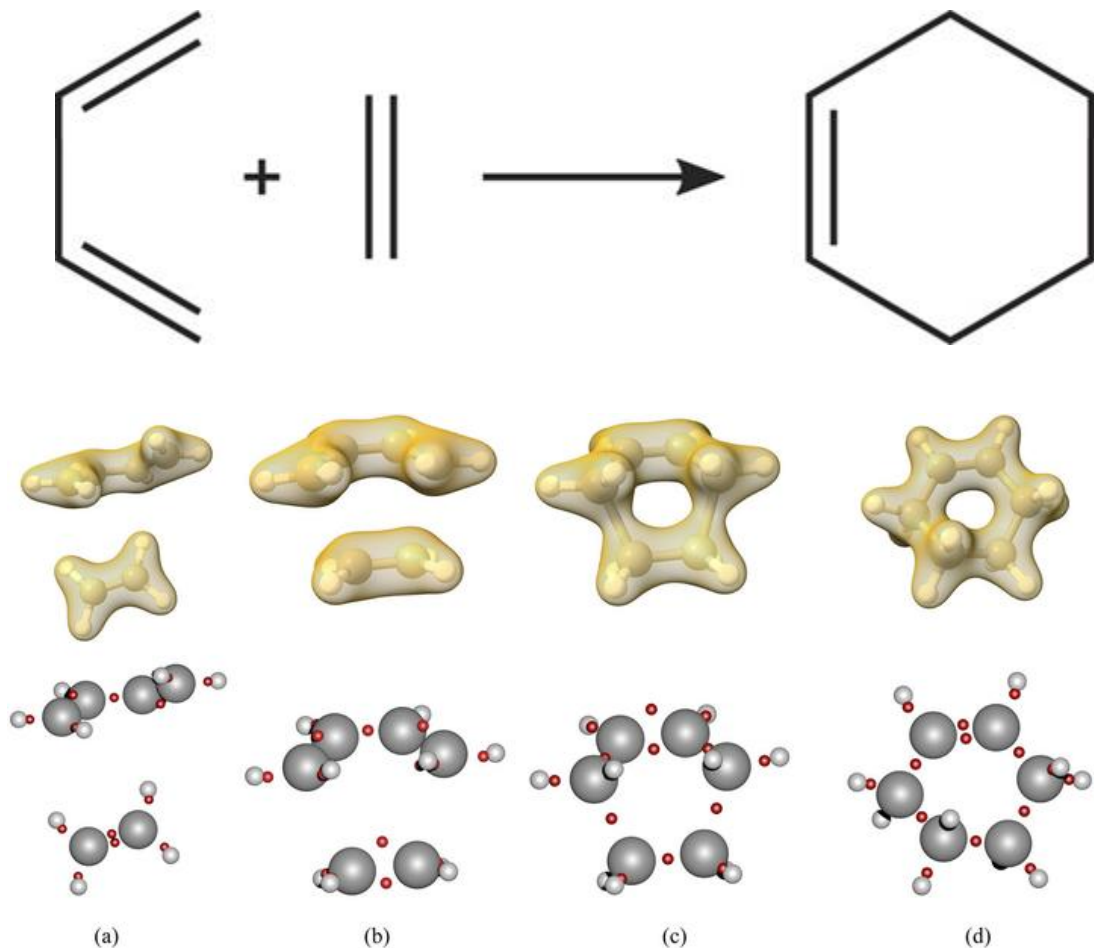


Demystifying AF2 ability to sample conformations

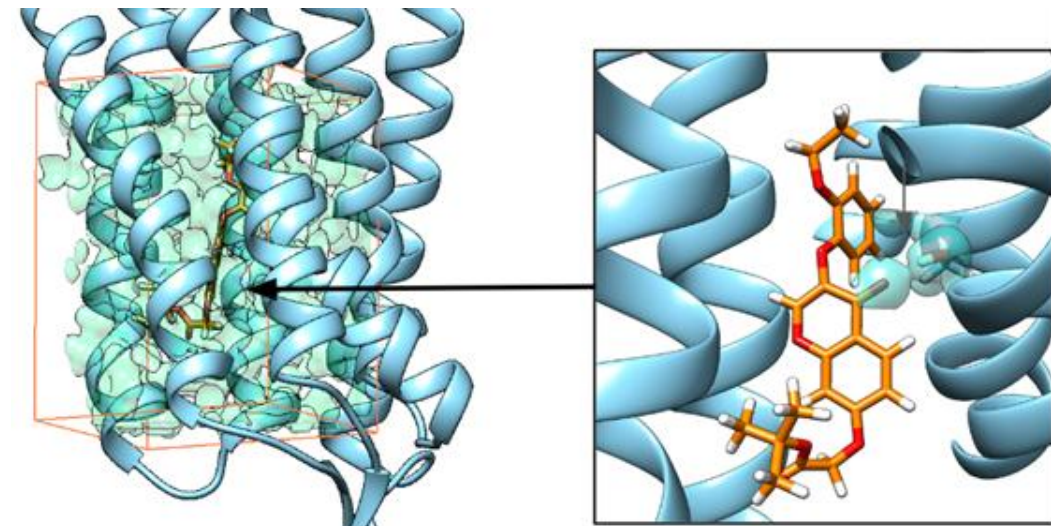
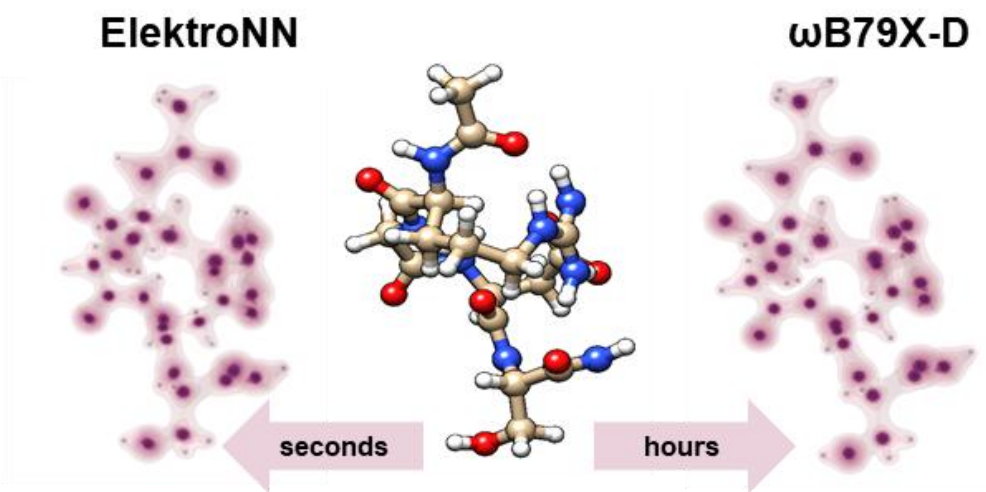
Respiratory Syncytial Virus Glycoprotein



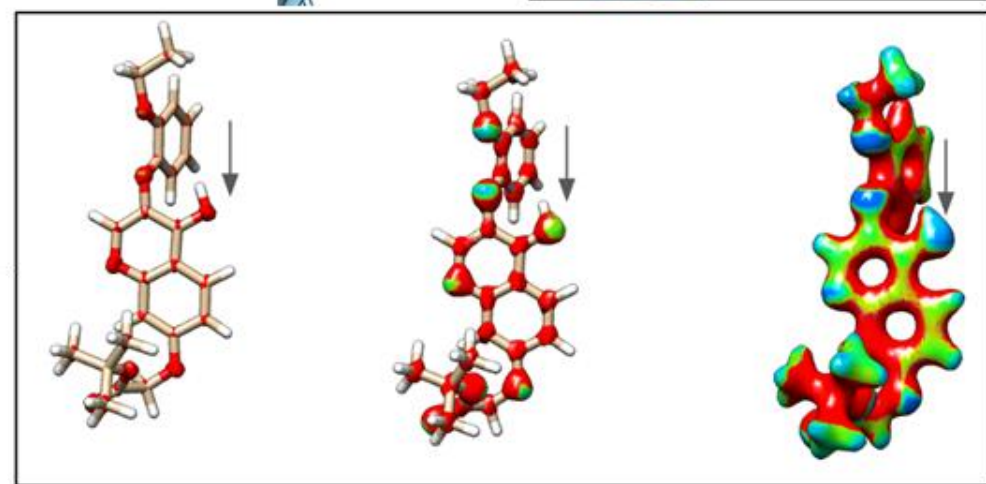
ElektroNN: Predicting Electron Density changes for the Diels-Alder Reaction



ElektroNN: Generalization to non-covalent Interactions in Ligands, Peptides, and Proteins

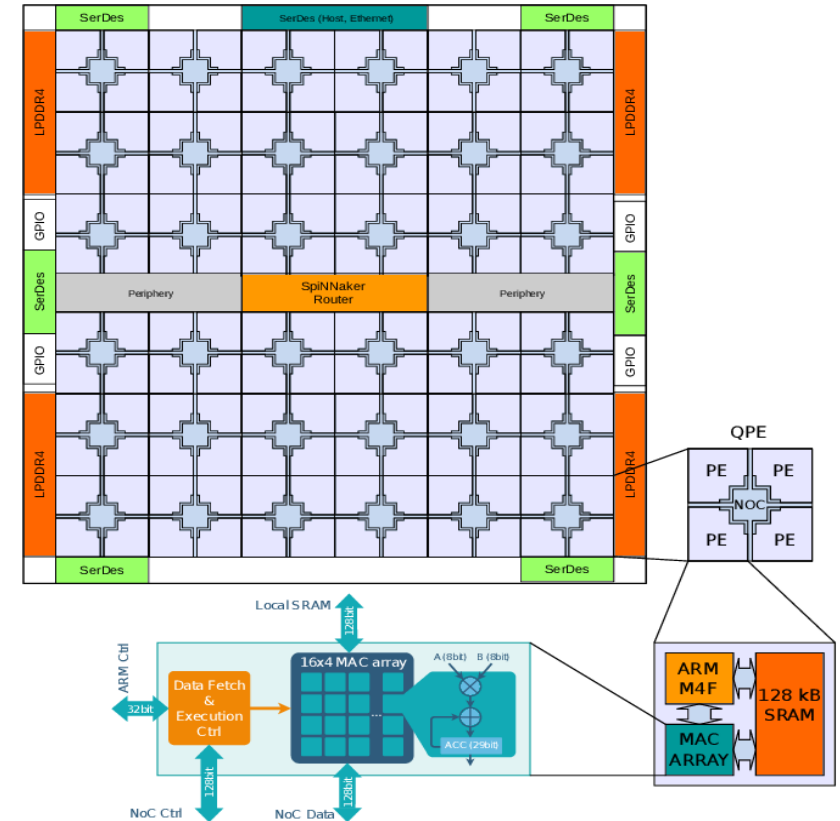
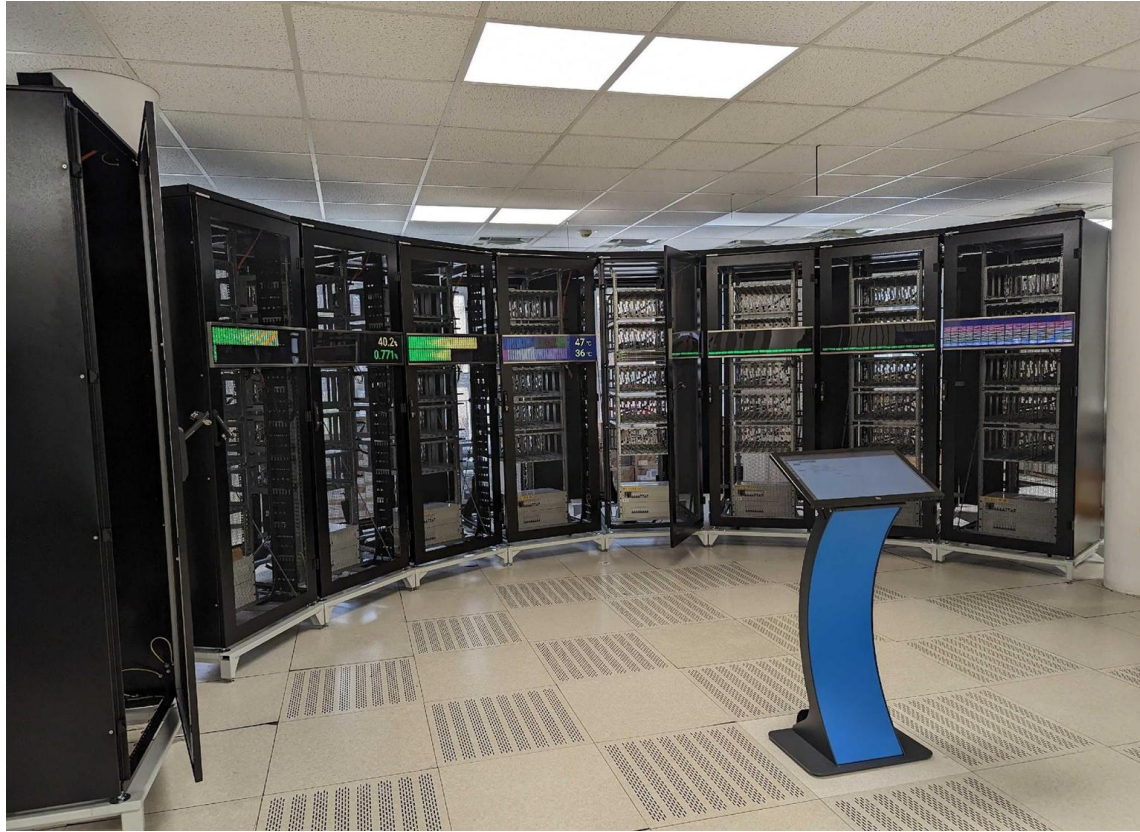


Peptide (5 conformations each)	Charge	Error per Atom (%)
VRN	1	0.30
KGD	0	0.35
DGEA	-2	0.29
APGL	0	0.25
RGD	0	0.30



The SpiNNaker2 Platform: Between GPUs and QC

10^{10} Cores and 10^{14} Connections



Yan, Stewart, Choo, Vogginger, Partzsch, Höppner, Kelber, Eliasmith, Furber, and **Mayr**: “Comparing Loihi with a SpiNNaker 2 prototype on low-latency keyword spotting and adaptive robotic control”; *Neuromorph. Comput. Eng.* **2021** || Höppner, Yan, Dixius, Scholze, Partzsch, Stolba, Kelber, Vogginger, Neumärker, Ellguth, Hartmann, Schiefer, Hocker, Walter, Liu, Garside, Furber, **Mayr**: “The SpiNNaker 2 Processing Element Architecture for Hybrid Digital Neuromorphic Computing”; preprint arXiv:2103.08392 [cs.AR]; **2021**

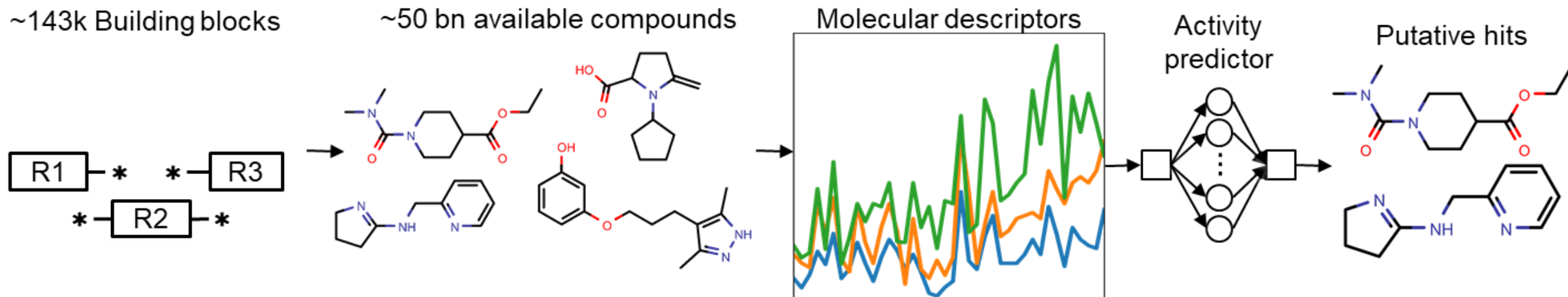


Advantages of SpiNNaker2 for Non-cognitive/non-AI Numerical Problems

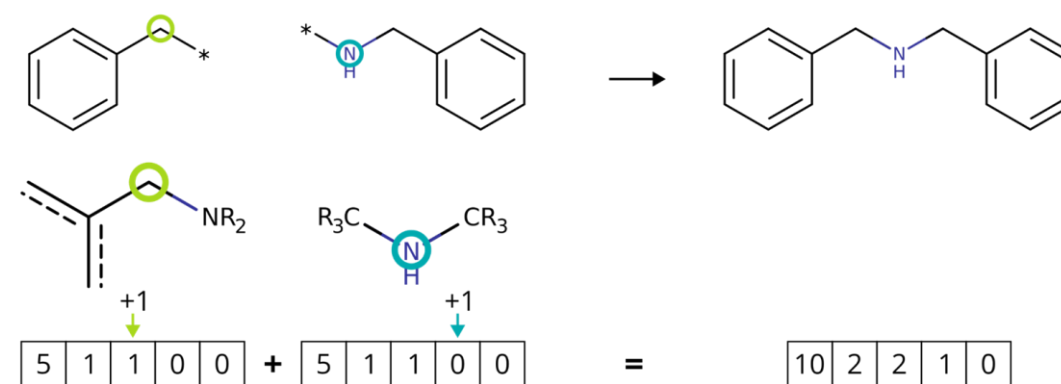
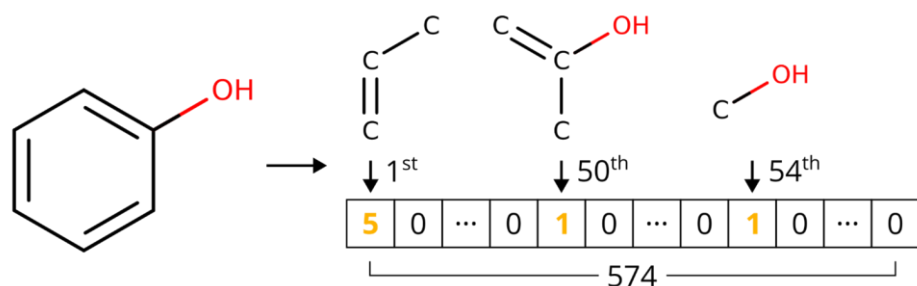
- Extreme parallelism of simple operations (think neurons...)
- (Search for) Sparse solutions in high-dimensional numerical rooms
- Stochastic computation/stochastic state representations
- Solving systems of locally coupled differential equations in a mesh/network topology (e.g. Neuron models, but also FEM and similar)

	HPC	SpiNNaker2	Quantum Computing
Parallelism	10^5 cores	10^{14} synaptic updates/msec	$>10^{25}$ quantum entanglements
Stochastic Computation	Only in software, 10^{10} stochastic decisions/sec	Hardware accelerators, 10^{17} stochastic decisions/sec	Inherent in Qubits, $>10^{30}$ stochastic decisions/sec
Sparsity in high-dimensional spaces	Not supported	Fully supported	Fully supported
FEM-type tessellations	10^5 elements, boundary condition updates μ s to ms	10^7 elements in torus, boundary condition updates $<10\mu$ s	Potentially very fast convergence, but tessellation limited to #Qubits: 10^2 - 10^3

Ultra-Large Library Screening on the SpiNNaker2 Platform



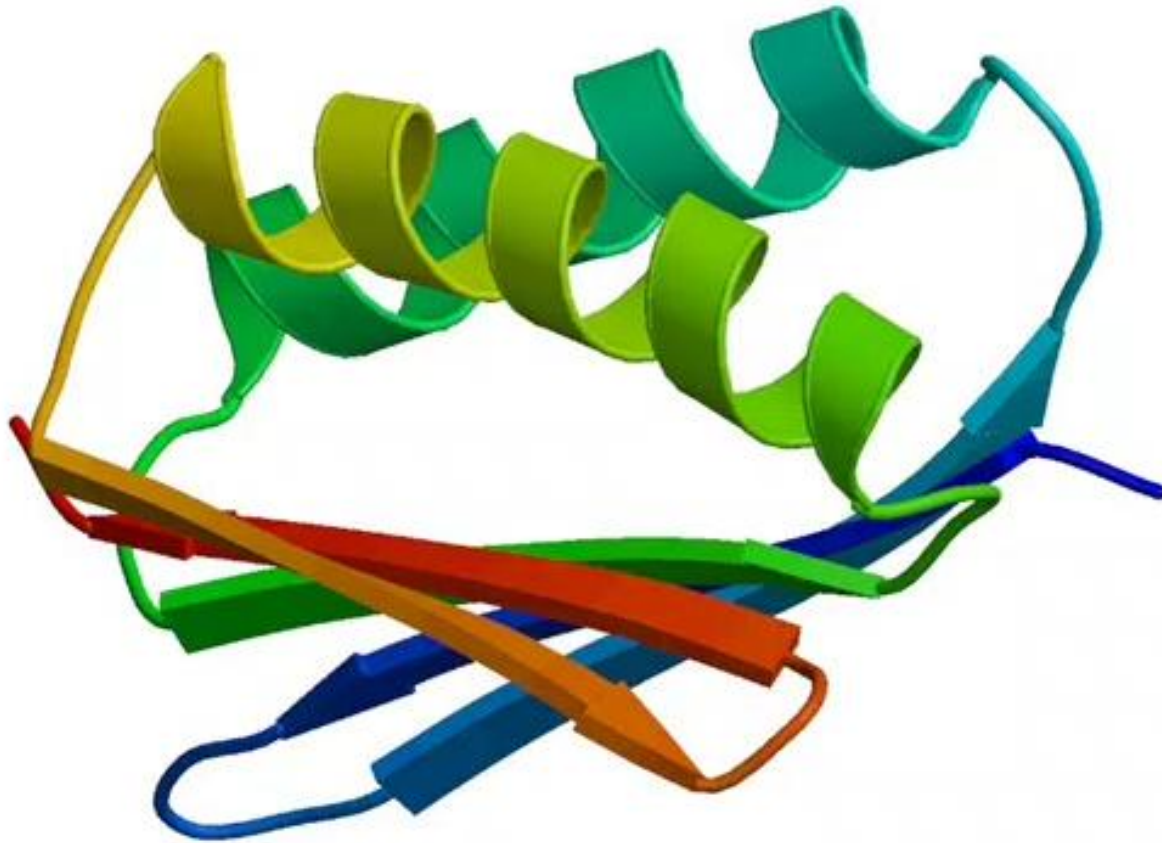
BCL::Mol2D descriptors from each atom's immediate neighborhood



Vu, O., Mendenhall, J., Altarawy, D., & Meiler, J. (2019). BCL::Mol2D-a robust atom environment descriptor for QSAR modeling and lead optimization. *J Comput Aided Mol Des*, 33(5), 477-486. <https://doi.org/10.1007/s10822-019-00199-8>



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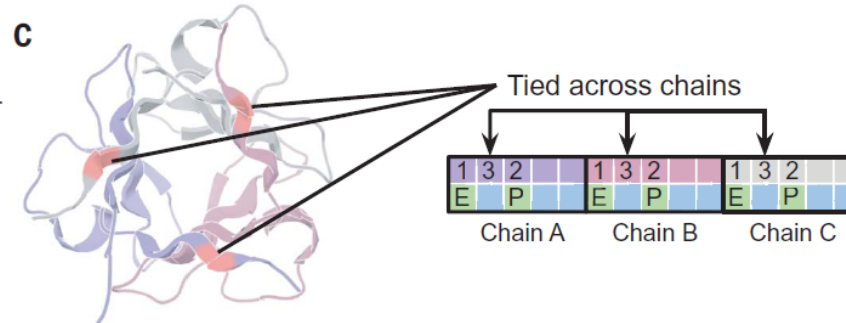
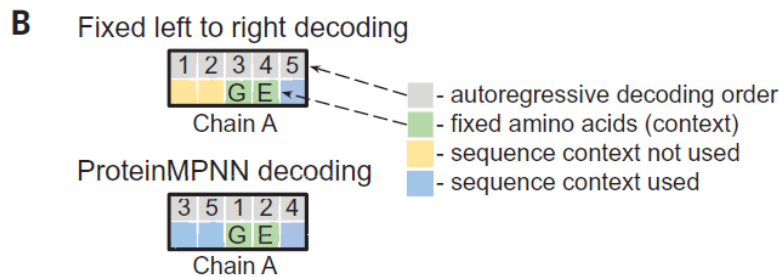
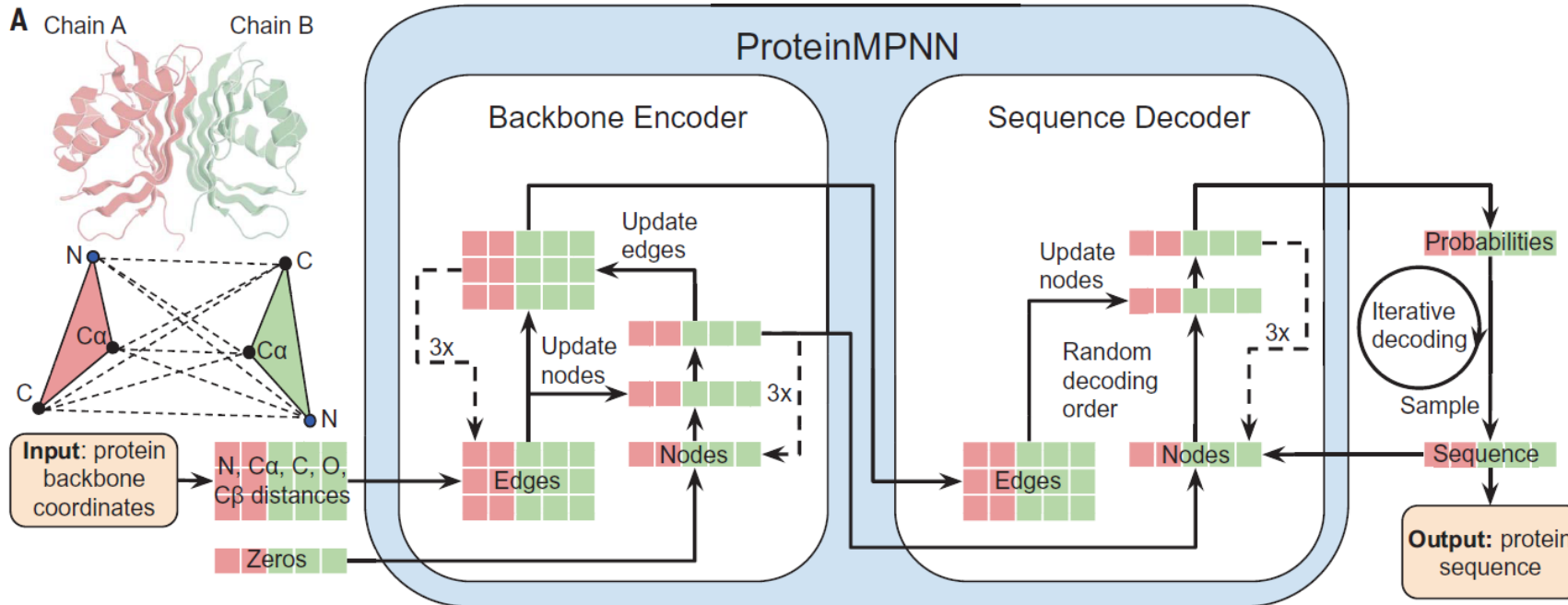
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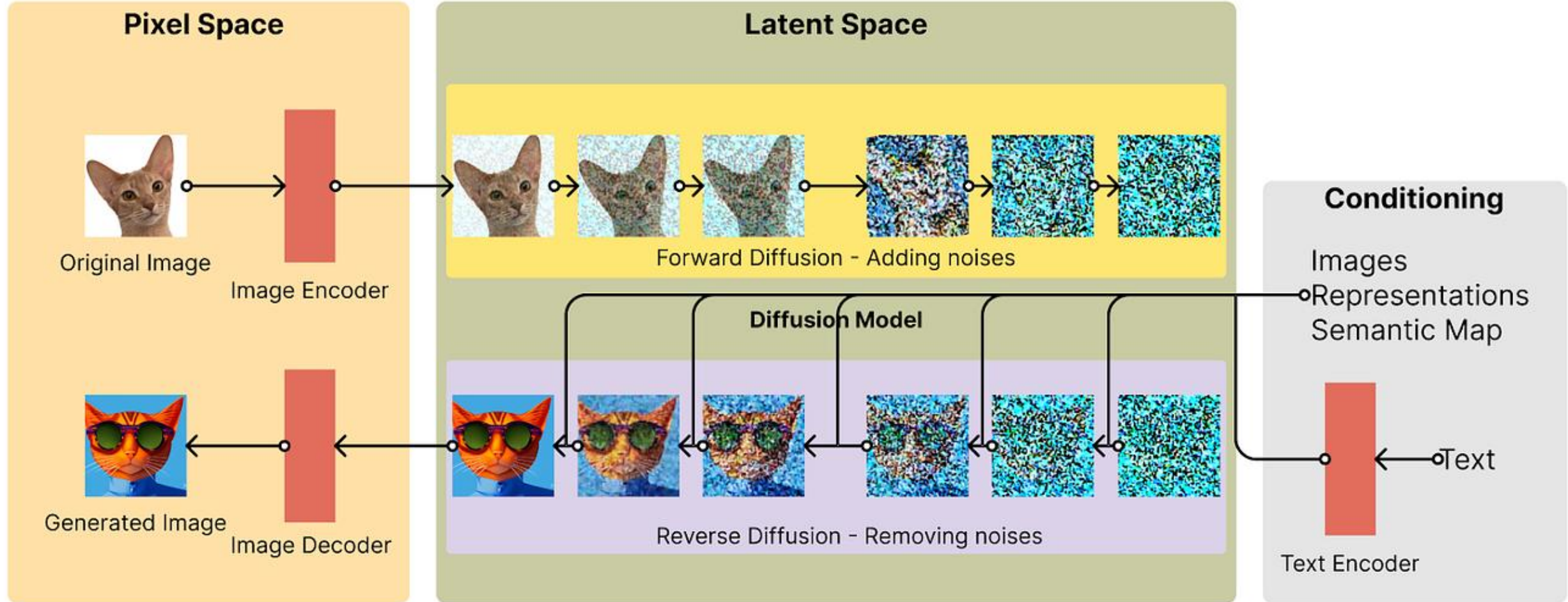


Robust deep learning-based protein sequence design using ProteinMPNN

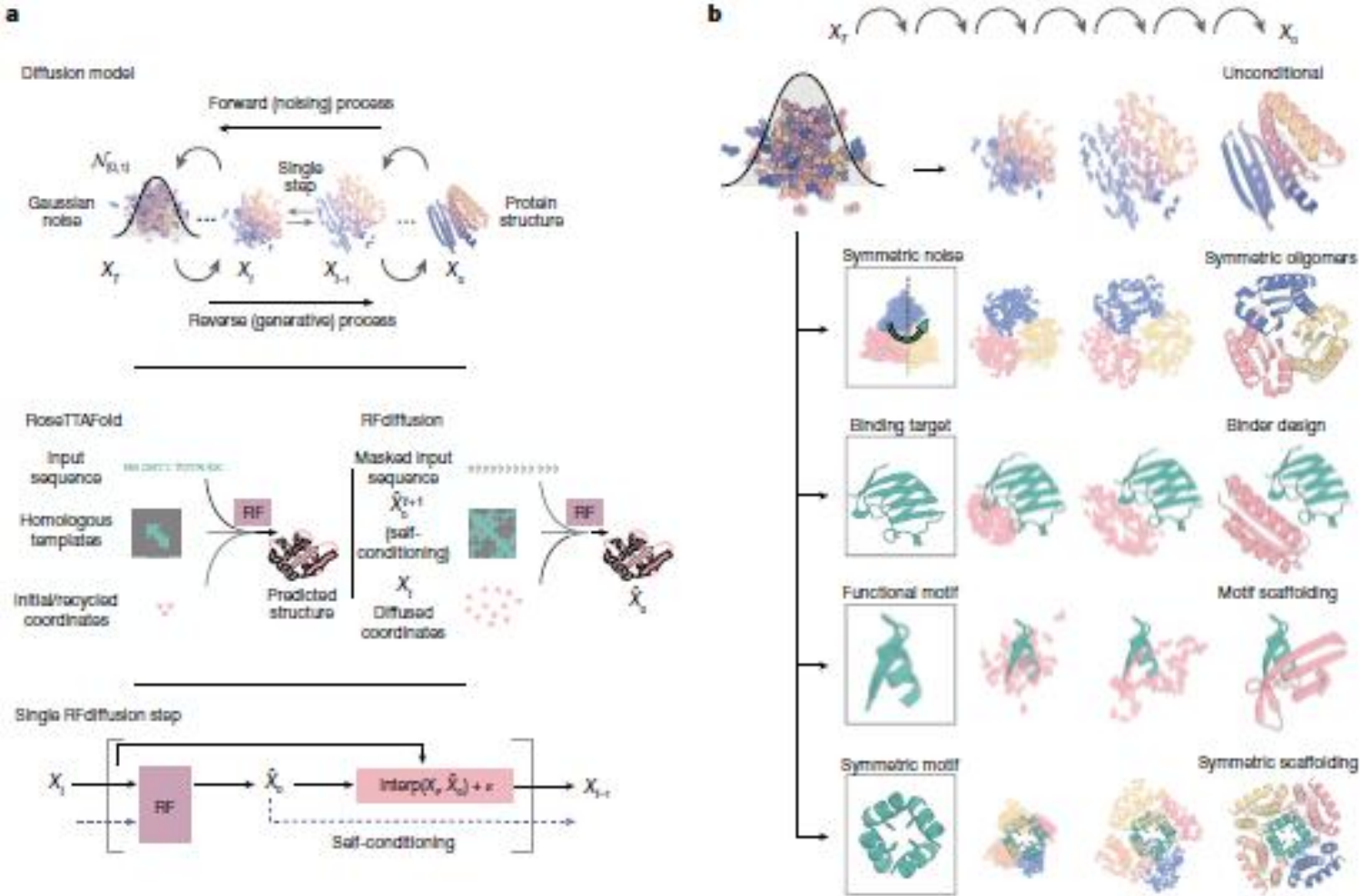


J. Dauparas, I. Anishchenko, N. Bennett, H. Bai, R. J. Ragotte, L. F. Milles, B. I. M. Wicky, A. Courbet, R. J. de Haas, N. Bethel, P. J. Y. Leung, T. F. Huddy, S. Pellock, D. Tischer, F. Chan, B. Koepnick, H. Nguyen, A. Kang, B. Sankaran, A. K. Bera, N. P. King and D. Baker; "Robust deep learning-based protein sequence design using ProteinMPNN"; *Science*; **2022**; Vol. **378** (6615): p. 49-56.

How Stable Diffusion works in a Nutshell



De novo design of protein structure and function with RFDiffusion



Watson, J. L., Juergens, D., Bennett, N. R., Trippe, B. L., Yim, J., Eisenach, H. E., Ahern, W., Borst, A. J., Ragotte, R. J., Milles, L. F., Wicky, B. I. M., Hanikel, N., Pellock, S. J., Courbet, A., Sheffler, W., Wang, J., Venkatesh, P., Sappington, I., Torres, S. V., . . . Baker, D. (2023). De novo design of protein structure and function with RFDiffusion. *Nature*, 620(7976), 1089-1100.

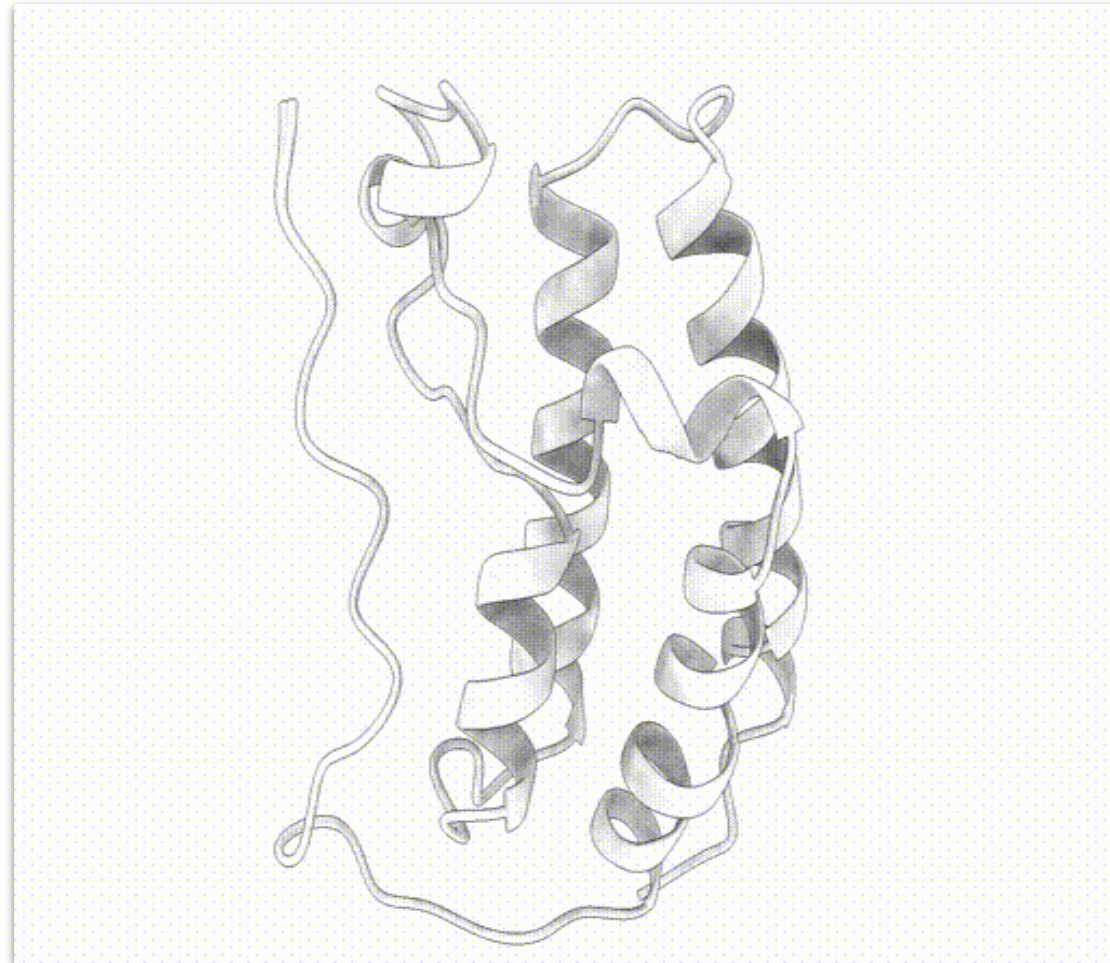
Modified Version of RFDiffusion for (Cyclic) Peptides

Normal positional embedding

0	1	2	3	4	5	6	7
-1	0	1	2	3	4	5	6
-2	-1	0	1	2	3	4	5
-3	-2	-1	0	1	2	3	4
-4	-3	-2	-1	0	1	2	3
-5	-4	-3	-2	-1	0	1	2
-6	-5	-4	-3	-2	-1	0	1
-7	-6	-5	-4	-3	-2	-1	0

Cyclic positional embedding

0	1	2	3	-4	-3	-2	-1
-1	0	1	2	3	-4	-3	-2
-2	-1	0	1	2	3	-4	-3
-3	-2	-1	0	1	2	3	-4
-4	-3	-2	-1	0	1	2	3
3	-4	-3	-2	-1	0	1	2
2	3	-4	-3	-2	-1	0	1
1	2	3	-4	-3	-2	-1	0



What does the Nobel Prize mean for 2025+? Integrating Chemical and Structural Biology!

Technology

- Expand the chemical space of AI design
- Learn to design protein dynamics for function
- Control at 0.1Å instead of 1.0Å



Application

- Medicine & Therapeutics
- Technology and Materials
- Sustainability



Acknowledgements

Current Members:

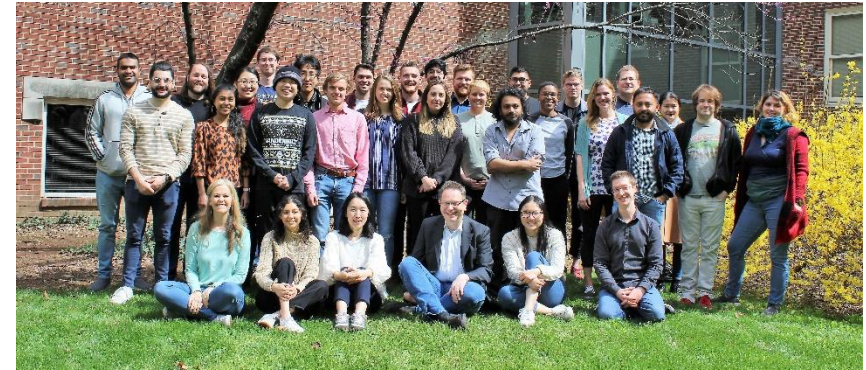
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Alexander Fülll
Alexander Zlobin
Ali Kanso
Ana Chang-Gonzalez
Andrea Pokrajcic
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Chris Moth
Christina Mercado
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Sophie Maclure
Te-Yu Kao
Thomas Scott
Tom Schlegel
Toni Oesterreich
Onur Elaldi
Vivian Ehrlich

Selected Alumni:

Abdullah Al Mamun
Alican Gulsevin
Annalen Bleckmann
Alex Sevy
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Benjamin Brown
Brett Kroncke
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Brian Shoichet, Hassane
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George, Lars Plate, Jonathan
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Sicking, Clara Schoeder, ...

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DFG CRC 1052, DFG TRR 386, DFG SPP
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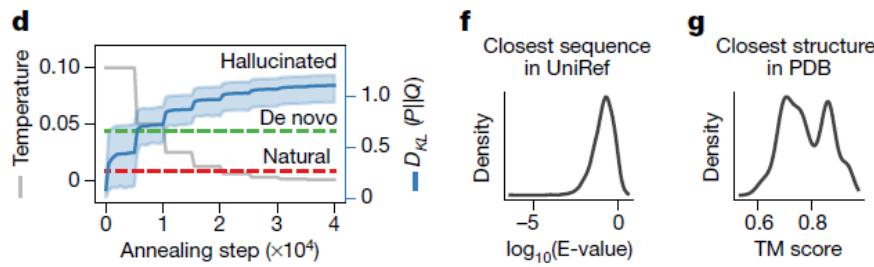
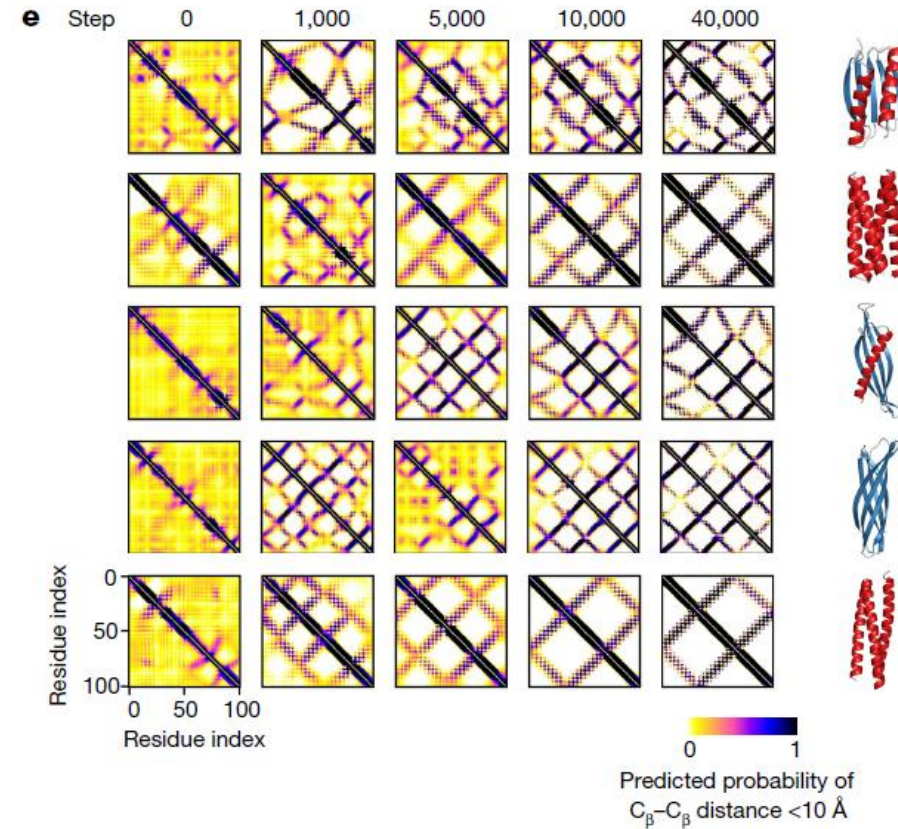
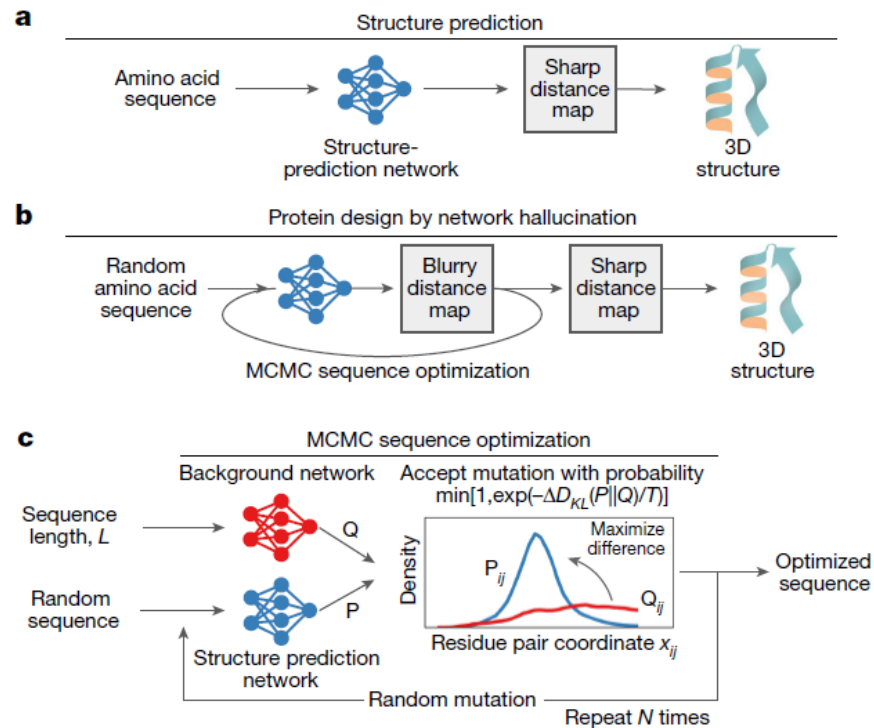


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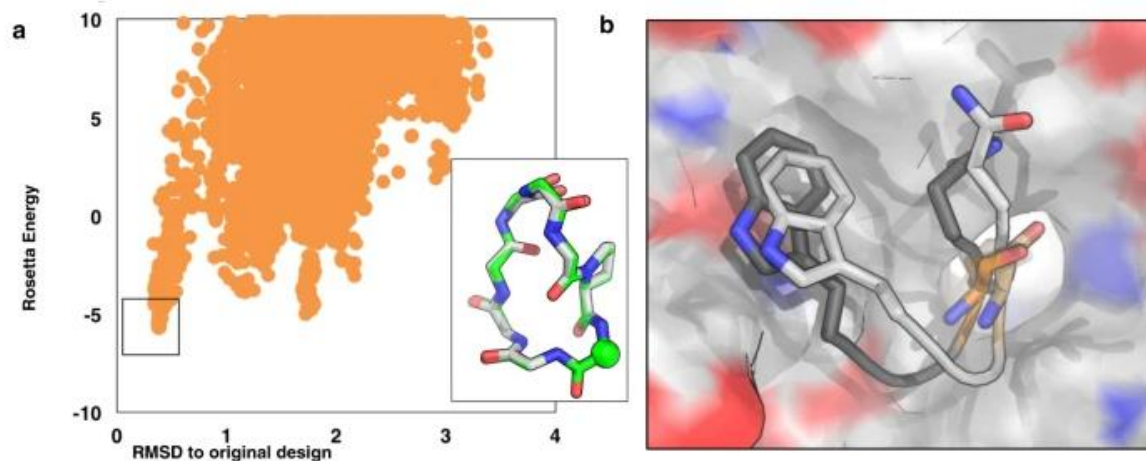
De novo Protein Design by Deep Network Hallucination



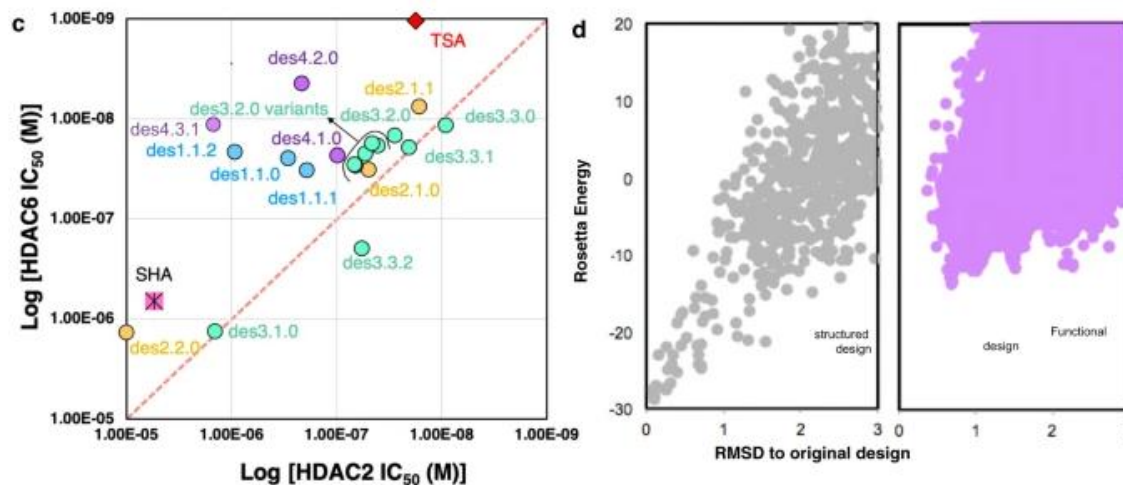
I. Anishchenko, S. J. Pellock, T. M. Chidyausiku, T. A. Ramelot, S. Ovchinnikov, J. Hao, K. Bafna, C. Norn, A. Kang, A. K. Bera, F. DiMaio, L. Carter, C. M. Chow, G. T. Montelione and D. Baker; "De novo protein design by deep network hallucination"; *Nature*; **2021**; Vol. 600 (7889): p. 547-552.



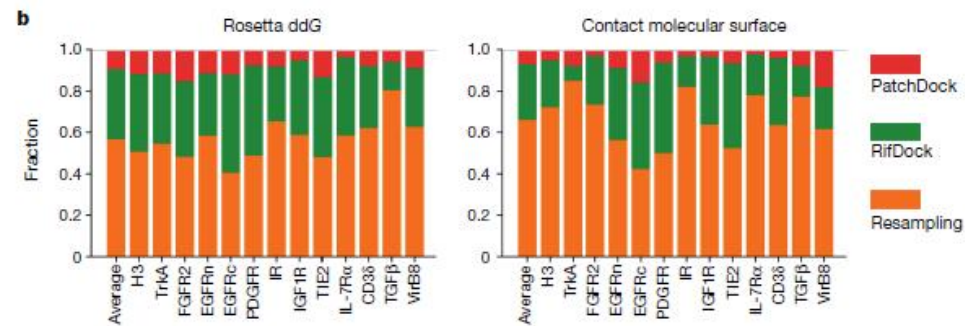
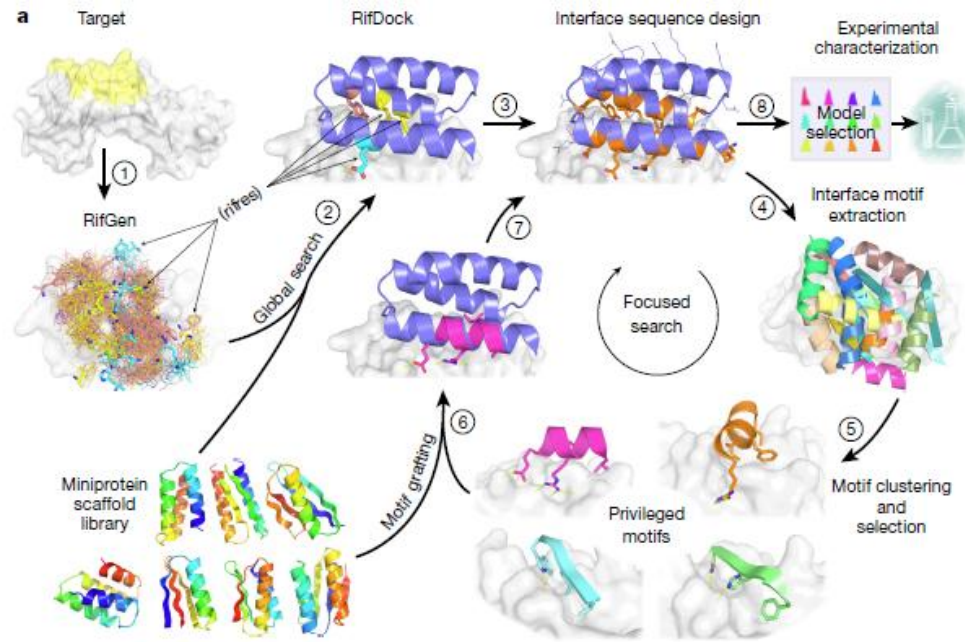
Anchor extension: a structure-guided approach to design cyclic peptides



P. Hosseinzadeh, P. R. Watson, T. W. Craven, X. Li, S. Rettie, F. Pardo-Avila, A. K. Bera, V. K. Mulligan, P. Lu, A. S. Ford, B. D. Weitzner, L. J. Stewart, A. P. Moyer, M. Di Piazza, J. G. Whalen, P. J. Greisen, D. W. Christianson and D. Baker; "Anchor extension: a structure-guided approach to design cyclic peptides targeting enzyme active sites"; *Nat Commun*; **2021**; *Vol. 12 (1)*: p. 3384.

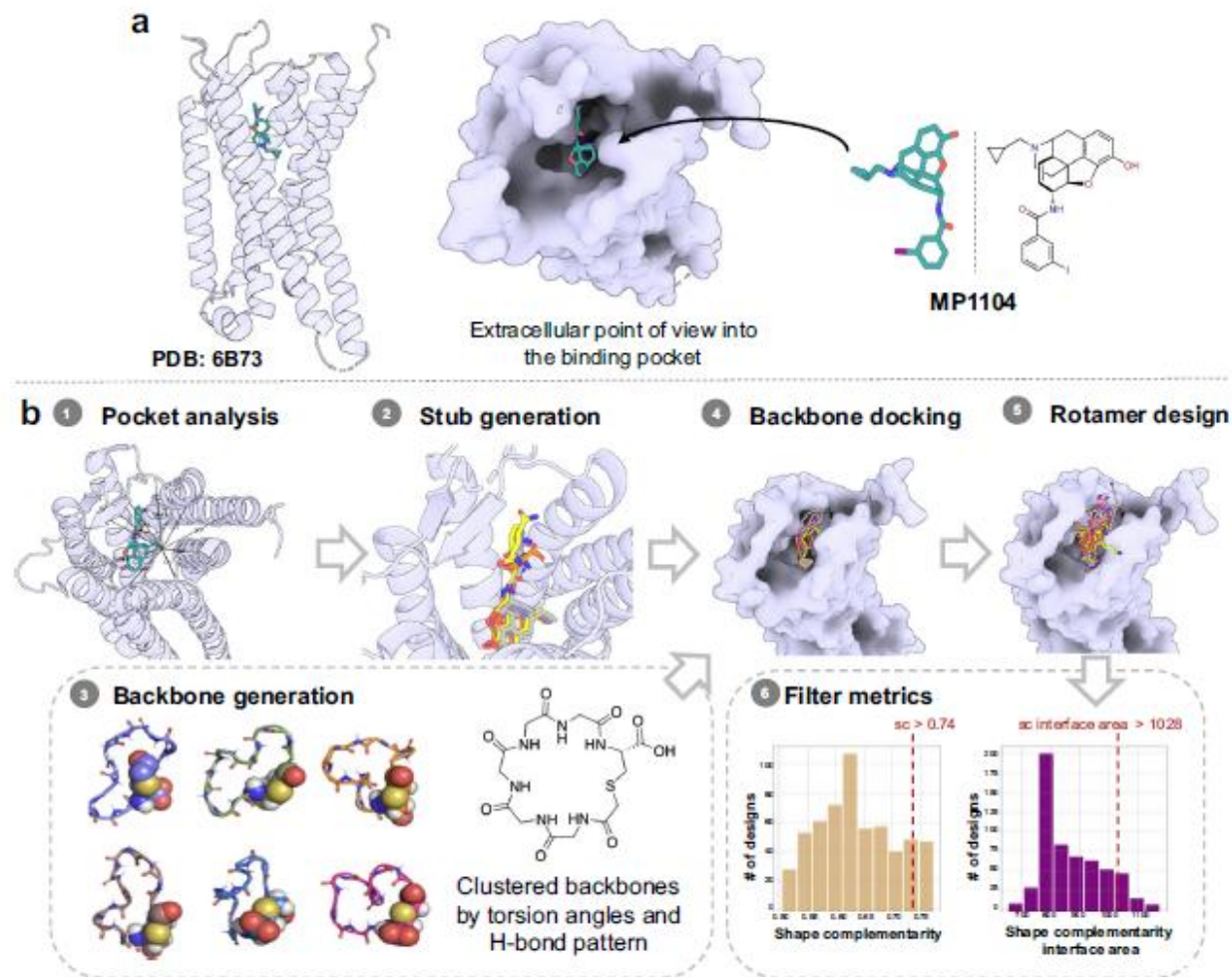


Design of protein-binding proteins from the target structure alone



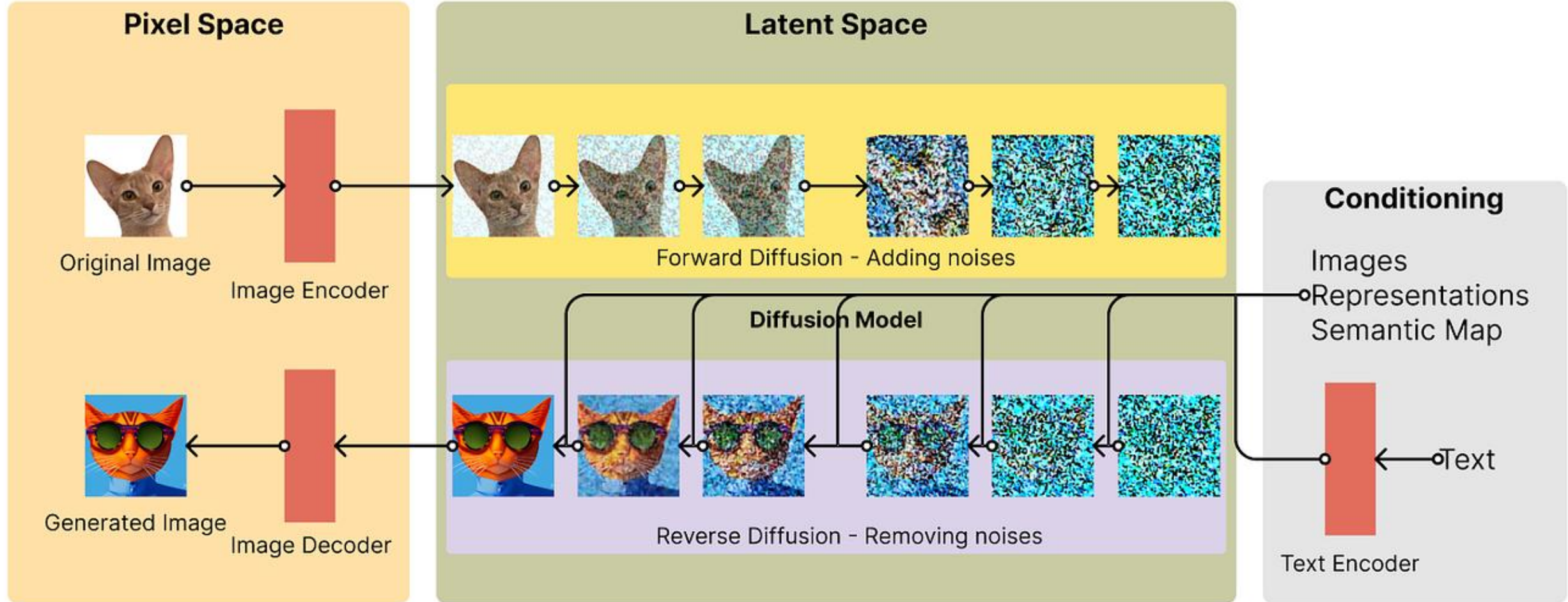
Cao, L. X., Coventry, B., Goreshnik, I., Huang, B. W., Sheffler, W., Park, J. S., Jude, K. M., Markovic, I., Kadam, R. U., Verschueren, K. H. G., Verstraete, K., Walsh, S. T. R., Bennett, N., Phal, A., Yang, A., Kozodoy, L., DeWitt, M., Picton, L., Miller, L., . . . Baker, D. (2022). Design of protein-binding proteins from the target structure alone. *Nature*, 605(7910), 551-+.

Design of peptide-drug conjugate ligands of the kappa-opioid receptor



Muratspahic, E., Deibler, K., Han, J., Tomasevic, N., Jadhav, K. B., Olive-Marti, A. L., Hochrainer, N., Hellinger, R., Koehbach, J., Fay, J. F., Rahman, M. H., Hegazy, L., Craven, T. W., Varga, B. R., Bhardwaj, G., Appourchaux, K., Majumdar, S., Muttenthaler, M., Hosseinzadeh, P., . . . Gruber, C. W. (2023). Design and structural validation of peptide-drug conjugate ligands of the kappa-opioid receptor. *Nat Commun*, 14(1), 8064.

How Stable Diffusion works in a Nutshell



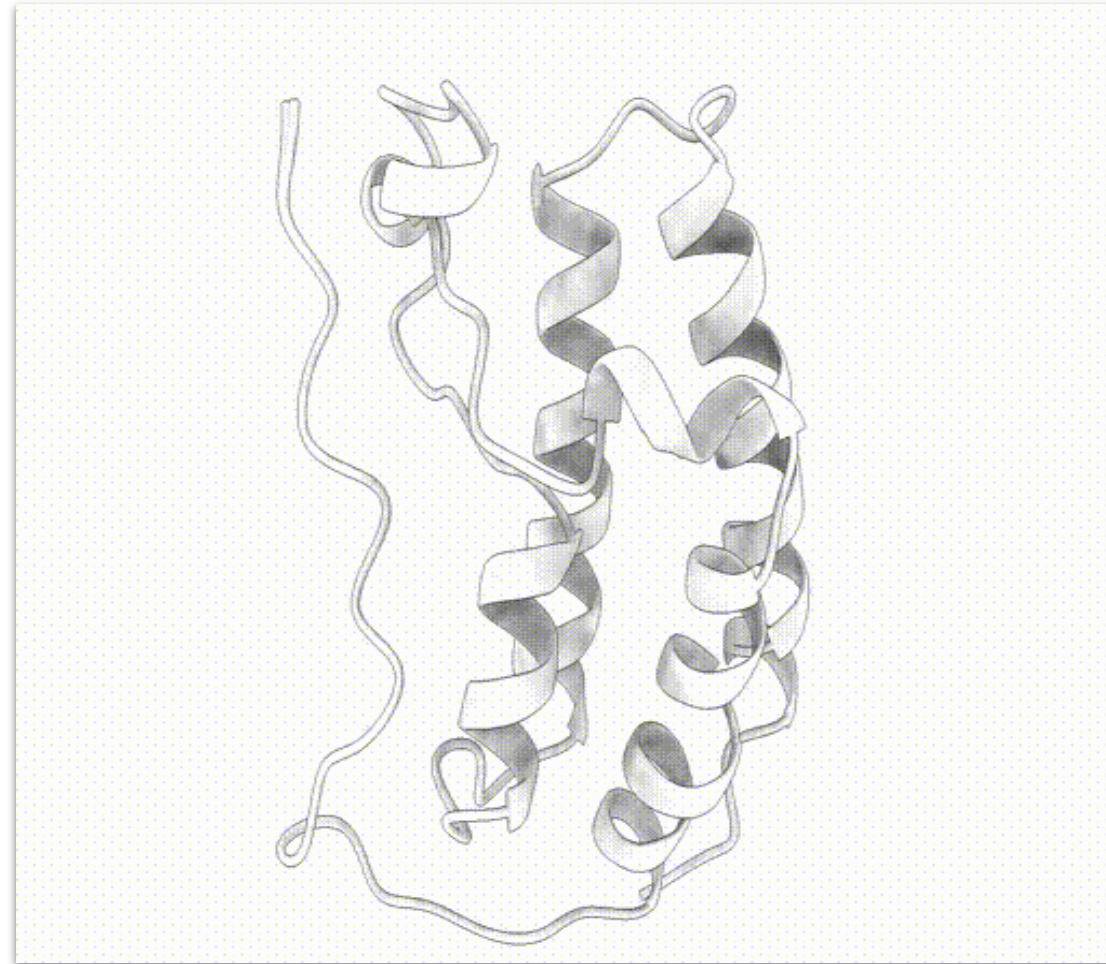
Modified Version of RFDiffusion for (Cyclic) Peptides

Normal positional embedding

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-3	-2	-1	0	1	2	3	4
-4	-3	-2	-1	0	1	2	3
-5	-4	-3	-2	-1	0	1	2
-6	-5	-4	-3	-2	-1	0	1
-7	-6	-5	-4	-3	-2	-1	0

Cyclic positional embedding

0	1	2	3	-4	-3	-2	-1
-1	0	1	2	3	-4	-3	-2
-2	-1	0	1	2	3	-4	-3
-3	-2	-1	0	1	2	3	-4
-4	-3	-2	-1	0	1	2	3
3	-4	-3	-2	-1	0	1	2
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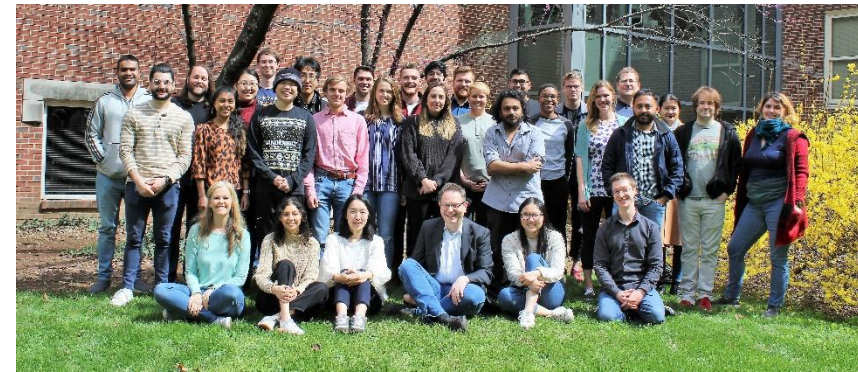
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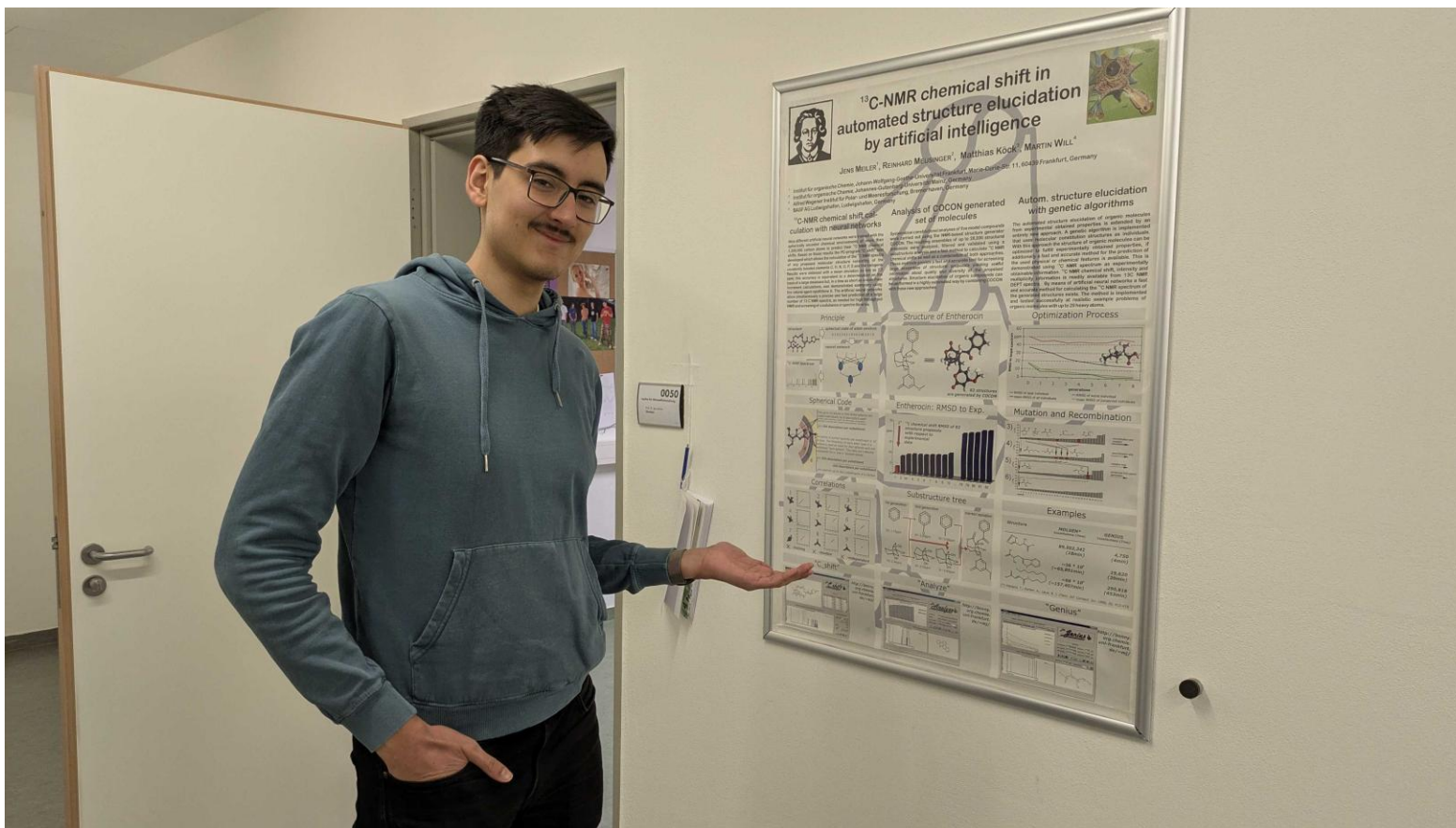


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- cancer & immunotherapy
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- chemical sensors
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Sustainability

- breaking down plastic
- artificial photosynthesis
- nanoscale manufacturing