

Accurate de novo design of high-affinity protein-binding macrocycles using deep learning

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 Check for updates

Stephen A. Rettie ^{1,2,3,15}, David Juergens ^{2,4,15}, Victor Adebomi^{1,2,15}, Yensi Flores Bueso ^{1,2,5,6}, Qinqin Zhao⁷, Alexandria N. Leveille⁸, Andi Liu ⁷, Asim K. Bera ², Joana A. Wilms ^{9,10}, Alina Üffing ^{9,10,14}, Alex Kang ², Evans Brackenbrough², Mila Lamb², Stacey R. Gerben ², Analisa Murray ², Paul M. Levine ², Maika Schneider ^{1,2,11}, Vibha Vasireddy^{1,2}, Sergey Ovchinnikov¹², Oliver H. Weiergräber¹⁰, Dieter Willbold ^{9,10}, Joshua A. Kritzer ⁸, Joseph D. Mougous ^{7,13}, David Baker ^{2,5,13} , Frank DiMaio ^{2,5}  & Gaurav Bhardwaj ^{1,2} 

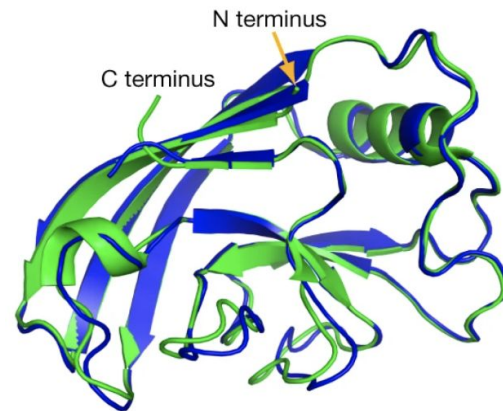
A bit of background

Sorry in case you already know it

Sequence → Structure

The protein folding revolution: AlphaFold2

- General solution to protein structure prediction
- Predicts accurate protein structures in minutes
 - As opposed to months - years for experimental determination



AlphaFold Experiment
r.m.s.d._{g5} = 0.8 Å; TM-score = 0.93

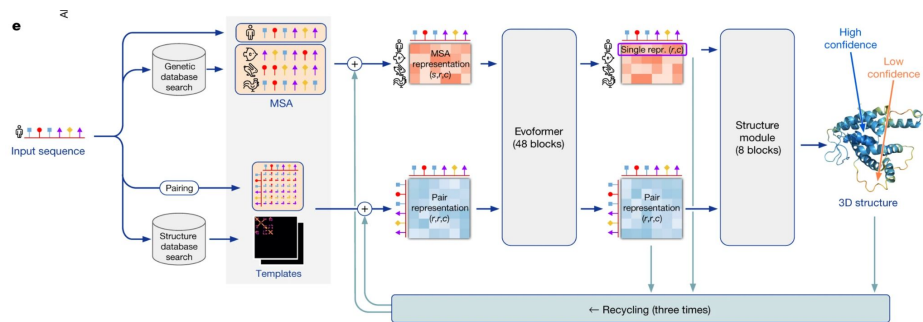
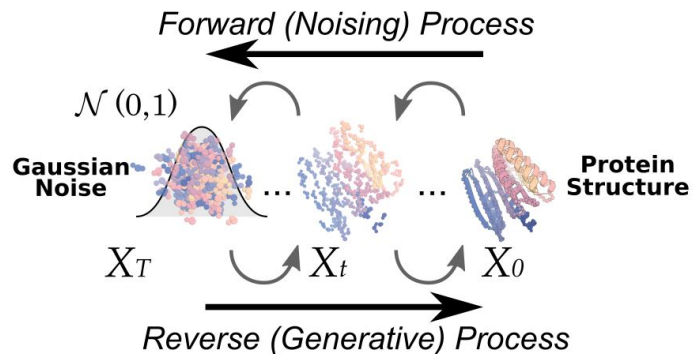
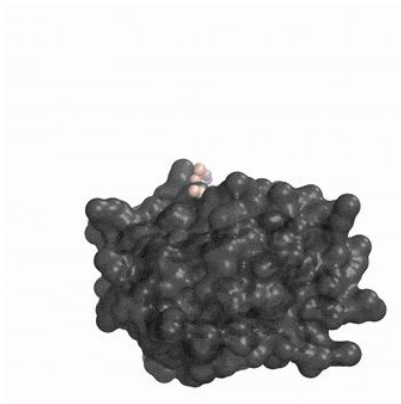


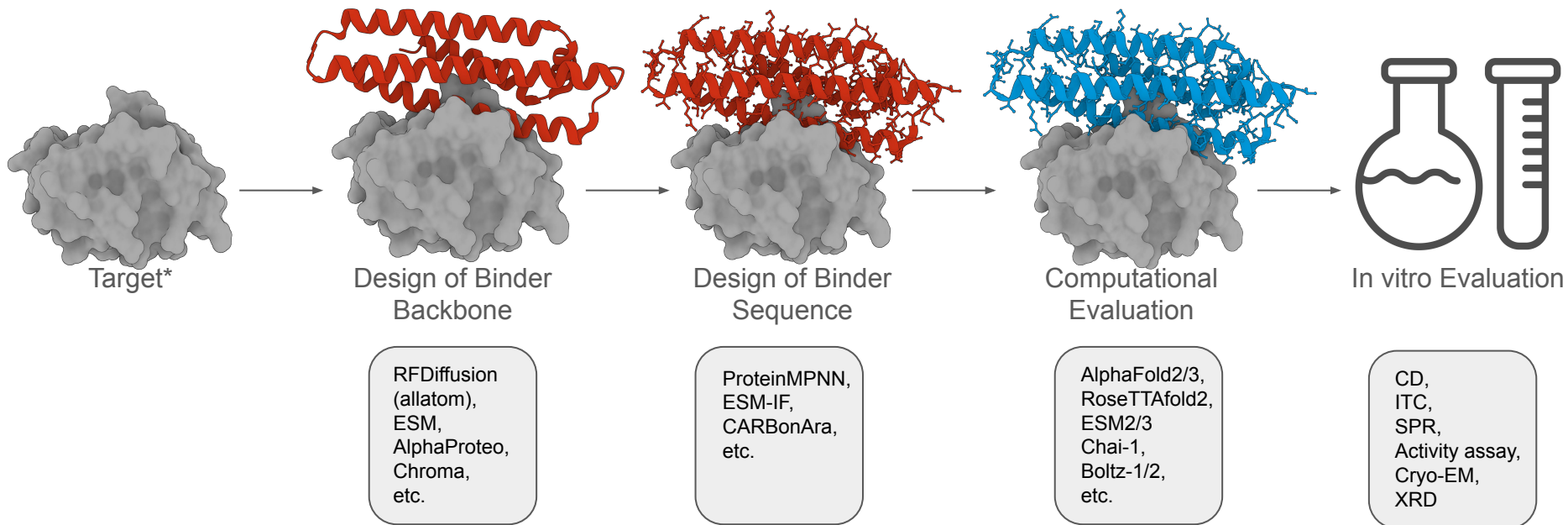
Image from: Jumper, J. *et al.* Highly accurate protein structure prediction with AlphaFold. *Nature* **596**, 583–589 (2021).

Very short intro to RFDiffusion

- DDPM on top of a structure prediction network
- conditioned on the target structure
- will yield backbone of binding protein

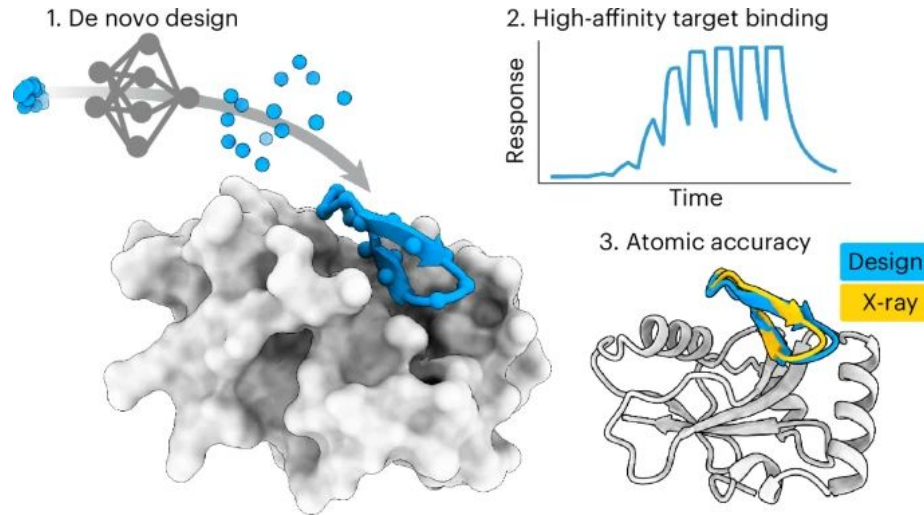


What is Protein Design - all together



* Target can also be of small molecule or nucleic acid type

The paper



Showcase

Like in a cooking show

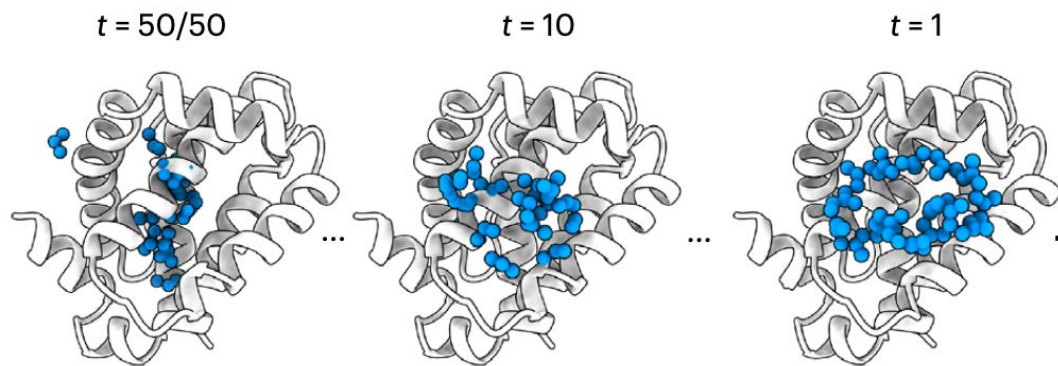
Challenges for current Macrocyclic design

- time and cost-intensive HTS for discovery
- low mutational tolerance for hits
- multi-objective optimisation not possible
 - target binding
 - selectivity and membrane permeability

The RFpeptides pipeline

e

Multimeric diffusion



MPNN



Prediction

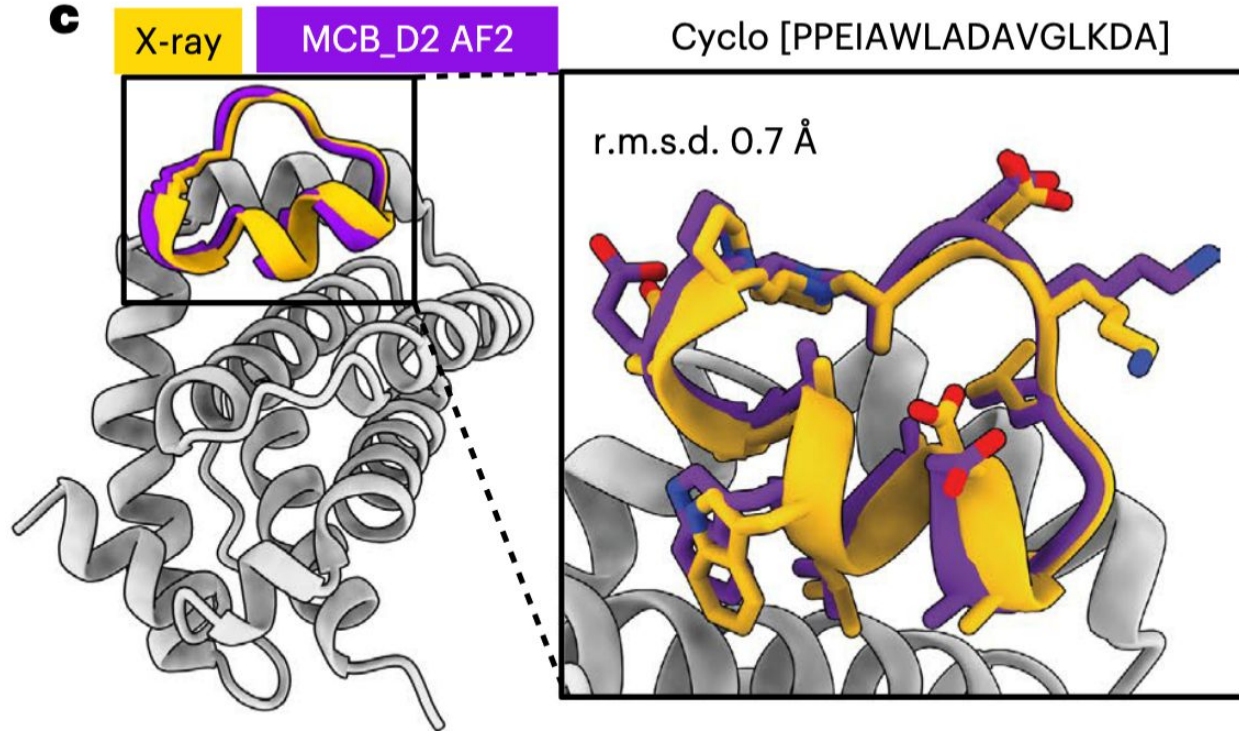


The design campaign

- myeloid cell leukemia 1 (MCL1) protein
 - part of the Bcl2 protein family (blocking cell death)
 - regulatory role in autophagy, cell survival, DNA repair and cellular proliferation
 - Aim: Inhibiting MCL1 to bind other proteins

- 9,965 structures → ProteinMPNN and AfCycDesign →
- 1,984 structures → physics-based in silico filtering →
- 27 designs for Fmoc-based solid-phase synthesis →
- 14 designs with acceptable yields→
- 3 designs binding →
- 1 best binder: 2 μ M (SPR)

The design campaign



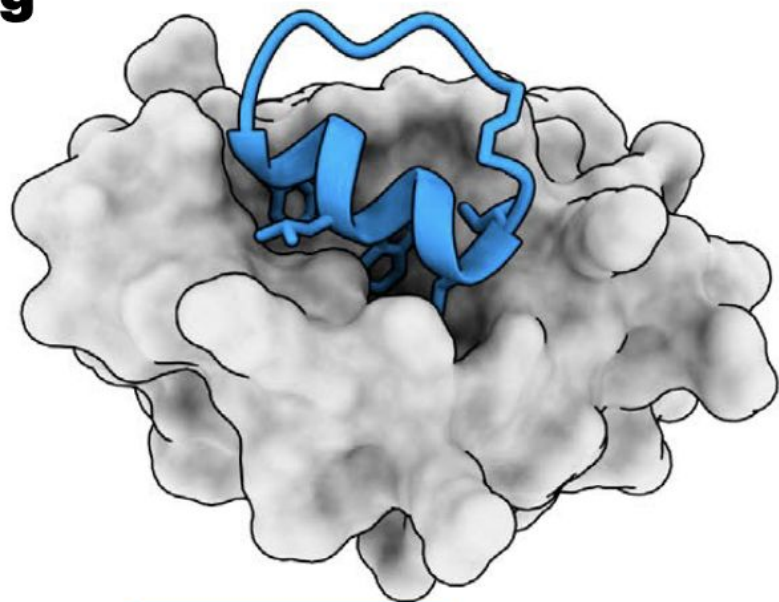
The design campaign

- MDM2
 - E3 ligase that interacts with tumor suppressor protein p53
 - critical roles in tumor growth and survival
 - Aim: Improve binding of ligase

- 40,000 structures → ProteinMPNN and AfCycDesign →
- 7,495 structures → physics-based in silico filtering →
- 11 designs for Fmoc-based solid-phase synthesis →
- 8 designs with acceptable yields →
- 3 designs binding →
- 1 best binder: 1.9 μM (SPR)

The design campaign

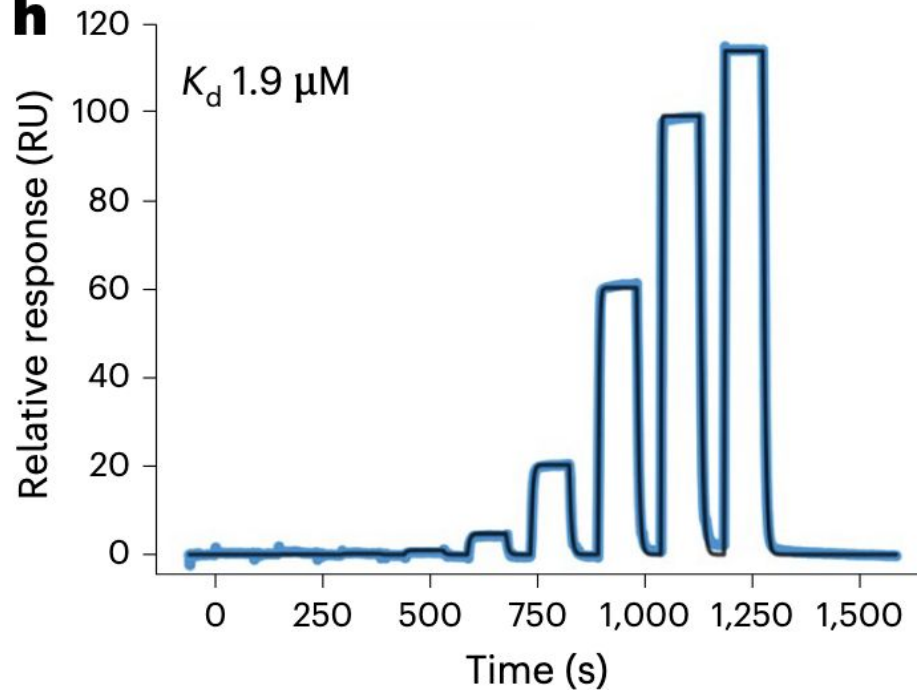
g



MDB_D8 AF2

MDM2

h



MDB_D8

Global fit

The design campaign

- γ -aminobutyric acid type A receptor-associated protein (GABARAP)
 - mediating autophagy through its role in autophagosome biogenesis and recruitment of cargo
 - resulting in lysosomal degradation of damaged or surplus proteins and organelles
 - therapeutic applications: late stage cancer, protein degradation

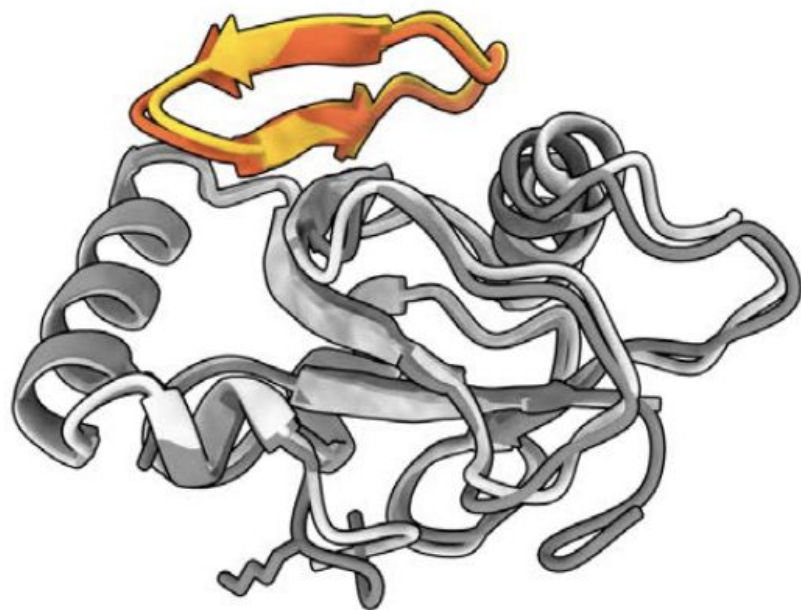
- 80,000 structures \rightarrow ProteinMPNN and AfCycDesign + clustering \rightarrow
- 335 structures \rightarrow physics-based in silico filtering \rightarrow
- 13 designs for Fmoc-based solid-phase synthesis \rightarrow
- 6 designs with acceptable yields \rightarrow
- 2 designs binding \rightarrow
- 1 best binder: 6 nM (SPR) \rightarrow capable of inhibiting the interface with K1

c

X-ray

GABARAPL1

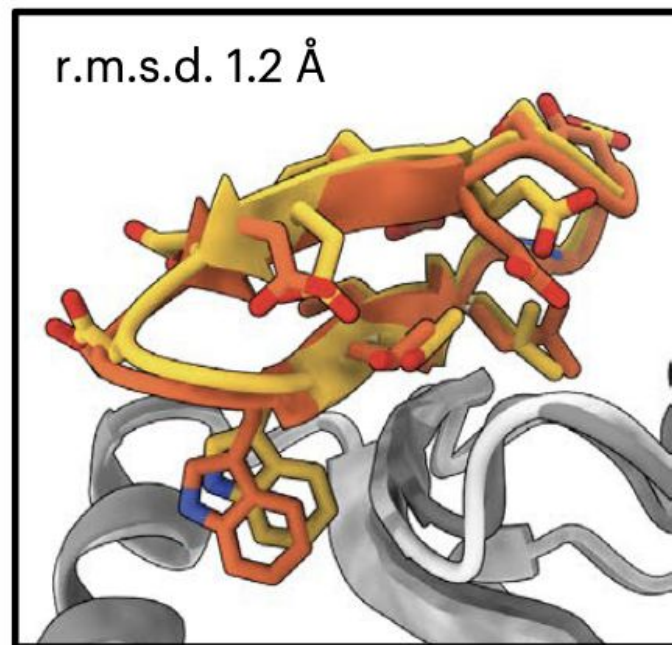
GAB_D8



AF2

GABARAP

GAB_D8



In short

- Achieves high-affinity binders with fewer than 20 candidates per target (vs. trillions in library-based approaches)
- Enables targeted design for specific protein patches and sites
- Provides atomically accurate models for structure-guided optimization
- Allows optimization for multiple properties simultaneously (binding affinity, membrane permeability)
- Bypasses complex structure determination bottlenecks

They do not address

- Oral availability
- Expensive synthesis
- Lack of broader applicability (only 3 targets)
- Off-target interactions
- Metabolic stability
- Pharmacokinetics, Immunogenicity, ADMET, Toxicity etc. (*)

Showcase

Present the results