

# MASSIVEFOLD

Practical work

Protein complex

IFB training - AlphaFold and beyond - 10th to 12th December 2025

# H1140

The H1140 CASP target is a nanobody/antigen target.

These targets are known to be hard to predict, certainly because of a lack of evolutionary signals.

We are going to model them with the 3 tools and all the neural network versions of AFmassive and ColabFold.

- Load MassiveFold 1.5

```
module load massivefold/1.5.0
```

- Edit the 'AFmassive\_params.json' and 'ColabFold\_params.json' files, change the "model\_preset" to "multimer" and add "recycles" to the plots (you may want to copy these files to another name which contains "\_multimer" for instance, to keep both file types, for monomers and multimers)
- The H1140.fasta is already present in the input folder because this is the example provided
- Run AFmassive and ColabFold:

```
./run_massivefold.sh -s input/H1140.fasta -r afm_default -p 5 -f AFmassive_params.json
```

```
./run_massivefold.sh -s input/H1140.fasta -r cf_default -p 5 -f ColabFold_params.json
```

For AlphaFold3, you need to specify that the fasta file contains 2 proteins

Edit the 'AlphaFold3\_params.json' file by setting:

- "fasta\_chains" to 2 protein chains
- "model\_preset" to "multimer"

```
"AF3_run": {  
  "fasta_chains": ["protein", "protein"],  
  "ligand": [  
    {"ccdCodes": [""], "smiles": ""}  
  ],  
  "modifications": [  
    [{"type": "", "sequence": "", "positions": []}]  
  ],  
  "model_preset": "multimer",  
  "max_template_date": "2024-11-28",  
  "num_diffusion_samples": "5",  
  "unpairedMsa": "true",  
  "pairedMsa": "true",  
  "templates": "true"  
},
```

## - Run AlphaFold3

```
./run_massivefold.sh -s input/H1140.fasta -r af3_default -p 5 -f AlphaFold3_params.json
```

# H1140 - Results

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- Get the results on your local computer to visualize them
- Alternatively, you can get or visualize them here

<https://nextcloud.univ-lille.fr/index.php/s/dnQLmBywbcQdx56>

# H1140 - Results

Compare the results for the 3 tools, looking at the 'ranking\_iptm.json' and the plots.

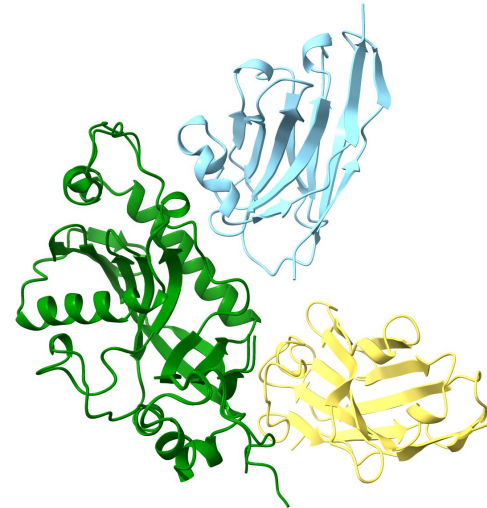
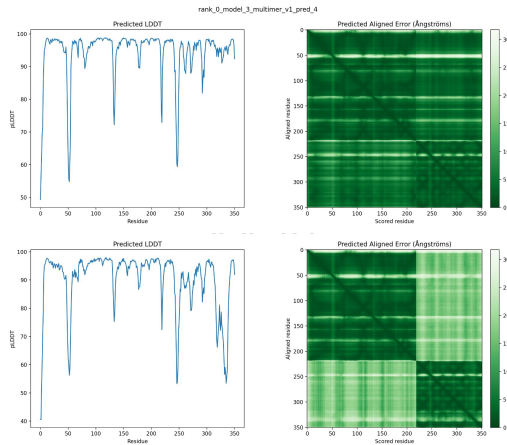
=> Which tool seems to give the best predictions between AFmassive and ColabFold ?

=> Which neural network version provides this prediction ?

=> What about AlphaFold3 predictions ?

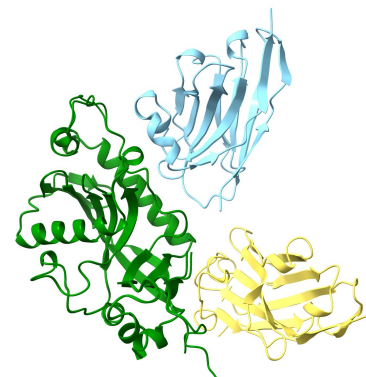
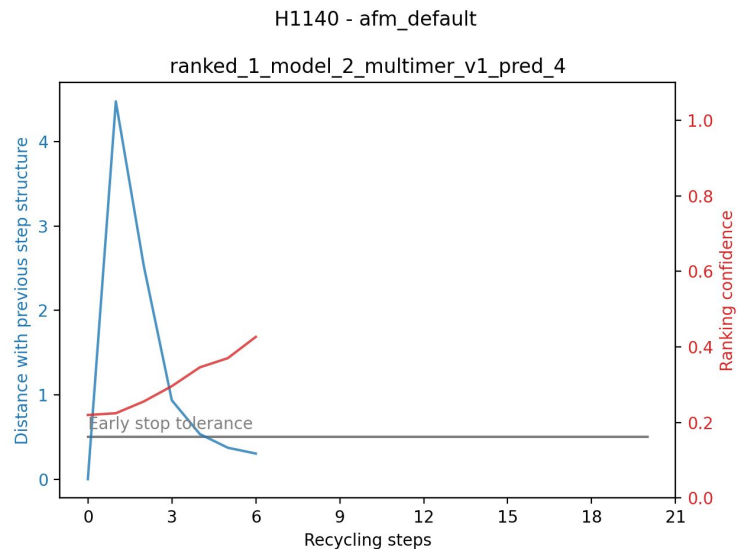
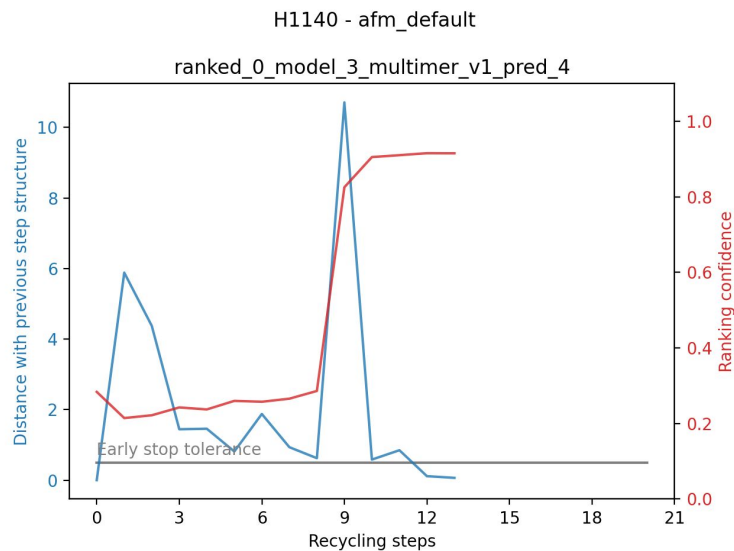
=> Open a high confident prediction in ChimeraX and align it to the released PDB 9ERT to visualize the quality of the prediction. Align the next prediction of AFmassive or ColabFold with a less good PAE matrix.

What do you see ?



# H1140 - Results

Look at the first recycles plots of AFmassive and ColabFold. What do you see ?



# T231

T231 is an antibody-antigen target, made of the heavy and light chains bound to a peptide.

- Get the T231.fasta sequence and perform the same runs with MassiveFold as for H1140

```
./run_massivefold.sh -s input/T231.fasta -r afm_default -p 5 -f AFmassive_params.json  
./run_massivefold.sh -s input/T231.fasta -r cf_default -p 5 -f ColabFold_params.json  
./run_massivefold.sh -s input/T231.fasta -r af3_default -p 5 -f AlphaFold3_params.json
```

**N.B.:** for AlphaFold3, add 3 times “protein” in the “fasta\_chains”

# T231 - Results

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- Get the results on your local computer to visualize them
- Alternatively, you can get them here

<https://nextcloud.univ-lille.fr/index.php/s/dnQLmBywbcQdx56>



# T231 - Results

- Compare the results of the 3 tools, looking at the 'ranking\_ipTM.json' and the plots.

=> What can you say about the results for AFmassive and ColabFold ?

=> Why do the ipTM scores are so homogeneous ?

