## AlphaFold: introduction + how-to

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Internship Major Systems Biology











## The Protein folding problem



#### Available sequences and structures

After filtering for redundancy

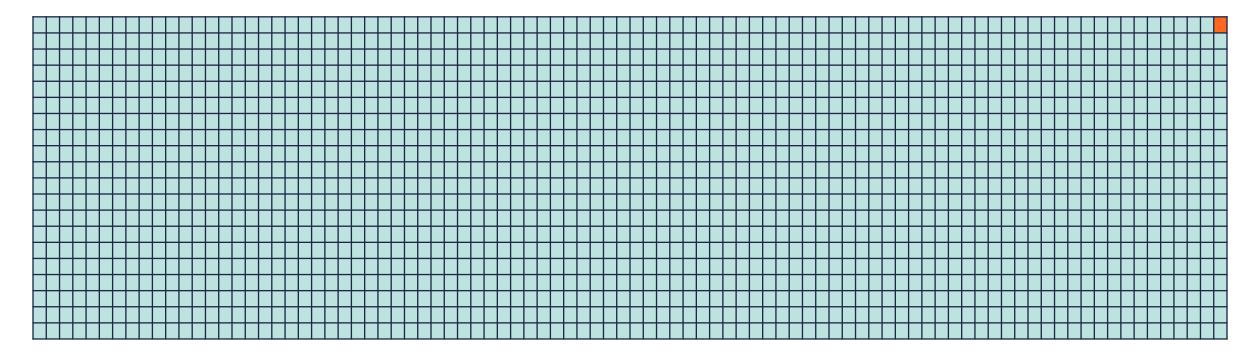
(50% similarity) --- (May 2023)

UniRef50

Protein Data Bank (PDB)

60 million unique sequences

~60 000 unique structures



 $\square$  = 60 000 proteins

### Main protein structure resource



~60 ooo unique structures

Main experimental technique: X-ray crystallography

- Costly
- Labour-intensive
- Slow

(NMR)



Image source: https://www.prweb.com/releases/2017/04/prweb14219956.htm

### Why 3-D structure prediction? Low effort, time efficient

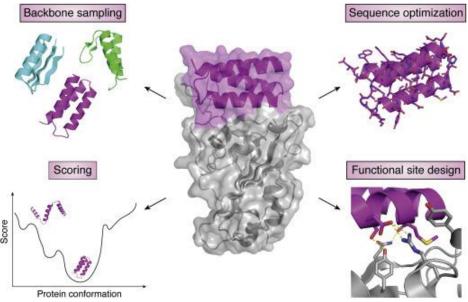
#### Examples:

- In silico drug candidate screening



- De novo protein design guidance

- +++



Source: https://www.jbc.org/article/Soo21-9258(21)00336-7/fulltext

## Structure prediction: homology models

Basic principle: 3-D structure more conserved than sequence

Query protein sequence: ELAIGILTVSYIPSAEKIRAPELTI

Sequence alignment:

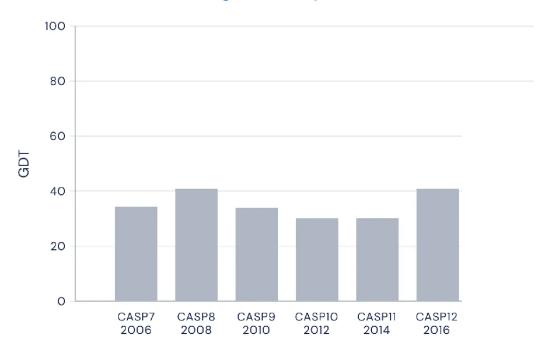
ELA-IGILTVSYIPSAEKIRAP--ELTI

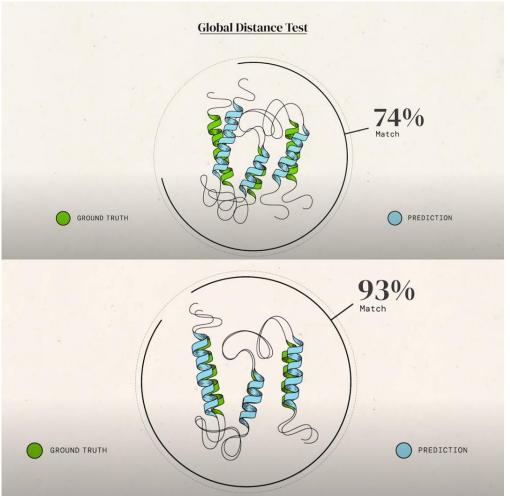
ELAGI-ILGVSYIPSAEKI-ARACELTI → structure in PDB?

Finetuning using statistical potentials and physics-based energy calculations

#### **Critical Assessment of Structure Prediction**

#### Median Free-Modelling Accuracy



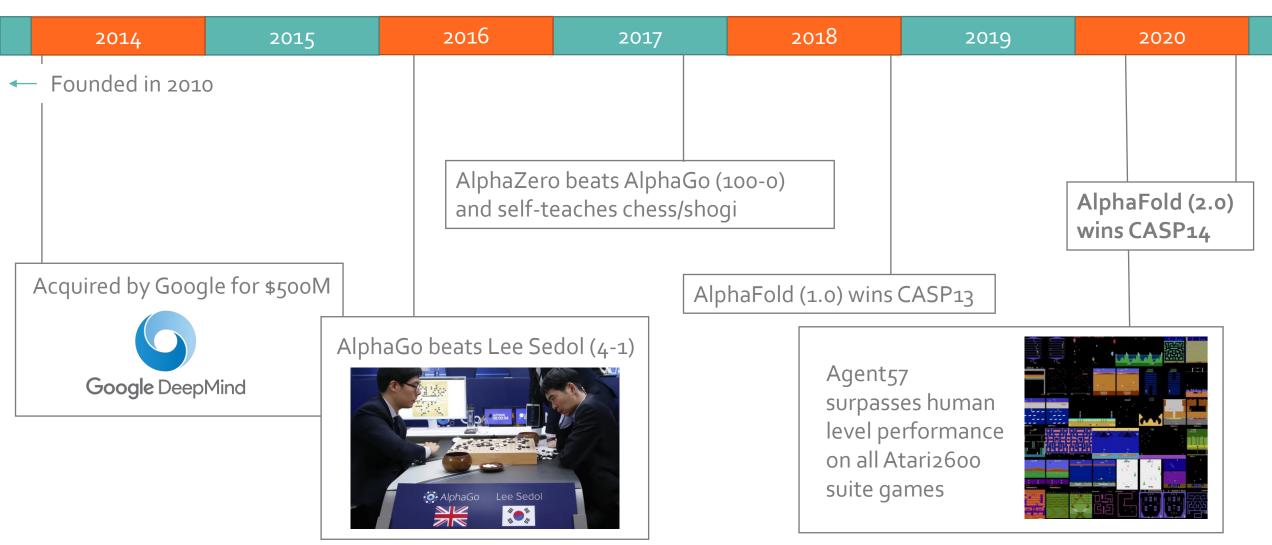


Source: https://news.machinelearning.sg/posts/alphafold2\_1o\_things\_you\_want\_to\_know\_about\_biologys\_imagenet\_moment/

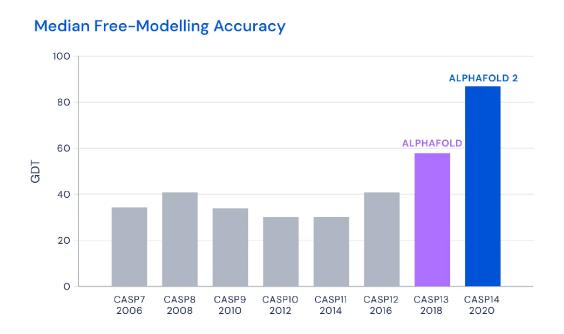
## AlphaFold introduction

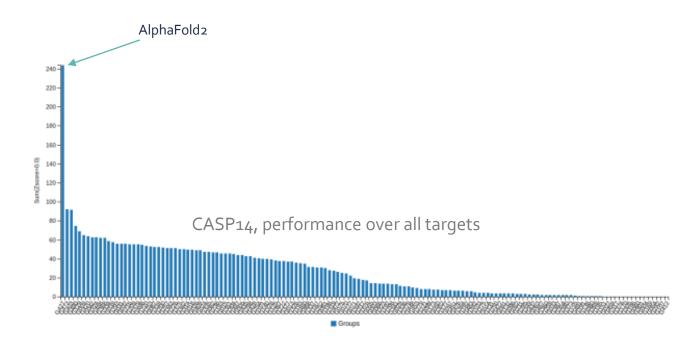


### DeepMind: a timeline



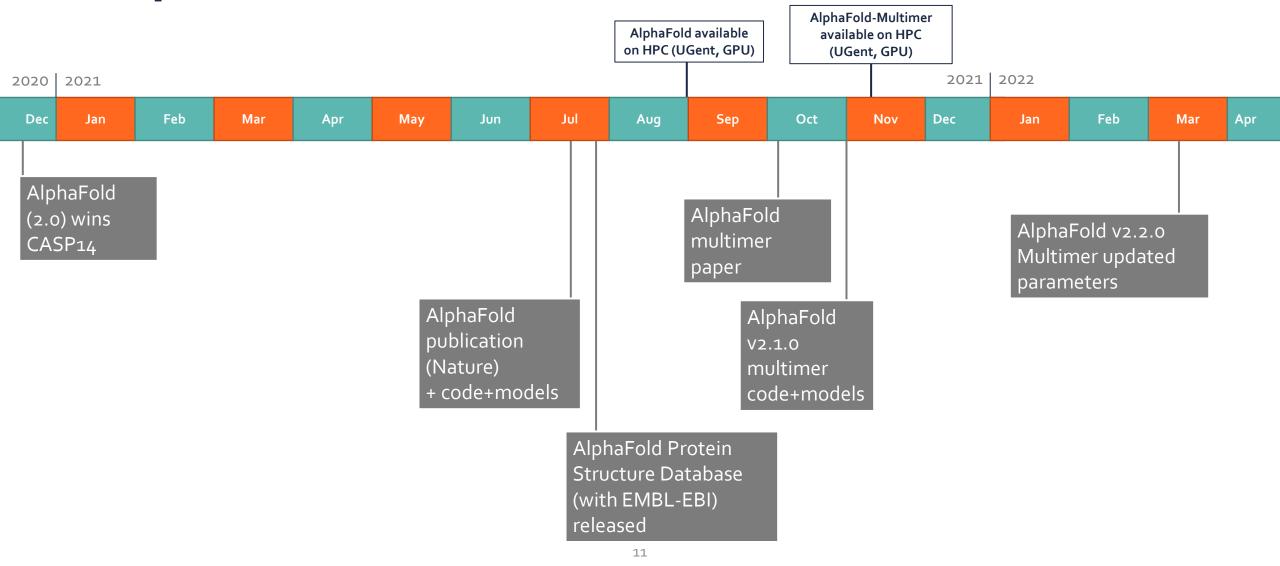
#### Critical Assessment of Structure Prediction (CASP14, 2020)



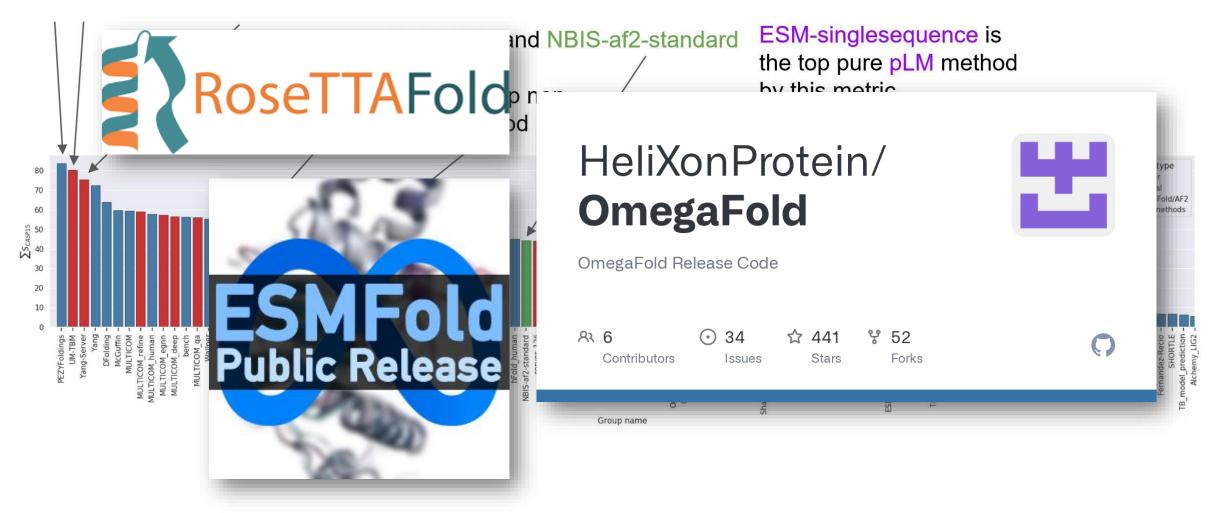


Median of 92.4 across all targets, 87.0 in free-modelling accuracy

### AlphaFold timeline



# CASP15: multi-run AlphaFold-centered approaches dominate



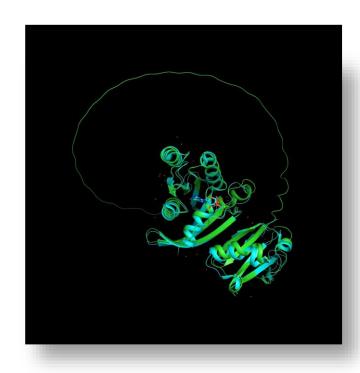
# CASP15: multi-run AlphaFold-centered approaches dominate

Winner of multimer competition

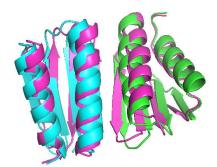
→ Afsample = <u>6000</u> predictions per target (+ select best)
<-> default AlphaFold: <u>5</u> (v₁) / <u>25</u> (v₂) per target

→ absence of evidence != evidence of absence

### AlphaFold-Multimer



Single chain with pseudolinker



2. Separate chains programmatically

Protein complex prediction with AlphaFold-Multimer

3. AlphaFold-Multimer



# Protein complex prediction with AlphaFold-Multimer

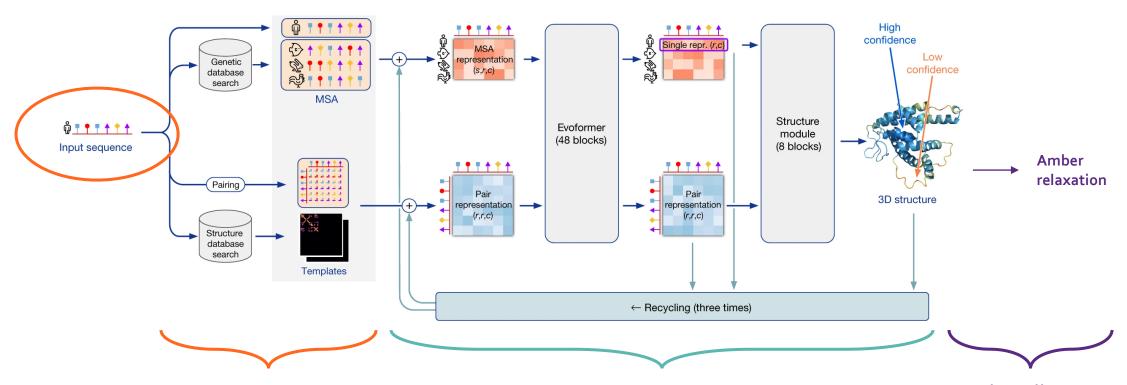
#### **Abstract**

While the vast majority of well-structured single protein chains can now be predicted to high accuracy due the recent AlphaFold [1] model, the prediction of multi-chain protein complexes remains a challenge in man cases. In this work, we demonstrate that an AlphaFold model trained specifically for multimeric inputs of kn stoichiometry, which we call AlphaFold-Multimer, significantly increases accuracy of predicted multimeric interfaces over input-adapted single-chain AlphaFold while maintaining high intra-chain accuracy. On a benchmark dataset of 17 heterodimer proteins without templates (introduced in [2]) we achieve at least

## The AlphaFold pipeline



## Stages of prediction



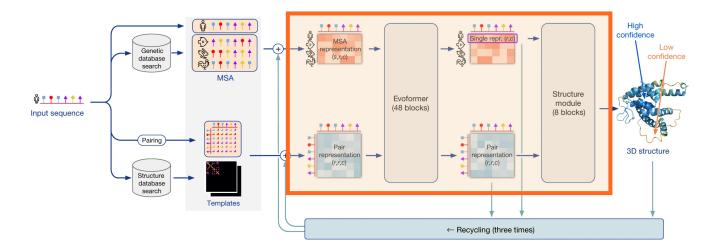
Multiple sequence alignment (CPU)
Structural templates search

Run inputs through 5 trained AlphaFold models (GPU)

- → 5 structures predicted
- → sorted on confidence

Relax all 5 structures (Amber)

## AlphaFold training



#### 5 prediction models

#### Trained **on**

- **PDB** protein fragments (256-384 residues)

#### Trained for

- Frame-aligned point error (FAPE)
  Minimize distance between predicted structure and ground truth
- + auxiliary losses (predict confidence by estimating when mistakes are made, physical constraints, ...)

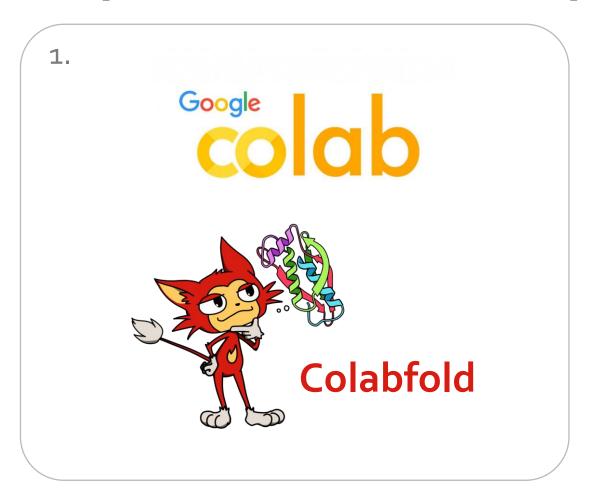
# Access to AlphaFold predictions



- Memory (RAM)
- GPUs
- Huge databases (~2.2 TB)



## AlphaFold availability



AlphaFold Protein Structure Database

Developed by DeepMind and EMBL-EBI



## **High-Performance Computing (HPC)**

Software and databases installed State-of-the-art GPU/CPU/RAM hardware Expert support Queuing system for allocating resources

#### AlphaFold on HPC

#### Advantages:

- Set up a large batches of experiments
- Faster + GPUs have greater memory → longer sequences
- No restricted availability (←→ Google Colab)



#### Available GPU servers (UGent):

| Available | joltik   | 10 nodes | 4x V100 GPU per node (32GB) | 256 GiB RAM p/n  | 800GB SSD p/n |
|-----------|----------|----------|-----------------------------|------------------|---------------|
| Available | accelgor | 9 nodes  | 4x A100 GPU per node (80GB) | ~500 GiB RAM p/n | 800GB SSD p/n |

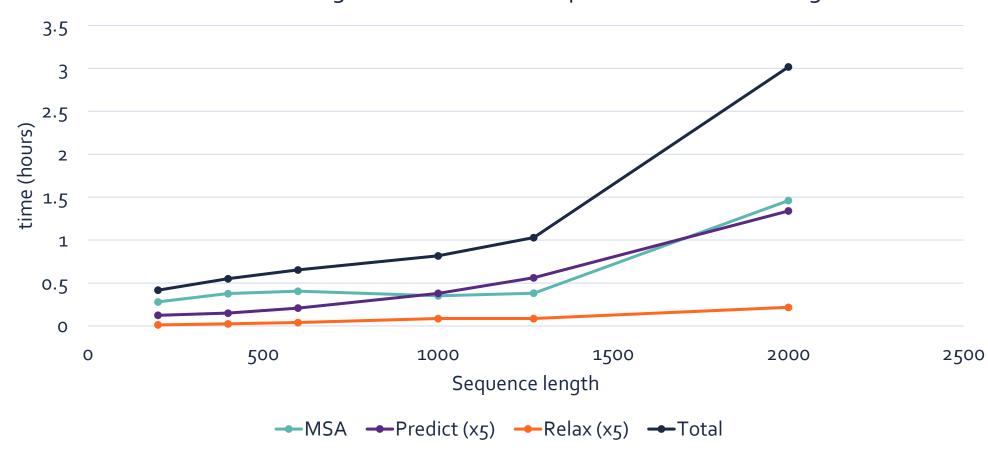
Available GPUs KUL: 20 x 4 P100 (16GB) + 2 x 4 V100 (32GB)

Available GPUs UA: 2 X 2 P100 (16GB)

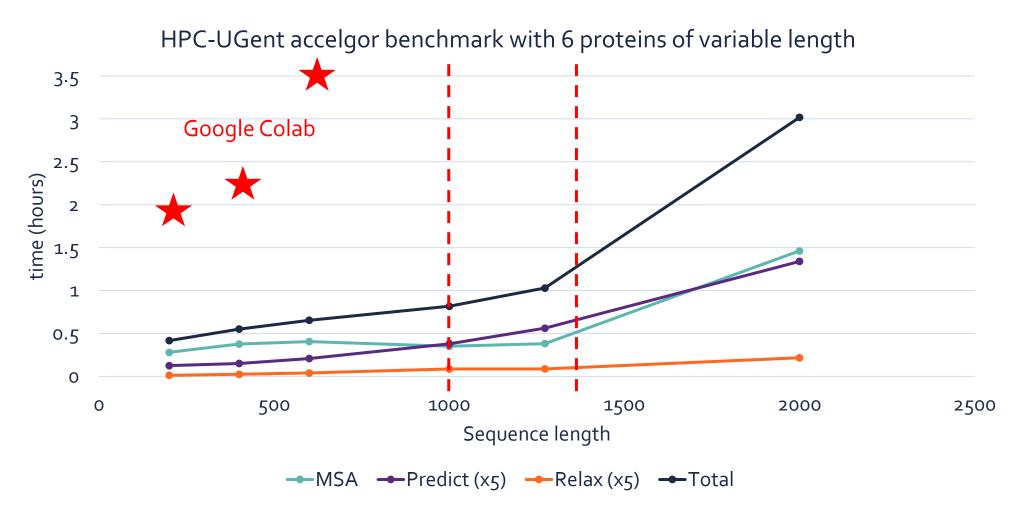
Available GPUs VUB: 6 x 2 K20Xm (6GB) + 4 x P100 (16GB) + 6 x A100 (40GB)

#### Computation time on the HPC (accelgor)

HPC-UGent accelgor benchmark with 6 proteins of variable length



### Online Google Colab limits

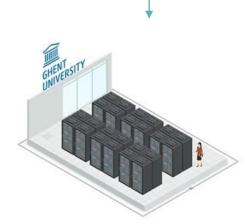


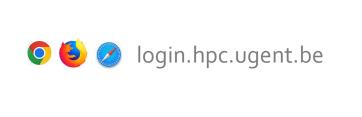
# Running AlphaFold on the HPC

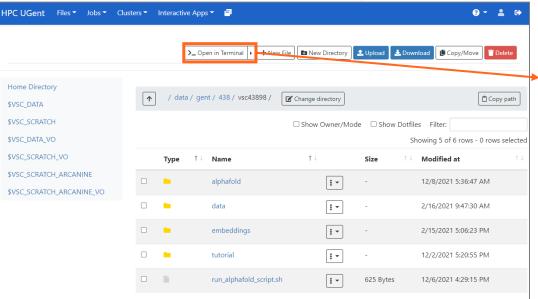


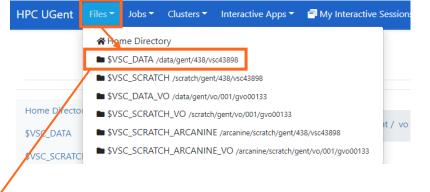
## Set-up: logging in











#### launch scripts



#### Command line commands

#### Introduction of needed commands (full command line workflow):

- <a href="https://elearning.bits.vib.be/courses/alphafold/lessons/alphafold-on-the-hpc/topic/basic-unix-commands/">https://elearning.bits.vib.be/courses/alphafold/lessons/alphafold-on-the-hpc/topic/basic-unix-commands/</a>

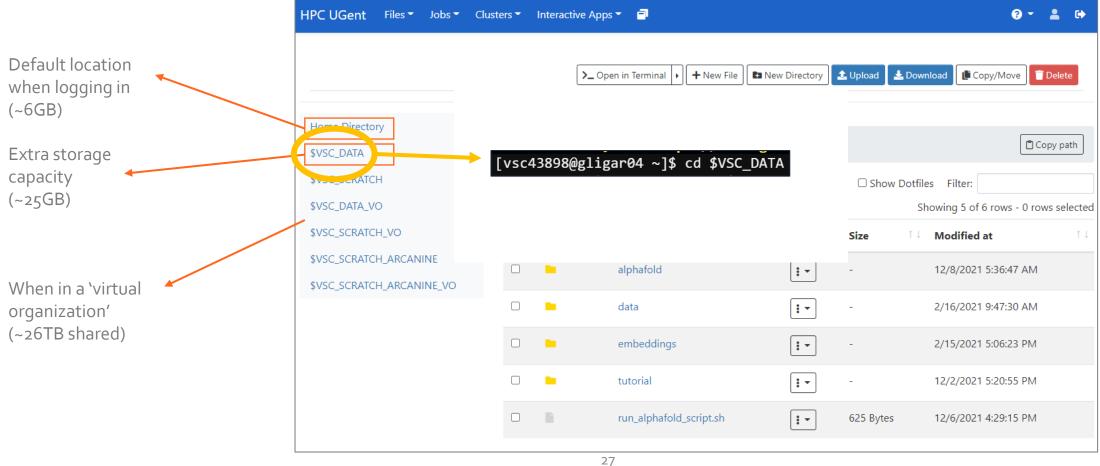
#### Needed commands for web interface workflow:

- pwd → print current directory
- $cd \rightarrow$  change directory
- Is  $\rightarrow$  list files in directory
- qsub/qdel/qstat → see later
- module → see later
- (mv/cp)

```
(base) [vsc43898@gligar05 alphafold]$ ls
some_directory some_text.txt
(base) [vsc43898@gligar05 alphafold]$ cp some_text.txt some_directory/
(base) [vsc43898@gligar05 alphafold]$ ls some_directory/
some_text.txt
(base) [vsc43898@gligar05 alphafold]$ mv some_text.txt new_name.txt
(base) [vsc43898@gligar05 alphafold]$ ls
new_name.txt some_directory
(base) [vsc43898@gligar05 alphafold]$ ls some_directory/
some_text.txt
(base) [vsc43898@gligar05 alphafold]$ mv new_name.txt some_directory/
(base) [vsc43898@gligar05 alphafold]$ ls
some_directory
(base) [vsc43898@gligar05 alphafold]$ ls some_directory/
new_name.txt some_text.txt
(base) [vsc43898@gligar05 alphafold]$
```

#### Set-up: directories and files

https://elearning.bits.vib.be/courses/alphafold/lessons/alphafold-on-the-hpc/topic/prepare-directories-and-files/



#### Set-up: directories and files

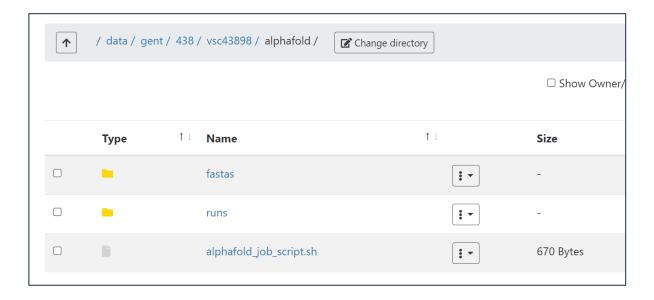
In \$DATA\_VSC: create directory alphafold

#### There:

- Create directory **fastas**
- Create directory runs
- Download <u>alphafold\_job\_script.sh</u>

OR copy contents to a file,
OR right-click on **Raw** > Save link as,
OR use *wget* with the raw link (**Raw** >
Copy link) on the command line directly





#### Set-up: directories and files

In fastas: create/upload your FASTA file

OR via command line (nano, vim, ...)
OR by uploading an existing file to the web interface
OR by creating + editing a new file via the web interface



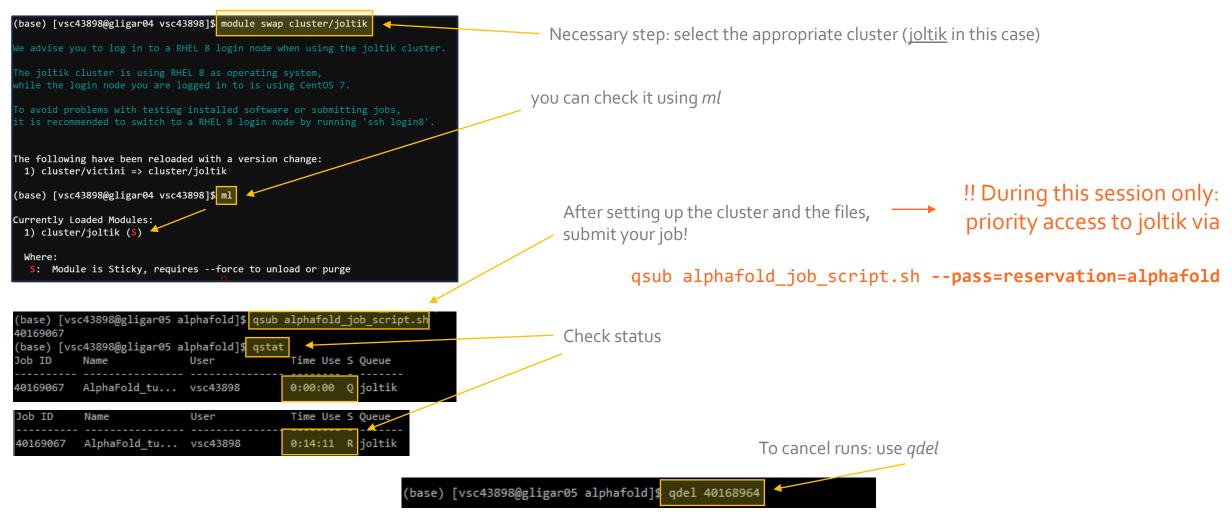
## Set-up: modify alphafold\_job\_script.sh

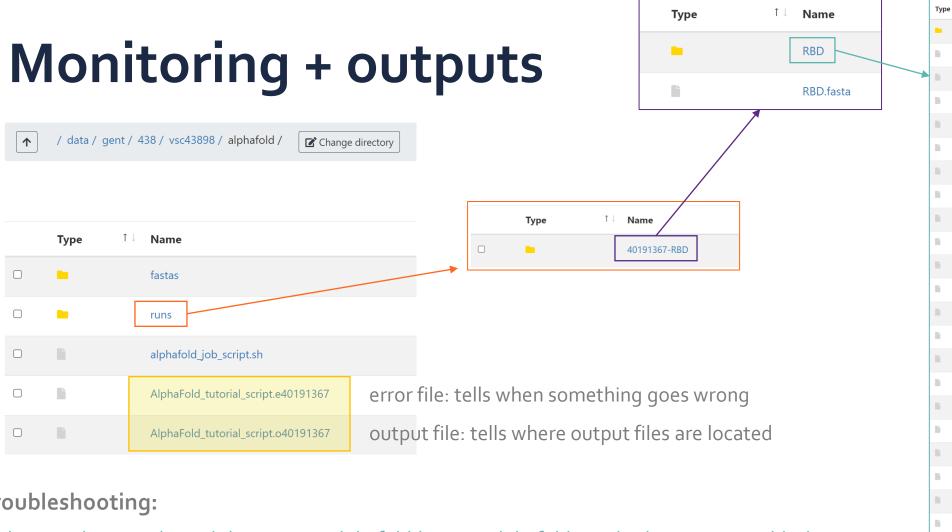
https://elearning.bits.vib.be/courses/alphafold/lessons/alphafold-on-the-hpc/topic/setting-up-the-job-script/

download link

```
Changes to make for individual experiments
#!/bin/bash
                                    Optional: change this name to anything for experiment monitoring
#PBS -N AlphaFold tutorial script
#PBS -l nodes=1:ppn=8:gpus=1
#PBS -l mem=64ab
#PBS -l walltime=24:0:0
                                                 Most important modification: the name of the FASTA
PROTEIN RBD
                                                 file (located in $VSC_DATA/alphafold/fastas/)
module load AlphaFold/2.1.1-fosscuda-2020b
export ALPHAFOLD DATA DIR=/arcanine/scratch/gent/apps/AlphaFold/20211201
                 /alphafold/runs/$PBS_JOBID-$PROTEIN
WORKDIR=
mkdir -p
                    R/fastas/SPROTEI
                                     N.fasta SWORKDIR/
echo Running
                      .fasta, output found at
alphafold --fasta_paths=
                                 .fasta
          --max template date=2020-05-14
          --db preset=full dbs
          --output dir=
           --model_preset=monomer_ptm | Change this to multimer if you want to run protein complexes
```

## Set-up: submitting and monitoring





#### **Troubleshooting:**

https://elearning.bits.vib.be/courses/alphafold/lessons/alphafold-on-the-hpc/topic/troubleshooting/

features.pkl ranked\_0.pdb

ranked\_1.pdb ranked\_2.pdb

ranked\_3.pdb ranked\_4.pdb ranking\_debug.json relaxed\_model\_1\_ptm.pdb

relaxed\_model\_2\_ptm.pdb

relaxed model 3 ptm.pdb

relaxed model 4 ptm.pdb

relaxed\_model\_5\_ptm.pdb

result\_model\_1\_ptm.pkl

result\_model\_2\_ptm.pkl result\_model\_3\_ptm.pkl

result\_model\_4\_ptm.pkl

result\_model\_5\_ptm.pkl

unrelaxed\_model\_1\_ptm.pdb

unrelaxed\_model\_2\_ptm.pdb unrelaxed\_model\_3\_ptm.pdb

unrelaxed model 4 ptm.pdb unrelaxed\_model\_5\_ptm.pdb

timings.json

## Exercise:

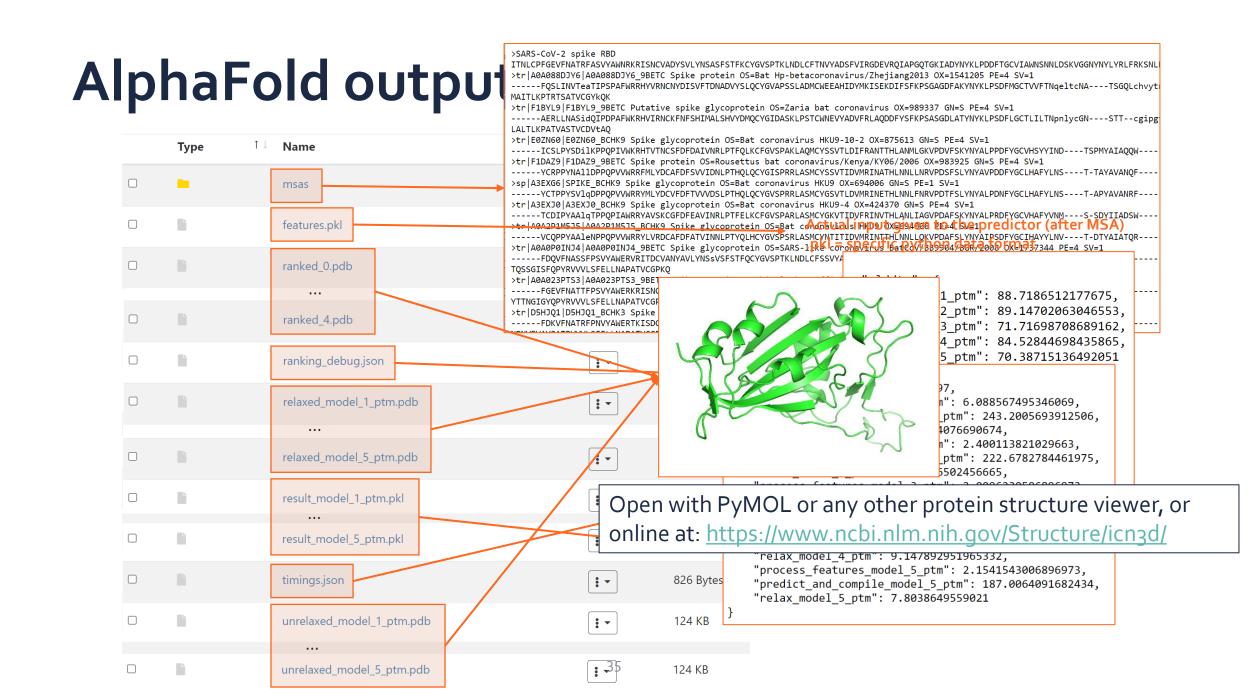
## run a prediction on the HPC

https://elearning.bits.vib.be/courses/alphafold/lessons/vib-training-session-alphafold/topic/first-experiment-on-the-hpc/



# AlphaFold outputs





## AlphaFold output folder

| Туре     | ↑↓ | Name                      | 1 ↓        | Size      |
|----------|----|---------------------------|------------|-----------|
| •        |    | msas                      | •          | -         |
|          |    | features.pkl              | • •        | 4.71 MB   |
|          |    | ranked_0.pdb              | • •        | 243 KB    |
|          |    | ranked_4.pdb              | • •        | 243 KB    |
|          |    | ranking_debug.json        | • •        | 369 Bytes |
| <b>L</b> |    | relaxed_model_1_ptm.pdb   |            | 243 KB    |
|          |    | relaxed_model_5_ptm.pdb   | • •        | 243 KB    |
|          |    | result_model_1_ptm.pkl    | <b>!</b> • | 28.3 MB   |
| lì .     |    | result_model_5_ptm.pkl    | • •        | 28.4 MB   |
|          |    | timings.json              | • •        | 826 Bytes |
|          |    | unrelaxed_model_1_ptm.pdb | • •        | 124 KB    |
|          |    | unrelaxed_model_5_ptm.pdb | : 36       | 124 KB    |

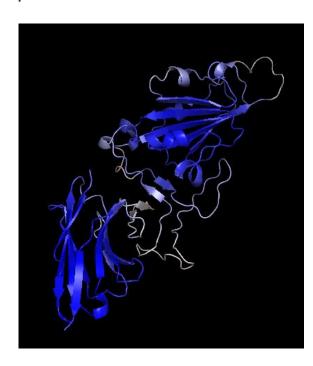
# Evaluating & interpreting AlphaFold predictions



## Metrics and scores to evaluate model accuracy

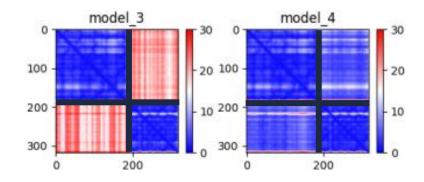
#### Local confidence

pLDDT (predicted local distance difference test)



#### Global confidence

PAE (predicted aligned error)



#### **Individual scores:**

- Average PLDDT
- **PTM** (predicted template modelling score)
- **iPTM** (interface PTM)

### pTM-score: intrinsic global accuracy estimate

- TM-score indicates the similarity between two structures by a score between o and 1, where 1 indicates a perfect match between two structures
- The TM-score is intended as a more accurate measure of the global similarity of full-length protein structures than the often used RMSD measure (independent of protein length, less sensitive to local dissimilarity).
- pTM score is the predicted TM-score between the structural prediction and the (unknown) true structure
- pTM for complexes == 0.2 \* PTM + 0.8 \* iPTM
- Alphafold predictions are ranked based on the pTM score
- References:

<u>TM-score</u>: Zhang Y and Skolnick J (2004). "Scoring function for automated assessment of protein structure template quality". *Proteins* <u>pTM score</u>: Jumper et al., (2021). "Highly accurate protein structure prediction with AlphaFold". *Nature* 

#### pLDDT reflects local model accuracy

- <u>pLDDT</u>: **p**redicted local **d**istance **d**ifference **t**est
- per residue confidence metric
- Estimates agreement of local environment between true structure and prediction
- structural models are coloured according to the pLDDTmetric (relative scale 1- 100):

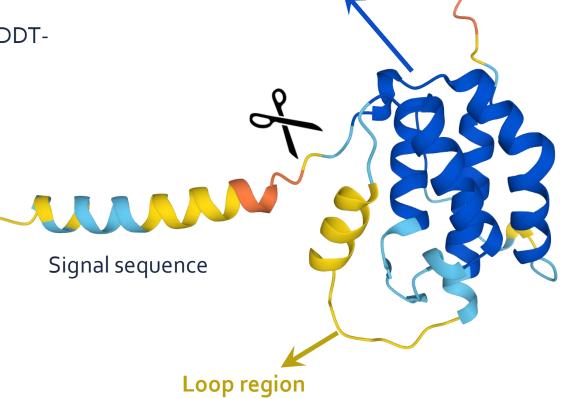
#### Model Confidence:

- Very high (pLDDT > 90)
- Confident (90 > pLDDT > 70)
- Low (70 > pLDDT > 50)
- Very low (pLDDT < 50)

AlphaFold produces a per-residue confidence score (pLDDT) between 0 and 100. Some regions below 50 pLDDT may be unstructured in isolation.

#### Structural model for Human TSLP

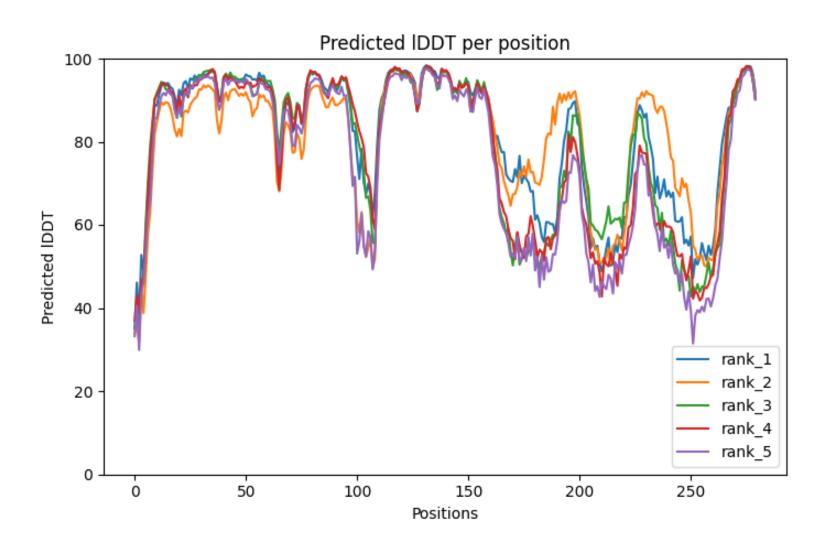
4-helix bundle region (high-confidence)



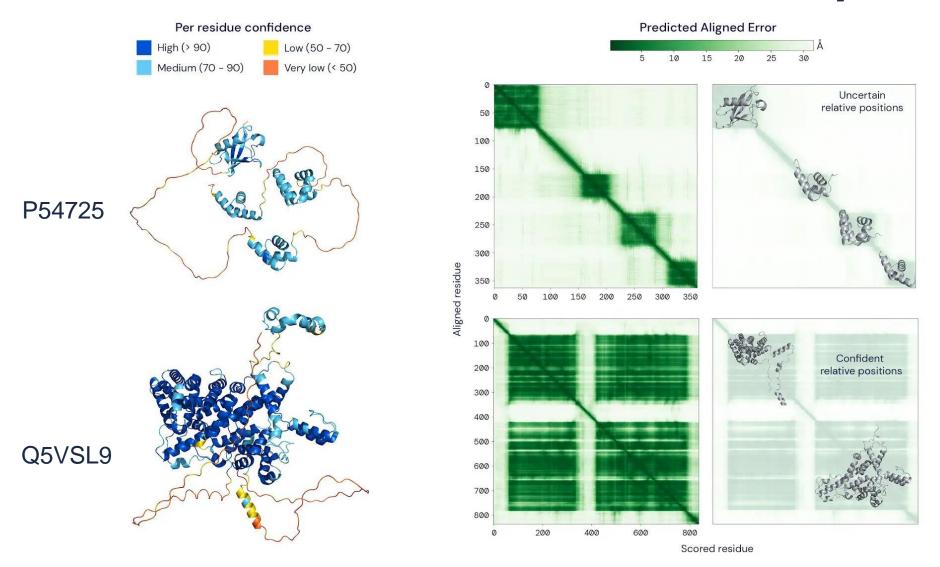
(low-confidence)

0

### pLDDT reflects local model accuracy



### PAE evaluates interdomain accuracy



### PAE-plot evaluates interdomain accuracy

Example:

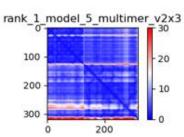
https://alphafold.ebi.ac.uk/entry/Q13049

(see Predicted aligned error tutorial (scroll down))

#### Example: VHH72 + SARS-CoV-1 (PDB:6WAQ)

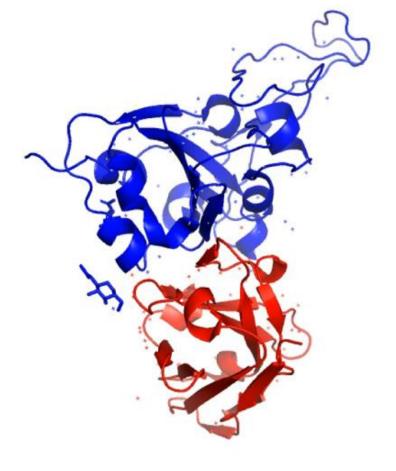


PTM=0.87



Green: PDB structure

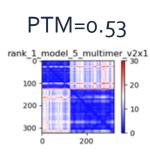
Cyan: predicted by AlphaFold

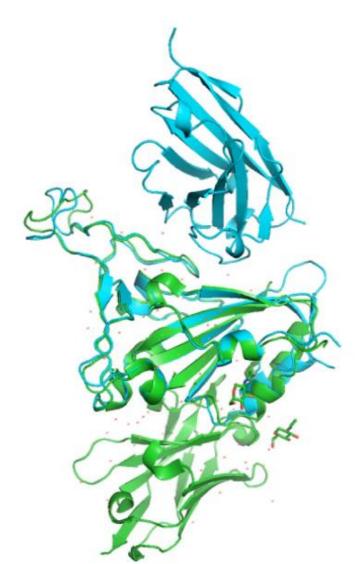


Blue: SARS-CoV-1 RBD

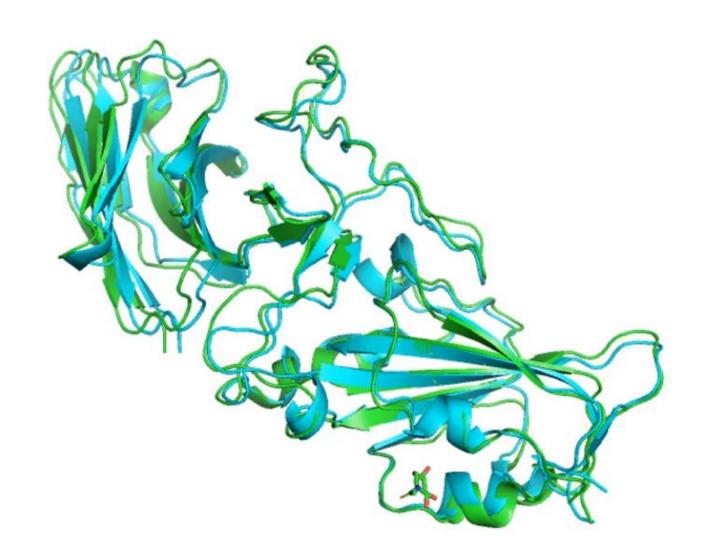
Red: VHH72

### Example: SARS-CoV-2 + VHH72





#### Example: SARS-CoV-2 + VHH E (PDB:7B14)



PTM=0.89

rank\_1\_model\_5\_multimer\_v2x
30
100
200
101
102

### Pitfalls of AlphaFold



### AlphaFold: uncertainties and potential pitfalls

- 1. Low accuracy of regions where **few sequences are available** for alignment
- 2. Low accuracy of intrinsically unstructured regions
- 3. Limited information on structural dynamics
- 4. Variable accuracy of multimeric assemblies
- 5. No information on co-factors/ligands/post-translational modifications
- 6. No post-translational processing, no compartmentalization of transmembrane assemblies

#### **Exercise:**

## Generate PAE/PLDDT using python script

!! Recent problem on joltik: if you have used "module load" after swapping to joltik, you need to open a new terminal session to run the python scripts in this exercise.

(otherwise you get the *Illegal instruction (core dumped)* error)



#### **SCIENCE MEETS LIFE**

#### Acknowledgements

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Kenneth Hoste













fwo