

class11

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Background

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

UniProtKB (the main protein 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 interested 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 protein (High values are good, above 70)
3. a **PAE score** that tells us about protein packing quality.

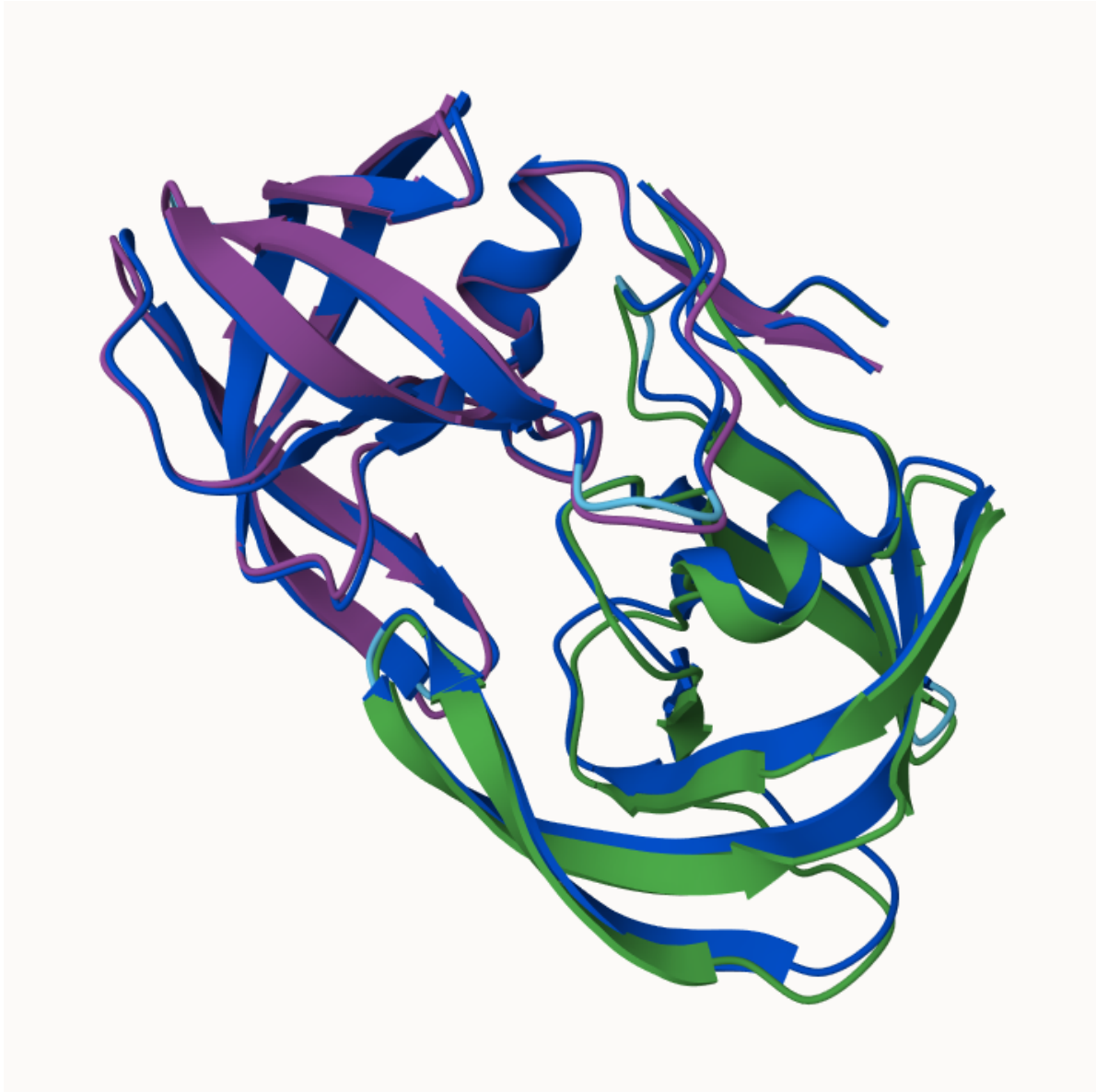
If you can't find a matching entry for the sequence you are interested in AFDB you can run AlphaFold yourself...

Running AlphaFold

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

Figure from AlphaFold here!

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



Interpreting results

custom 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
pdb/seq: 2   name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_002_alphafold2_multimer_v3_model_1_seed_000.pdb
pdb/seq: 3   name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_003_alphafold2_multimer_v3_model_5_seed_000.pdb
pdb/seq: 4   name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_004_alphafold2_multimer_v3_model_2_seed_000.pdb
pdb/seq: 5   name: hivpr_23119_0/hivpr_23119_0_unrelaxed_rank_005_alphafold2_multimer_v3_model_3_seed_000.pdb
```

```
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 GGFIVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP
[Truncated_Name:2]hivpr_2311 GGFIVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP
[Truncated_Name:3]hivpr_2311 GGFIVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP
[Truncated_Name:4]hivpr_2311 GGFIVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP
[Truncated_Name:5]hivpr_2311 GGFIVRQYDQILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFP
*****
51 . . . . 100

101 . . . . 150
[Truncated_Name:1]hivpr_2311 QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGIG
[Truncated_Name:2]hivpr_2311 QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGIG
[Truncated_Name:3]hivpr_2311 QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGIG
[Truncated_Name:4]hivpr_2311 QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGIG
[Truncated_Name:5]hivpr_2311 QITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKPMIGGIG
*****
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:

```
pdbs, fasta
```

Alignment dimensions:

```

5 sequence rows; 198 position columns (198 non-gap, 0 gap)

+ attr: xyz, resno, b, chain, id, ali, resid, sse, call

#pdb

```

Similarity and differences between the models

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

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

```
range(rd)
```

```
[1] 0.000 14.754
```

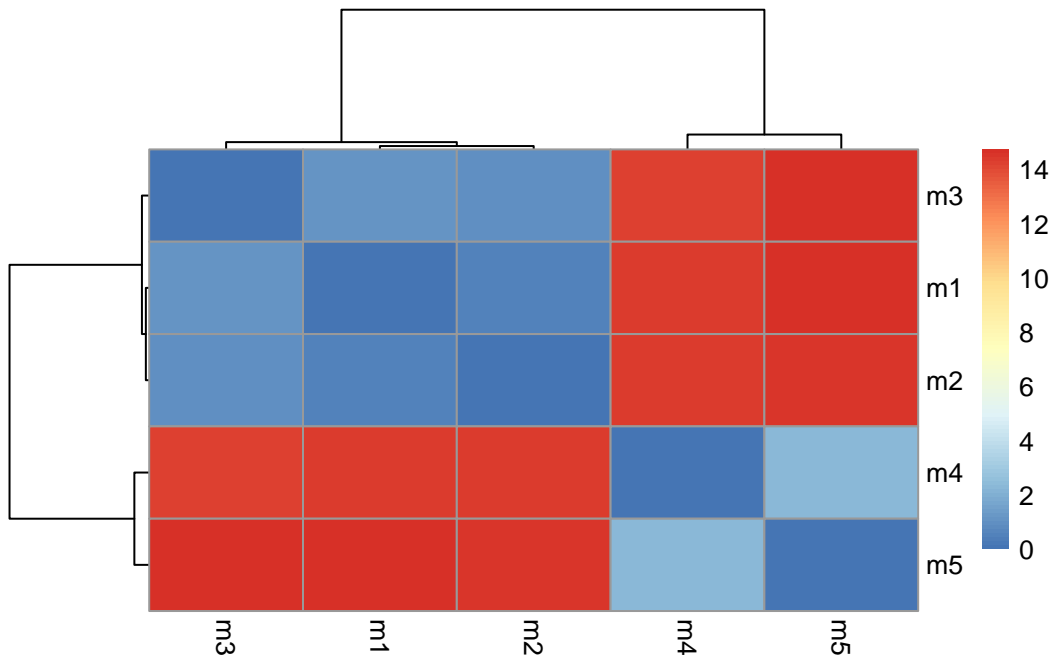
```

library(pheatmap)

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

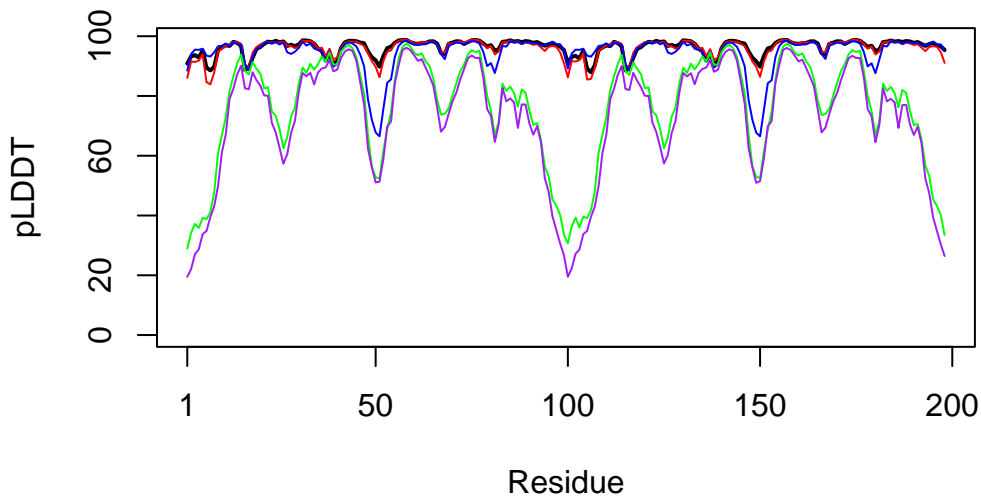
pheatmap(rd)

```



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

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

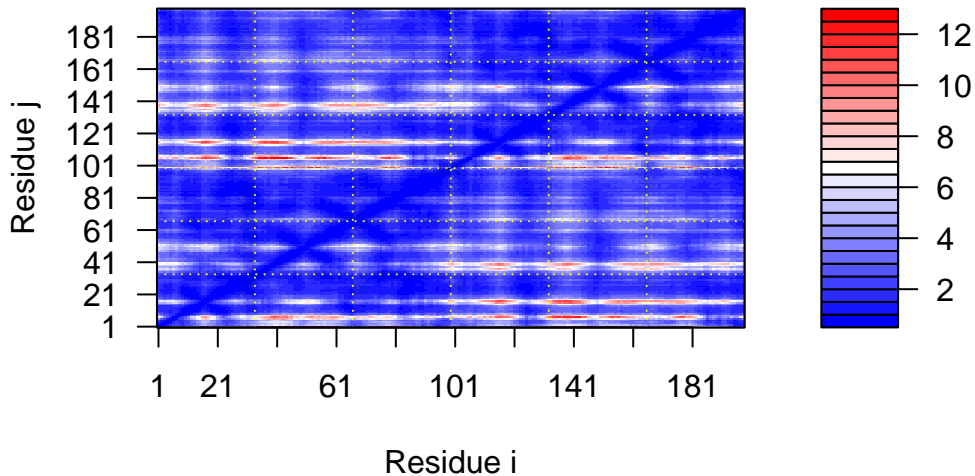
```
pae1 <- read_json(pae_files[1], simplifyVector=TRUE)

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.