

class11

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Background

We saw last day that the main repositonary for biomolecular structure (the PDB database) only has about 250,000 entries.

UniProtKB (the main protien sequence database) has over 200 million entries!

In this hands-on session we will utilize AlphaFold to predict protein structure from sequence (Jumper et al. 2021).

Without the aid of such approaches, it can take years of expensive laboratory work to determine the structure of just one protein. With AlphaFold we can now accurately compute a typical protein structure in as little as ten minutes.

The EBI AlphaFold database

The EBI alphafold database contains lots of computed structure models. It is increasing likely that the structure you are intreased in is already in this database at <https://alphafold.ebi.ac.uk/>

There are 3 major outputs from AlphaFold

1. A model of structure in PDB format,
2. a pLDDT score: that tells us how confident the model is for a given residue in your protien (High values are good, above 70)
3. a **PAE score** that tells us about protien packing quality.

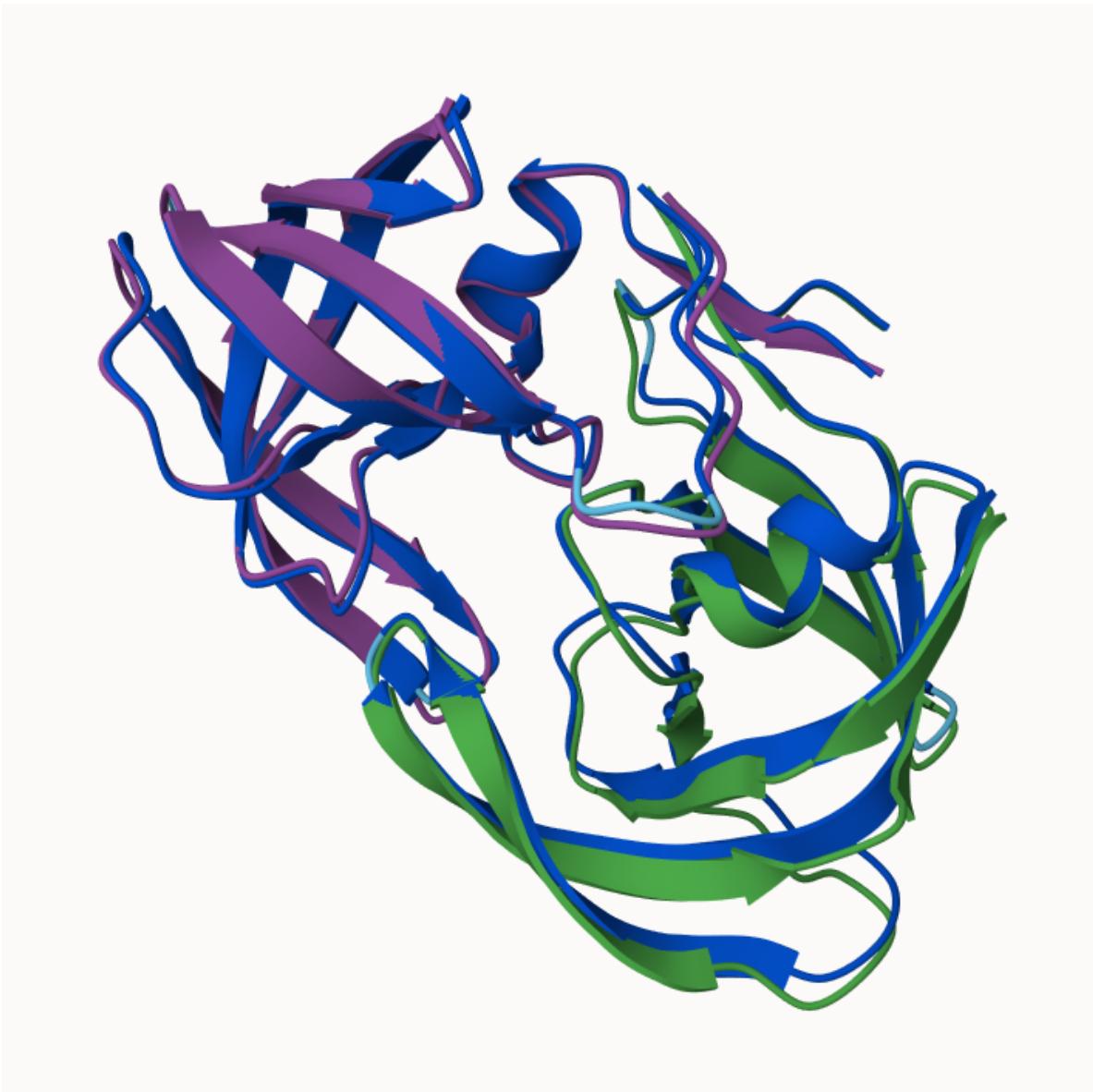
If you can't find a matching entry fir the sequence you are intreasted in AFDB you can run AlphaFold yourself...

Running AlphaFold

We will use Colab <https://github.com/sokrypton/ColabFold>

Figure from AlphaFold here!

```
knitr::include_graphics("Picture.png")
```



Interpreting results

coustom analysis of resulting models

we can read all AlphaFold results into R

```
results_dir <- "hivpr_23119_0"

pdb_files <- list.files(path = results_dir, pattern = "\\.pdb$",
                        full.names = TRUE)

basename(pdb_files)

[1] "hivpr_23119_0_unrelaxed_rank_001_alphaFold2_multimer_v3_model_4_seed_000.pdb"
[2] "hivpr_23119_0_unrelaxed_rank_002_alphaFold2_multimer_v3_model_1_seed_000.pdb"
[3] "hivpr_23119_0_unrelaxed_rank_003_alphaFold2_multimer_v3_model_5_seed_000.pdb"
[4] "hivpr_23119_0_unrelaxed_rank_004_alphaFold2_multimer_v3_model_2_seed_000.pdb"
[5] "hivpr_23119_0_unrelaxed_rank_005_alphaFold2_multimer_v3_model_3_seed_000.pdb"

library(bio3d)
pdbs <- pdbaln(pdb_files, fit = TRUE, exefile = "msa")

Reading PDB files:
hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_001_alphaFold2_multimer_v3_model_4_seed_000.pdb
hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_002_alphaFold2_multimer_v3_model_1_seed_000.pdb
hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_003_alphaFold2_multimer_v3_model_5_seed_000.pdb
hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_004_alphaFold2_multimer_v3_model_2_seed_000.pdb
hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_005_alphaFold2_multimer_v3_model_3_seed_000.pdb
.....
Extracting sequences

pdb/seq: 1  name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_001_alphaFold2_multimer_v3_model_4_seed_000
pdb/seq: 2  name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_002_alphaFold2_multimer_v3_model_1_seed_000
pdb/seq: 3  name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_003_alphaFold2_multimer_v3_model_5_seed_000
pdb/seq: 4  name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_004_alphaFold2_multimer_v3_model_2_seed_000
pdb/seq: 5  name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_005_alphaFold2_multimer_v3_model_3_seed_000

pdbs
```

	1	50
[Truncated_Name:1]hivpr_2311	PQITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:2]hivpr_2311	PQITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:3]hivpr_2311	PQITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:4]hivpr_2311	PQITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:5]hivpr_2311	PQITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	

	1	50
	51	100
[Truncated_Name:1]hivpr_2311	GGFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP	
[Truncated_Name:2]hivpr_2311	GGFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP	
[Truncated_Name:3]hivpr_2311	GGFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP	
[Truncated_Name:4]hivpr_2311	GGFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP	
[Truncated_Name:5]hivpr_2311	GGFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP	

	51	100
	101	150
[Truncated_Name:1]hivpr_2311	QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:2]hivpr_2311	QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:3]hivpr_2311	QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:4]hivpr_2311	QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	
[Truncated_Name:5]hivpr_2311	QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGI	

	101	150
	151	198
[Truncated_Name:1]hivpr_2311	GFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNF	
[Truncated_Name:2]hivpr_2311	GFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNF	
[Truncated_Name:3]hivpr_2311	GFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNF	
[Truncated_Name:4]hivpr_2311	GFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNF	
[Truncated_Name:5]hivpr_2311	GFIKVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNF	

	151	198

Call:

```
pdbaln(files = pdb_files, fit = TRUE, exefile = "msa")
```

Class:

```
pdb, fasta
```

Alignment dimensions:

```
5 sequence rows; 198 position columns (198 non-gap, 0 gap)  
+ attr: xyz, resno, b, chain, id, ali, resid, sse, call  
  
#pdbs
```

Similarity and differences between the models

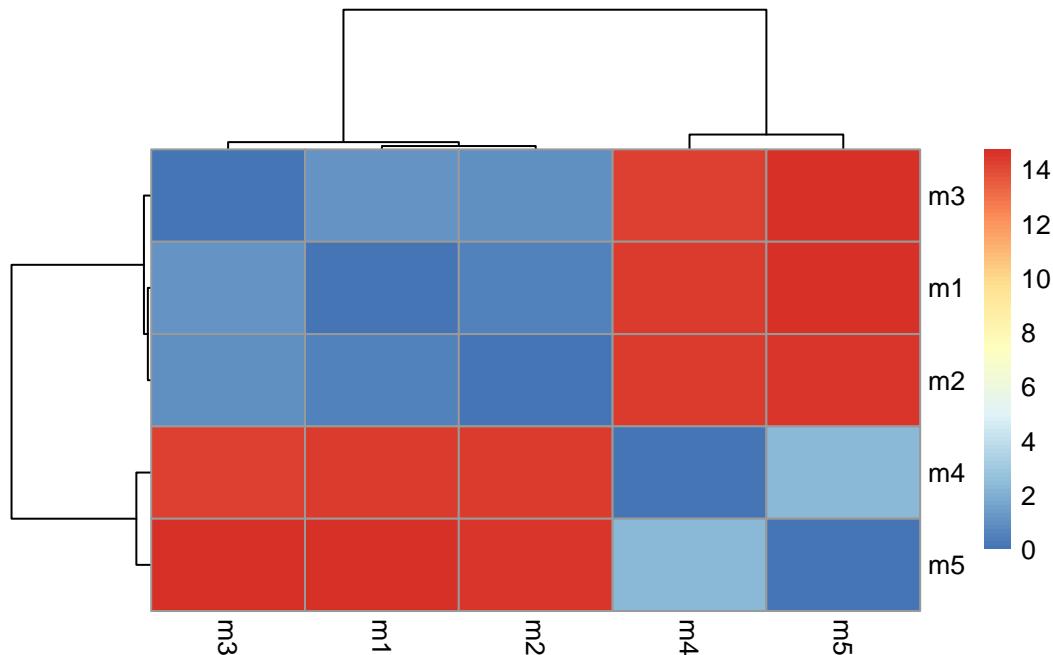
```
rd <- rmsd(pdbs, fit=TRUE)
```

Warning in rmsd(pdbs, fit = TRUE): No indices provided, using the 198 non NA positions

```
range(rd)
```

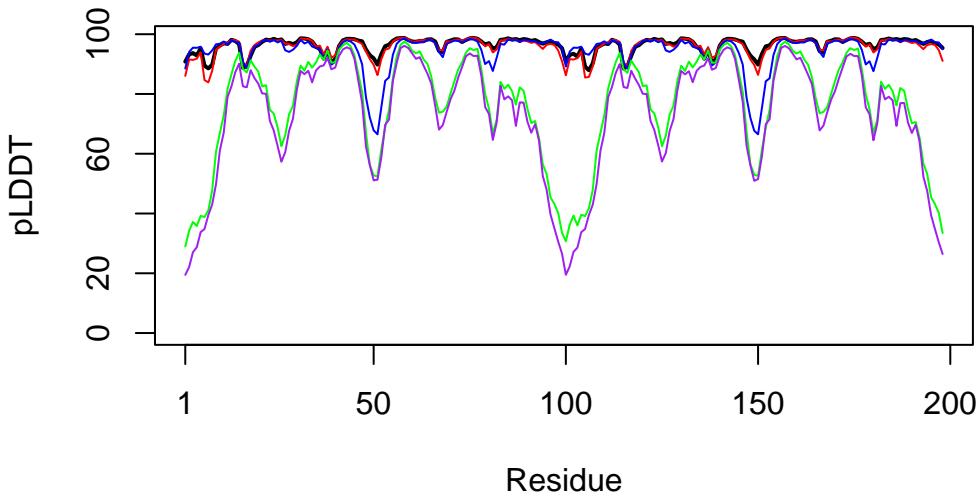
```
[1] 0.000 14.754
```

```
library(pheatmap)  
  
colnames(rd) <- paste0("m",1:5)  
rownames(rd) <- paste0("m",1:5)  
  
pheatmap(rd)
```



Models with lower RMSD values are more structurally similar. Clustering in the heatmap indicates which predictions agree most strongly.

```
plotb3(pdbs$b[1,], typ="l", lwd=2,  
       ylab="pLDDT", xlab="Residue")  
  
points(pdbs$b[2,], col="red", typ="l")  
points(pdbs$b[3,], col="blue", typ="l")  
points(pdbs$b[4,], col="green", typ="l")  
points(pdbs$b[5,], col="purple", typ="l")
```



Most show high pLDDT scores so they are reliable predictions. The Lower scores suggest flexible or disordered regions.

```
library(jsonlite)  
  
pae_files <- list.files(path=results_dir,  
                         pattern=".*model.*\\.json$",
                         full.names=TRUE)  
  
pae_files
```

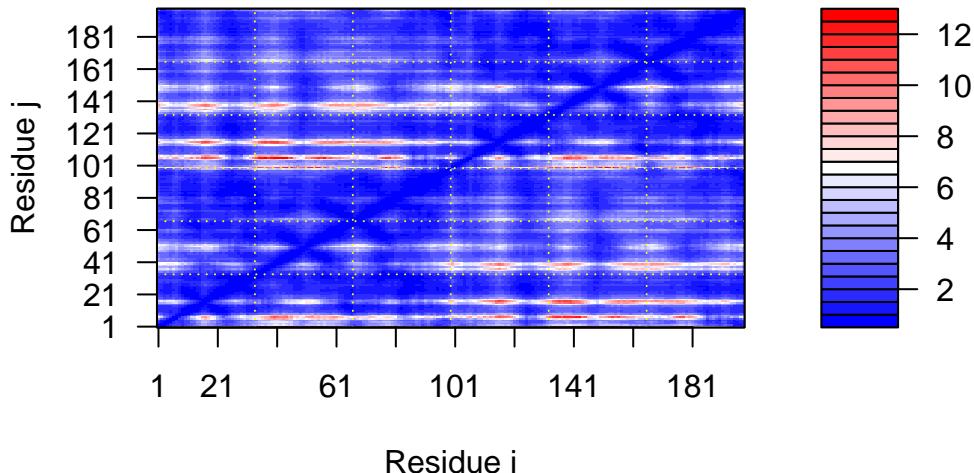
```
[1] "hivpr_23119_0/hivpr_23119_0_scores_rank_001_alphaFold2_multimer_v3_model_4_seed_000.json
```

```
[2] "hivpr_23119_0/hivpr_23119_0_scores_rank_002_alphaFold2_multimer_v3_model_1_seed_000.json"
[3] "hivpr_23119_0/hivpr_23119_0_scores_rank_003_alphaFold2_multimer_v3_model_5_seed_000.json"
[4] "hivpr_23119_0/hivpr_23119_0_scores_rank_004_alphaFold2_multimer_v3_model_2_seed_000.json"
[5] "hivpr_23119_0/hivpr_23119_0_scores_rank_005_alphaFold2_multimer_v3_model_3_seed_000.json"
```

```
pae1 <- read_json(pae_files[1], simplifyVector=TRUE)
str(pae1)
```

```
List of 5
$ plddt : num [1:198] 90.8 93.2 93.7 92.9 95.2 ...
$ max_pae: num 12.8
$ pae     : num [1:198, 1:198] 0.75 0.86 1.39 1.78 2.34 4.23 7.09 5.05 3.2 2.37 ...
$ ptm    : num 0.91
$ iptm   : num 0.9
```

```
plot.dmat(pae1$pae,
           xlab="Residue i",
           ylab="Residue j")
```



The PAE plot is mostly low (blue), indicating high confidence in residue positioning, with a few higher-error regions showing minor uncertainty in domain packing but overall a reliable model.