



RFdiffusion

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Article

De novo design of protein structure and function with RFdiffusion

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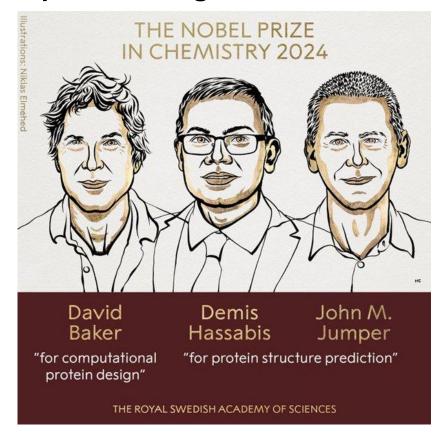
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There has been considerable recent progress in designing new proteins using deep-learning methods^{1–9}. Despite this progress, a general deep-learning framework for protein design that enables solution of a wide range of design challenges, including de novo binder design and design of higher-order symmetric architectures, has yet to be described. Diffusion models^{10,11} have had considerable success in image and language generative modelling but limited success when applied to protein modelling, probably due to the complexity of protein backbone geometry and sequence–structure relationships. Here we show that by fine-tuning the RoseTTAFold structure prediction network on protein structure denoising tasks, we obtain a generative model of protein backbones, that achieves outstanding performance on unconditional and topology-constrained protein monomer design, protein binder design, symmetric oligomer design, enzyme active site scaffolding and symmetric motif scaffolding for therapeutic



Why should we care about de novo protein design?

- Enable new functions
- Tune biochemical parameters
- Control protein function
- Create composite molecular machines, assemblies

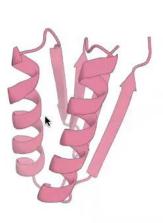




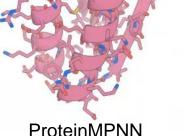
Backbone generation is a bottleneck in the protein design workflow

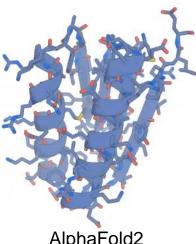
Backbone Generation Sequence Design Computational Filtering

Experimental Characterisation

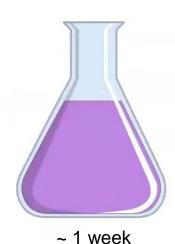


RFdiffusion









· Wook

Figure from video "RFdiffusion" by SBGrid Consortium.



Outline

- 1. What is diffusion?
- 2. Introduction to RFdiffusion
- 3. How to use RFdiffusion



Diffusion models introduce random noise iteratively and learn to reverse the process during training

Fixed forward diffusion process

Data

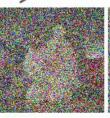














Noise

Generative reverse denoising process

https://cvpr2022-tutorial-diffusion-models.github.io/

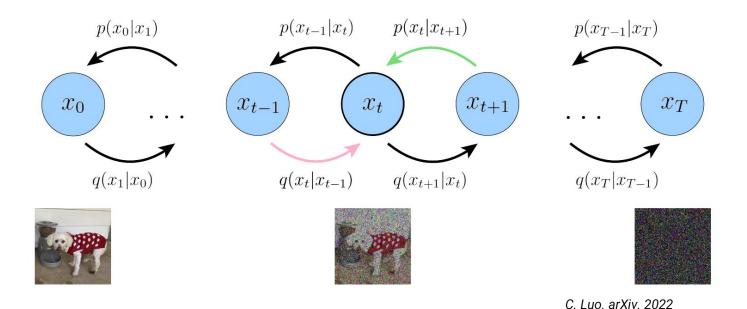
"Generate an image of a banana eating a hamburger while tap dancing."







Denoising diffusion probabilistic models



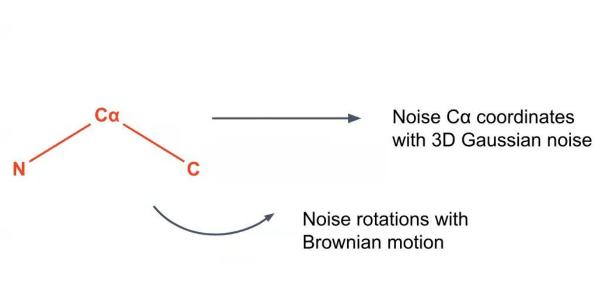


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Diffusion occurs across translations and rotations



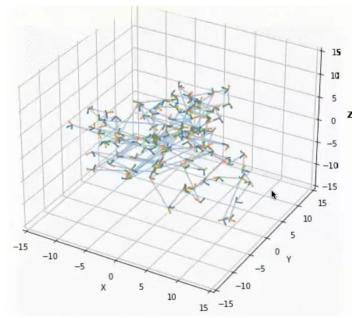
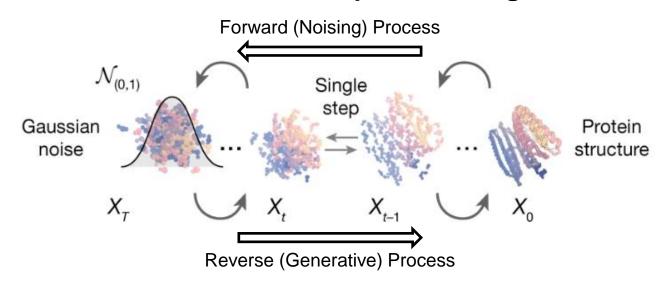


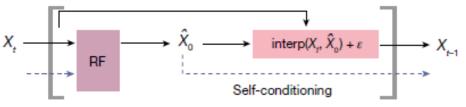
Figure from "RFDiffusion: Accurate protein design using structure prediction and diffusion generative models" video.



Utilizing diffusion models for use in protein design

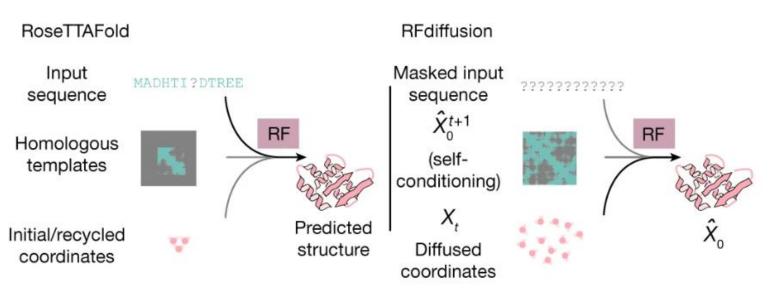


Single RFdiffusion step





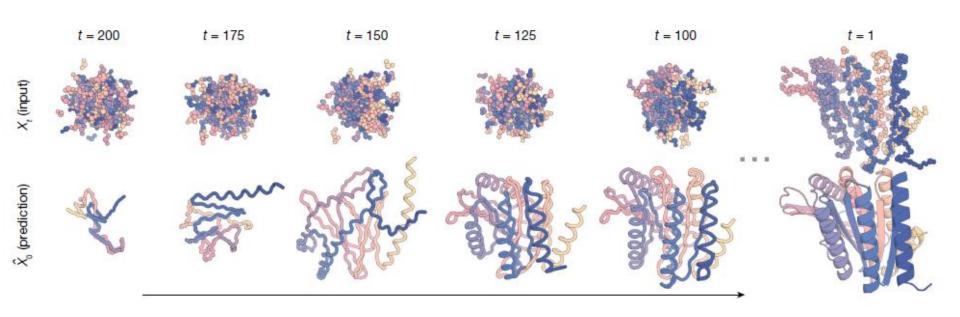
RoseTTAFold is used as the denoising framework



Watson et al. Nature, 2023.



Unconditional design trajectory for a 300-residue chain

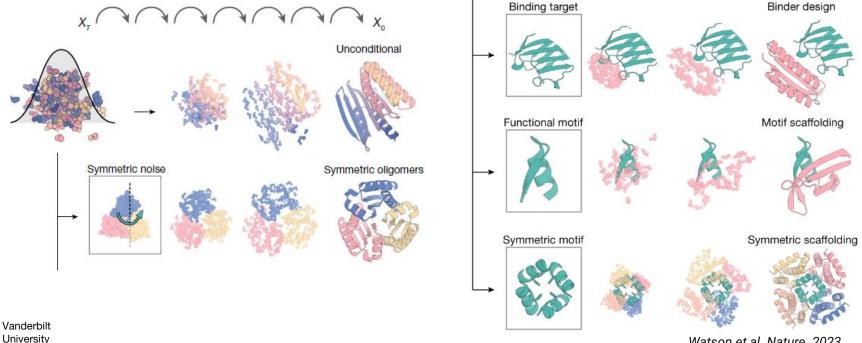


Watson et al. Nature. 2023.



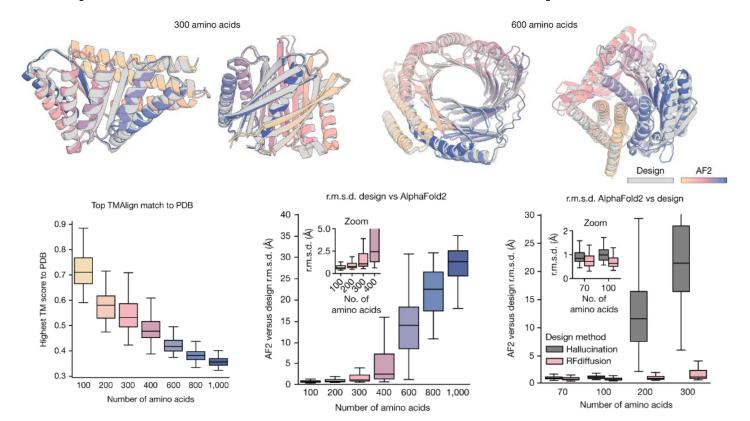
Broad applications of RFdiffusion

- "Generate a protein that has 300 amino acids."
- "Generate a protein that can bind to this target."



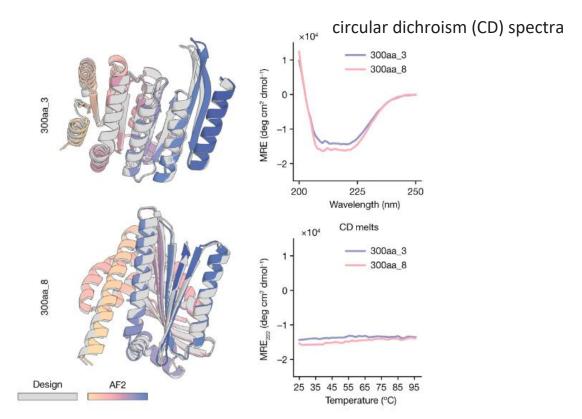


RFdiffusion produces diverse and new-to-nature proteins



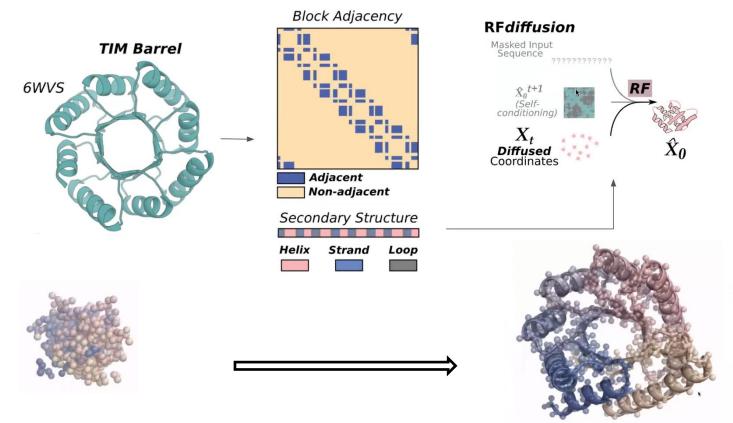


Unconditional designs expressed as soluble monomers



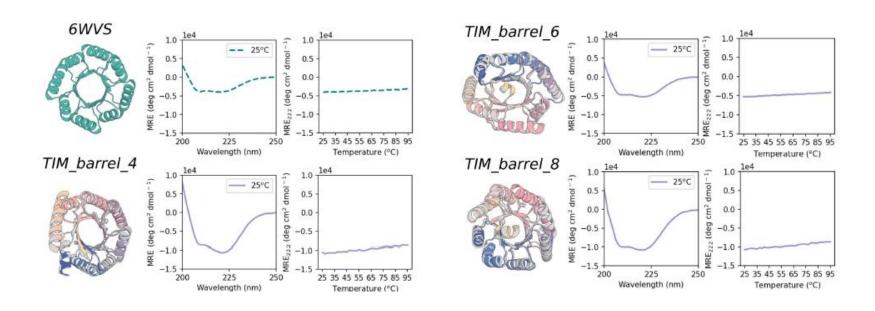


Condition designs to get a specific type of fold



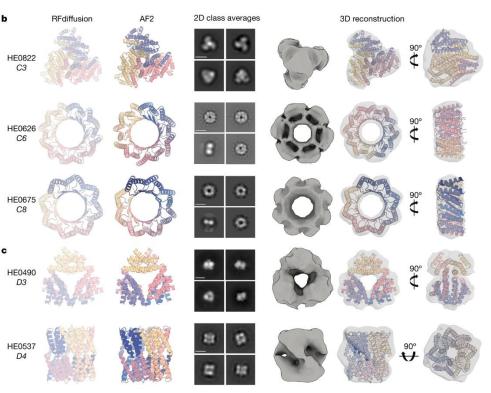


Example TIM barrel-conditioned designs capture fold and are diverse

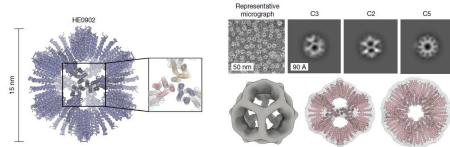


Symmetric oligomer design





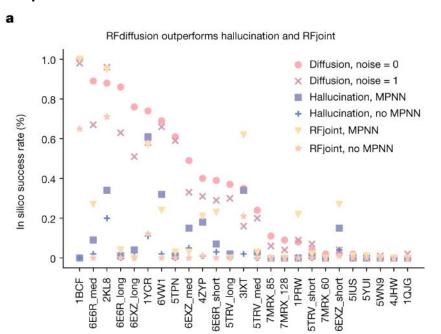
Design of icosahedral particle:

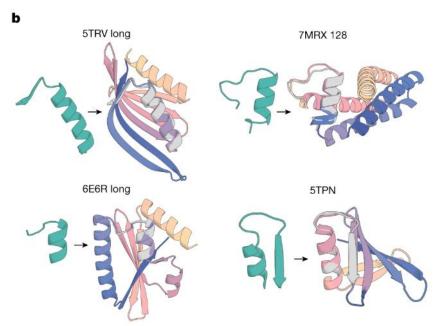




Generate proteins with given functional motifs

Given a functional motif, generate a protein which captures the structure of the input functional motif.







Designed binders show higher affinity for target than native binder

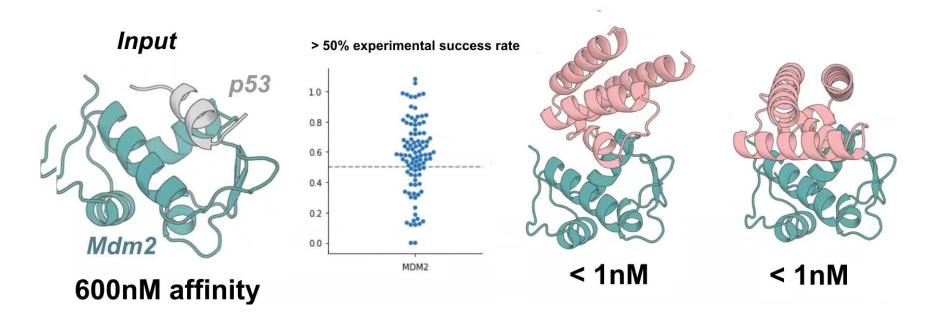
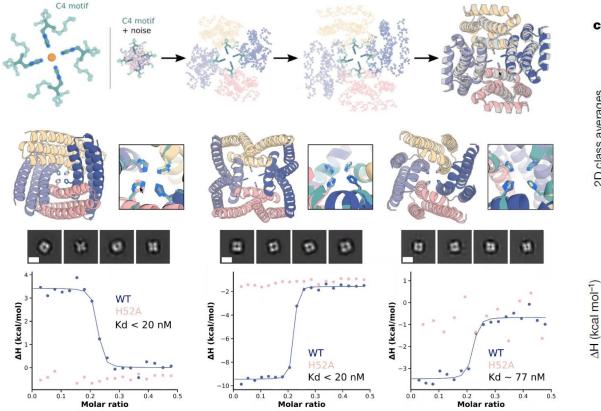
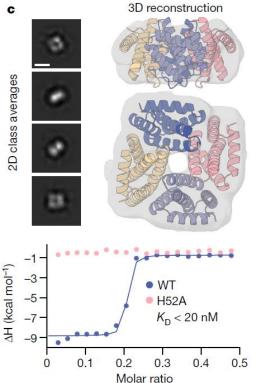


Figure from "RFDiffusion:
Accurate protein design using
structure prediction and diffusion
generative models" video.



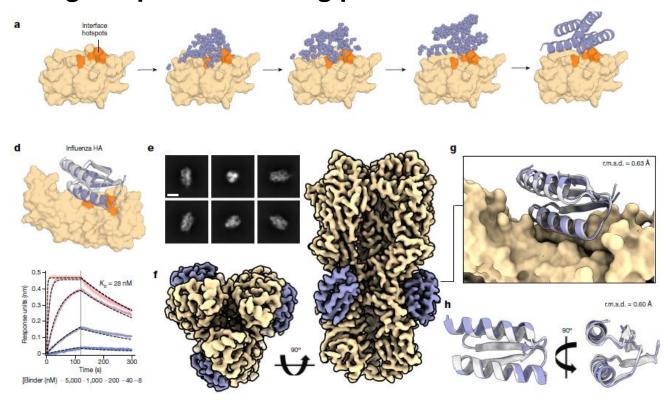
Symmetric metal-binding oligomers







De novo design of protein-binding proteins





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How to use RFdiffusion?

Activate your conda environment!

- Example for unconditional monomer generation:

```
../RFdiffusion/scripts/run_inference.py \
'contigmap.contigs=[150-150]' \
inference.output_prefix=./output_monomer/monomer \
inference.num_designs=5
```

Call the code: ~/rosetta_workshop/RFdiffusion/scripts/run_inference.py

- Specify length of the protein with **contigmap.contigs**Example: 'contigmap.contigs = [150-150]'
- Specify the output
 - Example: inference.output_prefix=path_to_output
- Set the number of backbones you want to generate
 - Example: inference.num_designs = 5



Output of unconditional monomer generation

Output files are in **output_monomer** folder.

All outputs are poly-Glycine sequence.

- PDB file contains generated backbone
- .trb file, contains all metadata with the run, including input options
- traj folder, contains trajectory files (pdb)

Use PyMOL for visualization

pymol monomer_0.pdb



Motif scaffolding

Example for generation conditioned on input motif:

```
../RFdiffusion/scripts/run_inference.py \
'contigmap.contigs=[20-30/E422-440/40-50]' \
inference.input_pdb=6UYG.pdb \
inference.output_prefix=./output_scaffold/scaffold \
inference.num_designs=1
```

- Specify motifs using 'contigmap.contigs'
 Example: 'contigmap.contigs=[20-30/E422-440/40-50]'
 - Motif is prefixed by letter followed by residue specification (**E422-440**)
 - Build a backbone with 20-30 residues at the N-terminal of the motif, then 422-440 from chain E of the input, followed by 40-50 residues
- Specify input PDB with 'inference.input_pdb=path'



Motif scaffolding with additional chains to avoid clashes

- Example where we can input structure from additional chains to avoid clashes
- Use /0 followed by a space to symbolize a chain break

```
cd ../../
../RFdiffusion/scripts/run_inference.py \
'contigmap.contigs=[10-20/E422-440/40-50]/0 H4-22/H29-71/H77-113]/0
L2-29/L31-95/L97-109]' \ Light chain info
inference.input_pdb=6UYG.pdb \
inference.output_prefix=./output_scaffold/scaffold_Ab \
inference.num_designs=1
```

Thank You!

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