Class_9_structural_bioinformatics

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10-26-2022

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| Getting started viewing pdb files | |

You can use multiple viewing tools such as Chimerax or the free online website molstart https://molstar.org/viewer/

Q1: What percentage of structures in the PDB are solved by X-Ray and Electron Microscopy.

```
total_xray <- 150417/171351
total_xray
```

[1] 0.8778297

```
total_em <- 8586/171351
total_em
```

```
[1] 0.05010767
```

Q2: What proportion of structures in the PDB are protein?

```
171351/(171351+10459+10919+4037+191+22)
```

[1] 0.8698948

Q3: Type HIV in the PDB website search box on the home page and determine how many HIV-1 protease structures are in the current PDB?

4707

bio3d and its function in loading PDB files

To load and interact with pdb files into R we need a package bio3d

```
library(bio3d)

example_pdb <-read.pdb("1hsg")

Note: Accessing on-line PDB file

example_pdb

Call: read.pdb(file = "1hsg")

Total Models#: 1
   Total Atoms#: 1686, XYZs#: 5058 Chains#: 2 (values: A B)

Protein Atoms#: 1514 (residues/Calpha atoms#: 198)
   Nucleic acid Atoms#: 0 (residues/phosphate atoms#: 0)

Non-protein/nucleic Atoms#: 172 (residues: 128)
   Non-protein/nucleic resid values: [ HOH (127), MK1 (1) ]

Protein sequence:
   PQITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWKPKMIGGIGGFIKVRQYD</pre>
```

QILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFPQITLWQRPLVTIKIGGQLKE ALLDTGADDTVLEEMSLPGRWKPKMIGGIGGFIKVRQYDQILIEICGHKAIGTVLVGPTP VNIIGRNLLTQIGCTLNF

+ attr: atom, xyz, seqres, helix, sheet,

```
calpha, remark, call
  head(example_pdb$atom)
  type eleno elety alt resid chain resno insert
                                                                z o
1 ATOM
          1
               N < NA >
                         PRO
                                A 1
                                          <NA> 29.361 39.686 5.862 1 38.10
                                Α
2 ATOM
          2
               CA <NA>
                         PRO
                                     1 <NA> 30.307 38.663 5.319 1 40.62
3 ATOM
              C <NA>
                         PRO
                                     1 <NA> 29.760 38.071 4.022 1 42.64
4 ATOM
          4
                         PRO
                                     1 <NA> 28.600 38.302 3.676 1 43.40
               O <NA>
                                Α
                                Α
                                      1 <NA> 30.508 37.541 6.342 1 37.87
5 ATOM
          5
               CB <NA>
                         PRO
6 ATOM
          6
               CG <NA>
                         PRO
                                      1 <NA> 29.296 37.591 7.162 1 38.40
  segid elesy charge
1 <NA>
           N
               <NA>
2 <NA>
           C <NA>
3 <NA>
           C <NA>
4 <NA>
           O <NA>
5 <NA>
           C <NA>
6 <NA>
               <NA>
Looking at other proteins: Adenylate Kinase
  adk <- read.pdb("6s36")
  Note: Accessing on-line PDB file
  PDB has ALT records, taking A only, rm.alt=TRUE
  adk
 Call: read.pdb(file = "6s36")
  Total Models#: 1
    Total Atoms#: 1898, XYZs#: 5694 Chains#: 1 (values: A)
    Protein Atoms#: 1654 (residues/Calpha atoms#: 214)
```

```
Nucleic acid Atoms#: 0 (residues/phosphate atoms#: 0)

Non-protein/nucleic Atoms#: 244 (residues: 244)

Non-protein/nucleic resid values: [ CL (3), HOH (238), MG (2), NA (1) ]

Protein sequence:

MRIILLGAPGAGKGTQAQFIMEKYGIPQISTGDMLRAAVKSGSELGKQAKDIMDAGKLVT

DELVIALVKERIAQEDCRNGFLLDGFPRTIPQADAMKEAGINVDYVLEFDVPDELIVDKI

VGRRVHAPSGRVYHVKFNPPKVEGKDDVTGEELTTRKDDQEETVRKRLVEYHQMTAPLIG

YYSKEAEAGNTKYAKVDGTKPVAEVRADLEKILG
```

```
+ attr: atom, xyz, seqres, helix, sheet, calpha, remark, call
```

We can use important information here such as the "(residues/Calpha atoms#: 214)" which tells us there are 214 amino acids.

We can look at all other attributes.

```
attributes(adk)

$names
[1] "atom" "xyz" "seqres" "helix" "sheet" "calpha" "remark" "call"

$class
[1] "pdb" "sse"
```

We can analyze the functional motions of structure

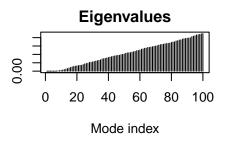
Predicting functional motions of a single structure

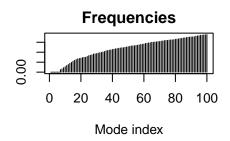
The first thing we will do is plot the normal mode analysis. This is a structural bioinformatics method to determine and predict the protein flexibility and potential functional motions.

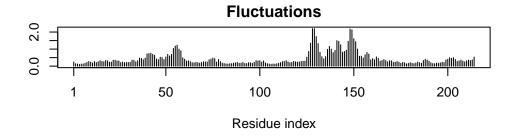
```
# Perform flexiblity prediction
m <- nma(adk)

Building Hessian... Done in 0.083 seconds.
Diagonalizing Hessian... Done in 0.257 seconds.</pre>
```

plot(m)







Now we want to use this to make a movie and plot the trajectory of certain residues

This function produces a file that has the multiple states and can be opened in molstar to view a movie of the enzyme different conformation changes as calculated by the normal mode analysis (which is performed via a force field calculation modeling atomic bonds as springs)

Now what we want to do is obtain all homolog of this protein and see what their conformation changes are like, and which ones are active and which ones are inactive, etc.

Using R to create a repository of homologs in order to do a comparative analysis across structures

We first need to load in all the predicted similar protein structures:

First we want to get our initial protein of interest, in this case it is ADK.

We will then get.seq() on this protein, then use this sequence to search a protein database (PDB database) with the function blast.pdb(), then we will get a list of all homologs and get their .pdb files with get.pdb()

```
#aminoacidseq <- get.seq("1ake_A")
#blast_results <- blast.pdb(aminoacidseq) #commented out but usually you would run this
#we can plot a summary plot
#hits <- plot(blast_results)
#this graph has the Y axis in units of -log(Evalue)

hits <- NULL
hits$pdb.id <- c('1AKE_A','6S36_A','6RZE_A','3HPR_A','1E4V_A','5EJE_A','1E4Y_A','3X2S_A','</pre>
```

We can see that there are some Nhits = # where there are some amount that are good enough based on their E value

```
hits$pdb.id

[1] "1AKE_A" "6S36_A" "6RZE_A" "3HPR_A" "1E4V_A" "5EJE_A" "1E4Y_A" "3X2S_A"

[9] "6HAP_A" "6HAM_A" "4K46_A" "3GMT_A" "4PZL_A"

#pdb.annotate(hits$pdb.id)
```

Now we want to obtain our zipped/folder package of our pdb files.

```
files <- get.pdb(hits$pdb.id, path="pdbs", split=TRUE,gzip=TRUE)

Warning in get.pdb(hits$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/
1AKE.pdb.gz exists. Skipping download

Warning in get.pdb(hits$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/
6S36.pdb.gz exists. Skipping download

Warning in get.pdb(hits$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/
6RZE.pdb.gz exists. Skipping download

Warning in get.pdb(hits$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/
3HPR.pdb.gz exists. Skipping download
```

Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 1E4V.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 5EJE.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 1E4Y.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 3X2S.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 6HAP.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 6HAM.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 4K46.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 3GMT.pdb.gz exists. Skipping download Warning in get.pdb(hits\$pdb.id, path = "pdbs", split = TRUE, gzip = TRUE): pdbs/ 4PZL.pdb.gz exists. Skipping download 0% 8% 15% 23% |========== | 31% 38%

| | | 46% |
|--|---|------|
| ======= | | 54% |
| ======= | | 62% |
| =================================== | | 69% |
| =================================== | | 77% |
| =================================== | | 85% |
| =================================== | | 92% |
| =================================== | 1 | 100% |

Now we want to overlap all of them:

. . .

```
pdbs <- pdbaln(files, fit = TRUE, exefile="msa")</pre>
```

```
Reading PDB files:
pdbs/split_chain/1AKE_A.pdb
pdbs/split_chain/6S36_A.pdb
pdbs/split_chain/6RZE_A.pdb
pdbs/split_chain/3HPR_A.pdb
pdbs/split_chain/1E4V_A.pdb
pdbs/split_chain/5EJE_A.pdb
pdbs/split_chain/1E4Y_A.pdb
pdbs/split_chain/3X2S_A.pdb
pdbs/split_chain/6HAP_A.pdb
pdbs/split_chain/6HAM_A.pdb
pdbs/split_chain/4K46_A.pdb
pdbs/split_chain/3GMT_A.pdb
pdbs/split_chain/4PZL_A.pdb
   PDB has ALT records, taking A only, rm.alt=TRUE
   PDB has ALT records, taking A only, rm.alt=TRUE
   PDB has ALT records, taking A only, rm.alt=TRUE
    PDB has ALT records, taking A only, rm.alt=TRUE
     PDB has ALT records, taking A only, rm.alt=TRUE
       PDB has ALT records, taking A only, rm.alt=TRUE
    PDB has ALT records, taking A only, rm.alt=TRUE
```

8

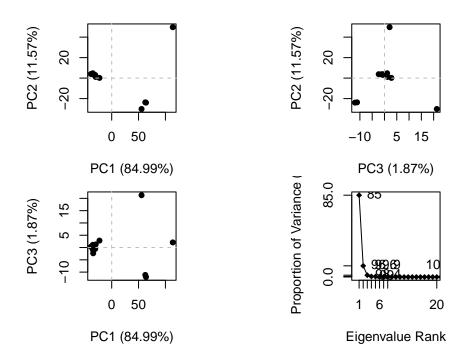
Extracting sequences

```
name: pdbs/split_chain/1AKE_A.pdb
   PDB has ALT records, taking A only, rm.alt=TRUE
             name: pdbs/split_chain/6S36_A.pdb
pdb/seq: 2
   PDB has ALT records, taking A only, rm.alt=TRUE
pdb/seq: 3
             name: pdbs/split_chain/6RZE_A.pdb
   PDB has ALT records, taking A only, rm.alt=TRUE
pdb/seq: 4
             name: pdbs/split_chain/3HPR_A.pdb
   PDB has ALT records, taking A only, rm.alt=TRUE
pdb/seq: 5
             name: pdbs/split_chain/1E4V_A.pdb
pdb/seq: 6
             name: pdbs/split_chain/5EJE_A.pdb
   PDB has ALT records, taking A only, rm.alt=TRUE
pdb/seq: 7
             name: pdbs/split_chain/1E4Y_A.pdb
pdb/seq: 8
             name: pdbs/split_chain/3X2S_A.pdb
pdb/seq: 9
             name: pdbs/split_chain/6HAP_A.pdb
              name: pdbs/split_chain/6HAM_A.pdb
pdb/seq: 10
   PDB has ALT records, taking A only, rm.alt=TRUE
              name: pdbs/split_chain/4K46_A.pdb
pdb/seq: 11
   PDB has ALT records, taking A only, rm.alt=TRUE
pdb/seq: 12
              name: pdbs/split_chain/3GMT_A.pdb
pdb/seq: 13
              name: pdbs/split_chain/4PZL_A.pdb
  #pdbs
  ids <- basename.pdb(pdbs$id)</pre>
  #plot(pdbs, labels=ids)
(sequencealignment.png)
```

Then we want to start our comparative analysis with a principal comparative analysis

Principal component analysis

```
pc.xray <- pca(pdbs)
plot(pc.xray)</pre>
```

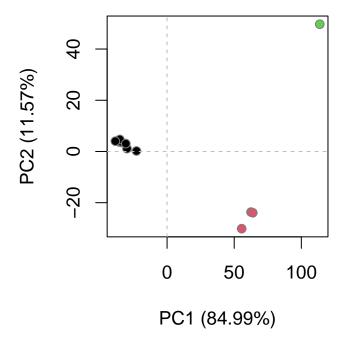


You can do calculate RMSD to do structure-based clustering

```
rd <- rmsd(pdbs)
```

Warning in rmsd(pdbs): No indices provided, using the 204 non NA positions

```
hc.rd <- hclust(dist(rd))
grps.rd <- cutree(hc.rd, k=3)
plot(pc.xray, 1:2, col="grey50", bg=grps.rd, pch=21, cex=1)</pre>
```

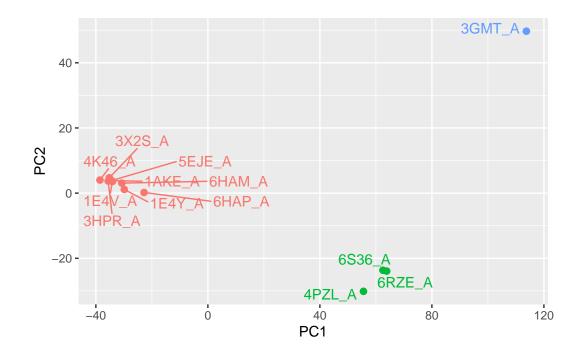


Here we see there are clearly two different groups

When we plot this and when we perform mktrj we are essentially creating dots between the two different structural groups (marked here in red and black)

When we compare inhibitor some are more inhibited than others, is there any way that this plot determines this? Short answer; no, not in this initial plotting at least. It is possible that if you were to characterize the inhibition of certain structures and make that another variable mask you could theoretically perform a PCA to determine a separating function that figures out how you could develop and make a trajectory between low inhibition and more inhibition

```
p <- ggplot(df) +
  aes(PC1, PC2, col=col, label=ids) +
  geom_point(size=2) +
  geom_text_repel(max.overlaps = 20) +
  theme(legend.position = "none")
p</pre>
```



modes <- nma(pdbs)</pre>

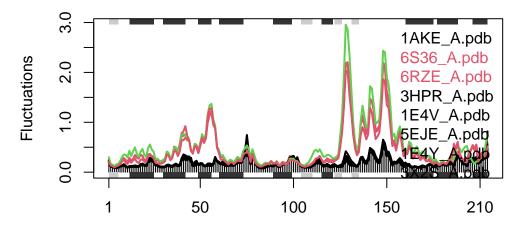
```
Details of Scheduled Calculation:
```

- ... 13 input structures
- \dots storing 606 eigenvectors for each structure
- ... dimension of x\$U.subspace: (612x606x13)
- \dots coordinate superposition prior to NM calculation
- ... aligned eigenvectors (gap containing positions removed)
- ... estimated memory usage of final 'eNMA' object: 36.9 Mb

0%

plot(modes, pdbs, col=grps.rd)

Extracting SSE from pdbs\$sse attribute



Residue number (reference PDB: 1AKE_A)