

# Class 09: Structural Bioinformatics pt. 1

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## The PDB Database

The main repository for biomolecular structure data is the Protein Data Bank (PDB) <https://www.rcsb.org>.

Let's have a quick look at the composition of this database:

```
pdb_stats <- read.csv("Data Export Summary.csv", row.names=1)
tail(pdb_stats)
```

	X.ray	EM	NMR	Integrative	Multiple.methods
Protein (only)	176,378	20,438	12,709	342	221
Protein/Oligosaccharide	10,284	3,396	34	8	11
Protein/NA	9,007	5,931	287	24	7
Nucleic acid (only)	3,077	200	1,554	2	15
Other	174	13	33	3	0
Oligosaccharide (only)	11	0	6	0	1
	Neutron	Other	Total		
Protein (only)	83	32	210,203		
Protein/Oligosaccharide	1	0	13,734		
Protein/NA	0	0	15,256		
Nucleic acid (only)	3	1	4,852		

Other	0	0	223
Oligosaccharide (only)	0	4	22

I need to convert to numeric

```
# This was my strategy but it sounds like Barry has a different approach

# pdb_stats$X.ray <- as.numeric(gsub(", ", "", pdb_stats$X.ray))
# pdb_stats$EM <- as.numeric(gsub(", ", "", pdb_stats$EM))
# pdb_stats$Total <- as.numeric(gsub(", ", "", pdb_stats$Total))

# Barry's approach

# install("readr")
library(readr)
stats <- read_csv("Data Export Summary.csv")
```

Rows: 6 Columns: 9

-- Column specification -----  
 Delimiter: ","  
 chr (1): Molecular Type  
 dbl (4): Integrative, Multiple methods, Neutron, Other  
 num (4): X-ray, EM, NMR, Total

i Use `spec()` to retrieve the full column specification for this data.  
 i Specify the column types or set `show\_col\_types = FALSE` to quiet this message.

```
head(stats)
```

```
# A tibble: 6 x 9
  `Molecular Type` `X-ray`    EM    NMR Integrative `Multiple methods` Neutron
  <chr>           <dbl> <dbl> <dbl>        <dbl>           <dbl> <dbl>
1 Protein (only)   176378 20438 12709       342            221   83
2 Protein/Oligosacch~ 10284  3396   34          8             11    1
3 Protein/NA        9007   5931   287         24              7    0
4 Nucleic acid (only) 3077    200   1554         2            15    3
5 Other              174     13    33          3              0    0
6 Oligosaccharide (o~ 11      0     6           0              1    0
# i 2 more variables: Other <dbl>, Total <dbl>
```

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

```
totstructures <- sum(stats$Total)
xray <- sum(stats$"X-ray")
em <- sum(stats$EM)

round((xray)/totstructures * 100, 2)
```

[1] 81.43

```
round(em/totstructures * 100, 2)
```

[1] 12.27

Xray methods are 81.43 % and EM methods are 12.27% of structures.

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

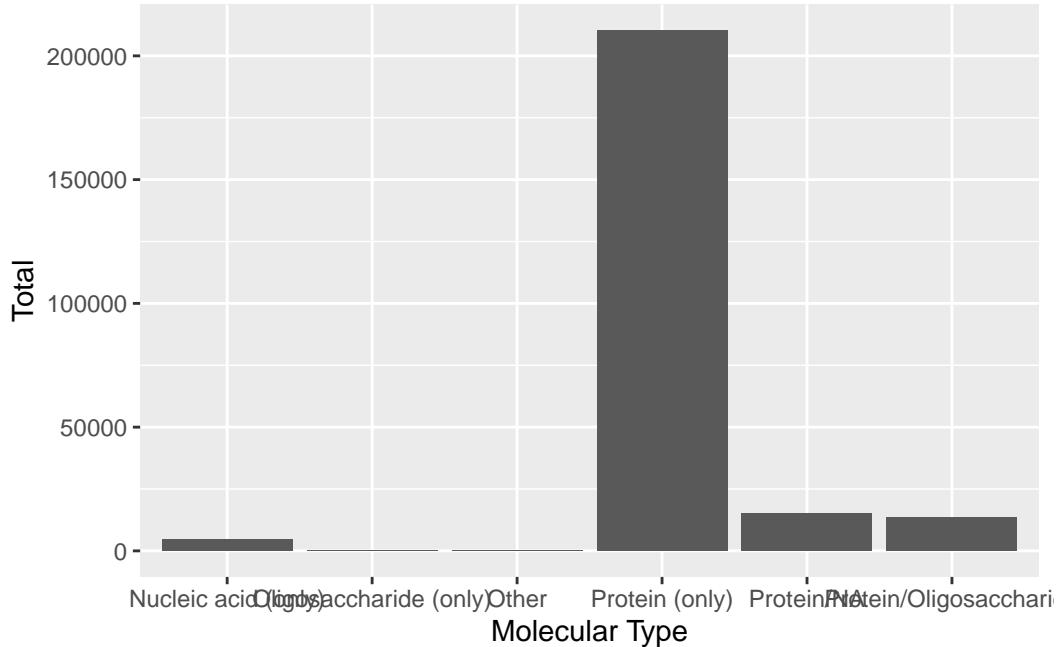
```
pdb_proteins <- stats$Total[1]
round(pdb_proteins/totstructures * 100,2)
```

[1] 86.05

Q3: Make a bar plot overview of molecular type composition using gplot

```
library(ggplot2)

ggplot(stats) + aes(x=`Molecular Type`, y=Total) + geom_col()
```



Extra exercise:

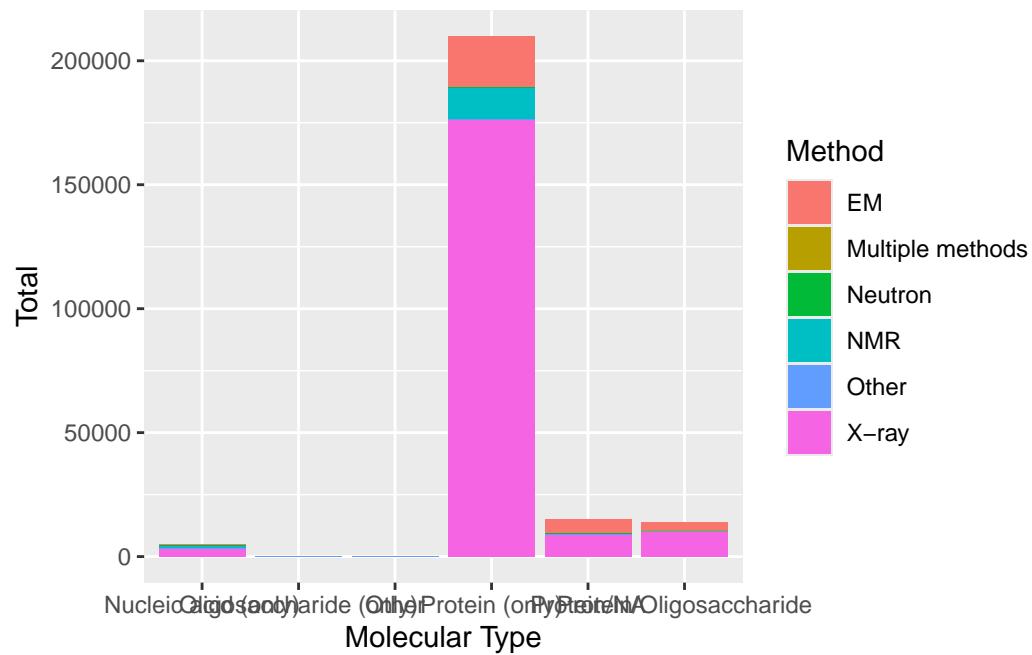
```
library(tidyr)

# Pivot to long format
df_long <- pivot_longer(
  stats,
  cols = c("X-ray", "EM", "NMR", "Multiple methods", "Neutron", "Other"),
  names_to = "Method",
  values_to = "Count"
)

head(df_long)

# A tibble: 6 x 5
`Molecular Type` Integrative Total Method      Count
<chr>           <dbl>   <dbl> <chr>      <dbl>
1 Protein (only)    342  210203 X-ray     176378
2 Protein (only)    342  210203 EM        20438
3 Protein (only)    342  210203 NMR       12709
4 Protein (only)    342  210203 Multiple methods 221
5 Protein (only)    342  210203 Neutron     83
6 Protein (only)    342  210203 Other        32
```

```
ggplot(df_long) + aes(x=Molecular Type, y= Count, fill= Method) + geom_bar(stat = "identity")
```



## Visualizing structure data

The mol\* viewer is embedded in many bioinformatics websites. The homepage is <https://molstar.org>

I can insert/render any figure or image file using markdown format:

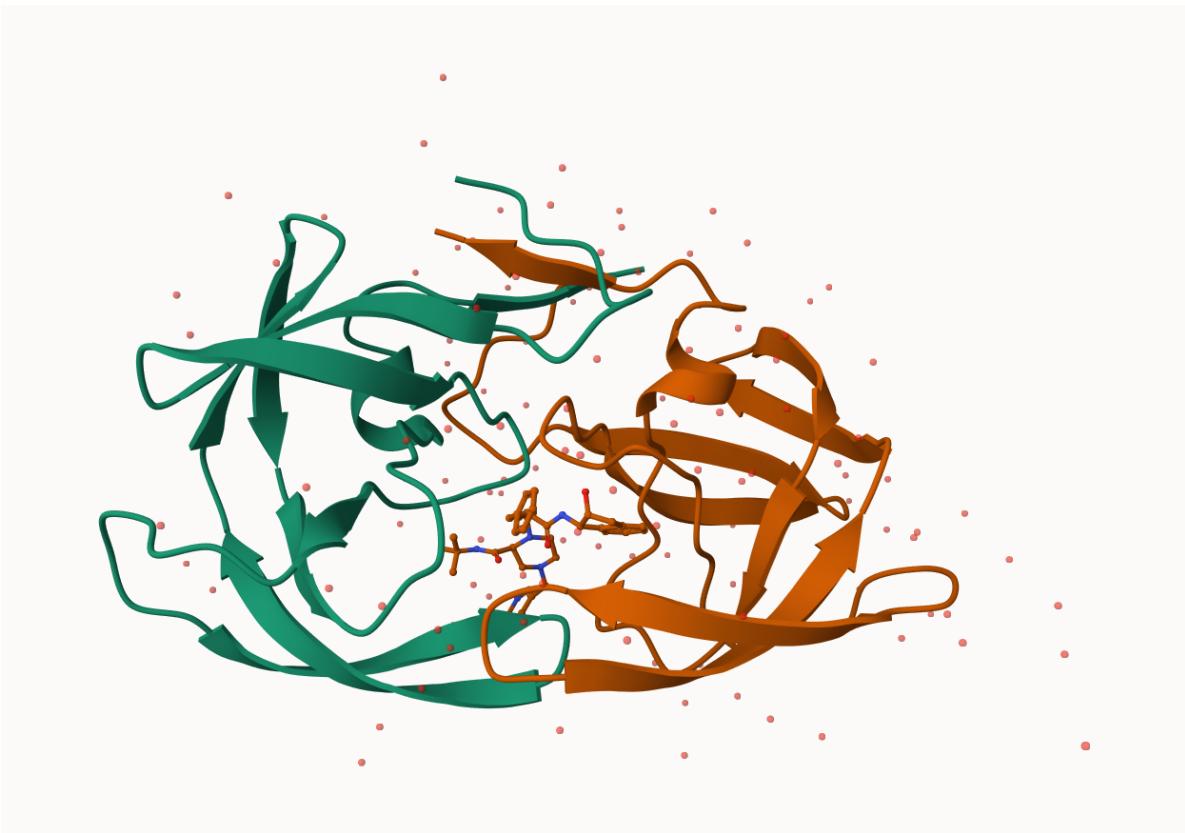


Figure 1: The HIV-Pr dimer with bound inhibitor



Figure 2: The HIV-Pr dimer, reactive aspartic acid (D25) and important H2O (308) are highlighted

### Bio3D package for structural bioinformatics

We can use the bio3d package to read and analyze biomolecular data in R:

```
library(bio3d)  
  
hiv <- read.pdb("1hsg")
```

Note: Accessing on-line PDB file

```
hiv
```

```
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:
PQITLWQRPLVTIKIGGQLKEALLDTGADDTVLEEMSLPGRWPKMIGGGIGGFIKVRQYD
QILIEICGHKAIGTVLVGPTPVNIIGRNLLTQIGCTLNFPQITLWQRPLVTIKIGGQLKE
ALLDTGADDTVLEEMSLPGRWPKMIGGGIGGFIKVRQYDQILIEICGHKAIGTVLVGPTP
VNIIGRNLLTQIGCTLNF

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

```
head(hiv$atom)
```

	type	eleno	elety	alt	resid	chain	resno	insert	x	y	z	o	b
1	ATOM	1	N	<NA>	PRO	A	1	<NA>	29.361	39.686	5.862	1	38.10
2	ATOM	2	CA	<NA>	PRO	A	1	<NA>	30.307	38.663	5.319	1	40.62
3	ATOM	3	C	<NA>	PRO	A	1	<NA>	29.760	38.071	4.022	1	42.64
4	ATOM	4	O	<NA>	PRO	A	1	<NA>	28.600	38.302	3.676	1	43.40
5	ATOM	5	CB	<NA>	PRO	A	1	<NA>	30.508	37.541	6.342	1	37.87
6	ATOM	6	CG	<NA>	PRO	A	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>	C	<NA>										

Let's get the sequence

```
 pdbseq(hiv)
```

```
 1  2  3  4  5  6  7  8  9  10 11 12 13 14 15 16 17 18 19 20
"P" "Q" "I" "T" "L" "W" "Q" "R" "P" "L" "V" "T" "I" "K" "I" "G" "G" "Q" "L" "K"
21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
"E" "A" "L" "L" "D" "T" "G" "A" "D" "D" "T" "V" "L" "E" "E" "M" "S" "L" "P" "G"
41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
"R" "W" "K" "P" "K" "M" "I" "G" "G" "I" "G" "G" "F" "I" "K" "V" "R" "Q" "Y" "D"
61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
"Q" "I" "L" "I" "E" "I" "C" "G" "H" "K" "A" "I" "G" "T" "V" "L" "V" "G" "P" "T"
81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 1
"P" "V" "N" "I" "I" "G" "R" "N" "L" "L" "T" "Q" "I" "G" "C" "T" "L" "N" "F" "P"
 2  3  4  5  6  7  8  9  10 11 12 13 14 15 16 17 18 19 20 21
"Q" "I" "T" "L" "W" "Q" "R" "P" "L" "V" "T" "I" "K" "I" "G" "G" "Q" "L" "K" "E"
22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41
"A" "L" "L" "D" "T" "G" "A" "D" "D" "T" "V" "L" "E" "E" "M" "S" "L" "P" "G" "R"
42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61
"W" "K" "P" "K" "M" "I" "G" "G" "I" "G" "G" "F" "I" "K" "V" "R" "Q" "Y" "D" "Q"
62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80 81
"I" "L" "I" "E" "I" "C" "G" "H" "K" "A" "I" "G" "T" "V" "L" "V" "G" "P" "T" "P"
82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99
"V" "N" "I" "I" "G" "R" "N" "L" "L" "T" "Q" "I" "G" "C" "T" "L" "N" "F"
```

Let's trim to just chain A and get just that chain's sequence:

```
HIV_A <- trim.pdb(hiv, chain="A")
HIV_A.seq <- pdbseq(HIV_A)
HIV_A.seq
```

```
 1  2  3  4  5  6  7  8  9  10 11 12 13 14 15 16 17 18 19 20
"P" "Q" "I" "T" "L" "W" "Q" "R" "P" "L" "V" "T" "I" "K" "I" "G" "G" "Q" "L" "K"
21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
"E" "A" "L" "L" "D" "T" "G" "A" "D" "D" "T" "V" "L" "E" "E" "M" "S" "L" "P" "G"
41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
"R" "W" "K" "P" "K" "M" "I" "G" "G" "I" "G" "G" "F" "I" "K" "V" "R" "Q" "Y" "D"
61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80
"Q" "I" "L" "I" "E" "I" "C" "G" "H" "K" "A" "I" "G" "T" "V" "L" "V" "G" "P" "T"
81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99
"P" "V" "N" "I" "I" "G" "R" "N" "L" "L" "T" "Q" "I" "G" "C" "T" "L" "N" "F"
```

I want to blast it

```
# the above extra code is useful if you'll need to render the document but you don't want to
blast <- blast.pdb(HIV_A.seq)
```

```
Searching ... please wait (updates every 5 seconds) RID = G5FAV7EF014
.
.
.
Reporting 249 hits
```

```
head(blast$hit.tbl)
```

	queryid	subjectids	identity	alignmentlength	mismatches	gapopens	q.start
1	Query_3578775	1W5V_A	100.00		99	0	0
2	Query_3578775	2FDE_A	100.00		99	0	0
3	Query_3578775	1AJV_A	100.00		99	0	0
4	Query_3578775	2R38_A	98.99		99	1	0
5	Query_3578775	2R3T_A	98.99		99	1	0
6	Query_3578775	1HXB_A	98.99		99	1	0

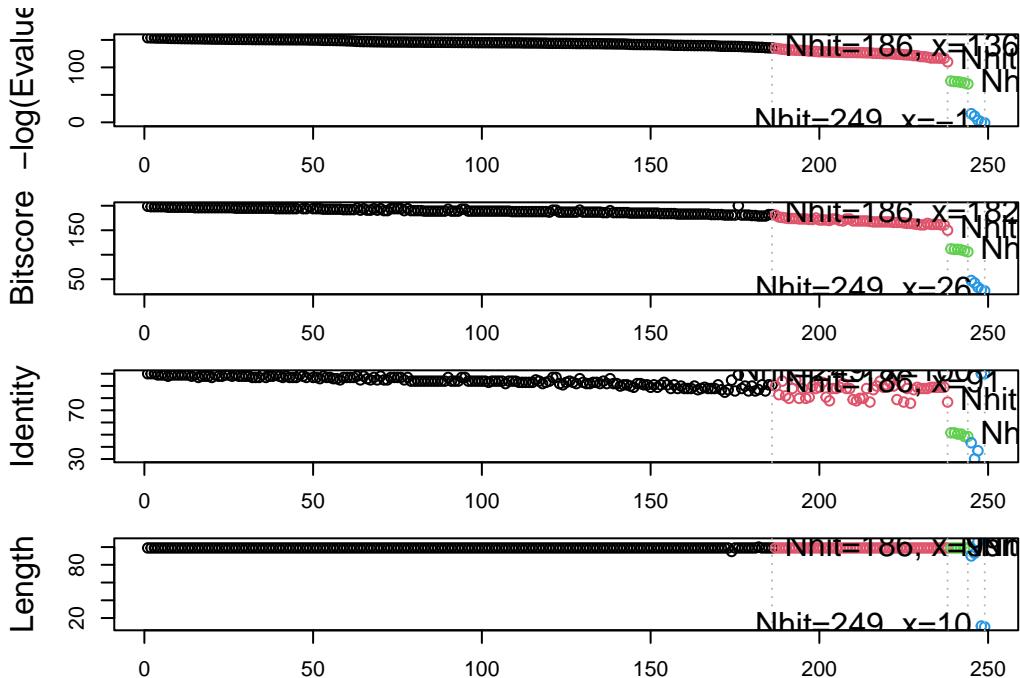
	q.end	s.start	s.end	evalue	bitscore	positives	mlog.evalue	pdb.id	acc
1	99	12	110	1.38e-67		199	100	153.9511	1W5V_A
2	99	2	100	1.70e-67		198	100	153.7426	2FDE_A
3	99	1	99	1.99e-67		198	100	153.5851	1AJV_A
4	99	1	99	2.50e-67		198	100	153.3569	2R38_A
5	99	1	99	2.50e-67		198	100	153.3569	2R3T_A
6	99	1	99	2.50e-67		198	100	153.3569	1HXB_A

What if you plot the blast results

```
hits<-plot(blast)
```

```
* Possible cutoff values:    135 110 69 -2
Yielding Nhits:          186 238 244 249

* Chosen cutoff value of:   69
Yielding Nhits:          244
```



```
hits$pdb.id
```

```
[1] "1W5V_A" "2FDE_A" "1AJV_A" "2R38_A" "2R3T_A" "1HXB_A" "1BV9_A" "1AAQ_A"
[9] "1AXA_A" "1HVS_A" "1ZP8_A" "2QHC_A" "1A8G_A" "204L_A" "5COK_A" "1TCX_A"
[17] "2Z54_A" "1D4S_A" "1BV7_A" "1BWA_A" "1A9M_A" "2FLE_A" "1ODY_A" "1GNN_A"
[25] "1GNM_A" "5YRS_B" "1HEF_E" "10DX_A" "4QGI_A" "1BVE_A" "2AZ8_A" "1A30_A"
[33] "6DH6_A" "6DHO_A" "2I4D_A" "600S_A" "1RL8_A" "5YRS_A" "1ZSF_A" "2Q64_A"
[41] "6DH3_A" "2NPH_A" "2Q63_A" "1LZQ_A" "1FB7_A" "1G6L_A" "1HIV_A" "600U_A"
[49] "1HVC_A" "2I4V_A" "2AZ9_A" "600T_A" "2P3B_B" "5KAO_A" "2WLO_A" "6OPT_A"
[57] "1IZI_A" "1MRX_A" "2PYM_A" "2PYN_A" "1DMP_A" "4K4P_A" "1LV1_A" "1AID_A"
[65] "1LV1_A" "1ZBG_A" "3TKG_A" "1HVC_A" "5YOK_A" "1G6L_A" "1FGC_C" "3K4V_A"
[73] "3KT5_A" "3KT5_A" "4QLH_A" "4QLH_A" "2F3K_A" "4Q5M_A" "2AOC_A" "3B80_A"
[81] "3VF5_A" "2AVQ_A" "1DW6_C" "1KZK_A" "2HS1_A" "1K6C_A" "1MTB_A" "4Q1X_A"
[89] "4Q1W_A" "4Q5M_A" "3D1X_A" "2AVM_A" "3PWM_A" "3KT2_A" "3KT2_A" "1SDV_A"
[97] "3JWV_A" "3OY4_A" "1A94_A" "2HS2_A" "4EJ8_A" "2FGU_A" "2AVV_A" "3JW2_A"
[105] "3BVA_A" "1FFF_C" "3S43_B" "2NXD_A" "1FG6_C" "1EBK_C" "4Q1Y_A" "3EL4_A"
[113] "1F7A_A" "1K2B_A" "2FGV_A" "1Z8C_A" "2G69_A" "3EL9_A" "30XV_A" "1BDR_A"
[121] "3N3I_A" "3N3I_A" "30XW_A" "3S43_A" "3EM3_A" "3CYW_A" "5KQX_A" "2B60_A"
[129] "7DOZ_A" "1K2C_A" "1MT7_A" "3EM4_A" "4QJ9_A" "1BDL_A" "3LZS_A" "5T84_A"
[137] "4DQB_A" "7DOZ_A" "4QJ2_A" "3LZV_A" "1SGU_A" "2FXE_A" "1BDQ_A" "3U71_A"
[145] "2R5P_A" "4OBD_A" "7MAS_A" "3IXO_A" "3D3T_A" "5YOJ_A" "3LZU_A" "4NJS_A"
[153] "3EKP_A" "1B6J_A" "3EKQ_A" "2RKF_A" "1C6X_A" "7MAR_A" "4DQF_A" "1RPI_A"
```

```
[161] "3OU1_B" "3PJ6_A" "2P3A_A" "60GQ_A" "30Q7_A" "5KR1_A" "30QD_A" "4RVI_A"
[169] "30QA_A" "1B6K_A" "30UD_B" "6MK9_A" "3S09_A" "1Q9P_A" "6I45_A" "7SEP_A"
[177] "4NJT_A" "3BXR_A" "4YOA_A" "4DQC_A" "2FDD_A" "2RKG_A" "4DQH_A" "2P3C_A"
[185] "4EP2_A" "4EP2_A" "4EQO_A" "4NPT_A" "6OPU_A" "4NPU_A" "3U7S_A" "3HAW_A"
[193] "2AZB_A" "3TTP_A" "3HBO_A" "3GGU_A" "7N6T_A" "60PV_A" "4EQO_A" "60PX_A"
[201] "204N_A" "5T2E_A" "3UCB_A" "3KA2_A" "3FSM_A" "60PW_A" "2AZC_A" "3FSM_A"
[209] "3HLO_A" "2P3D_A" "3T3C_A" "7MYP_A" "6054_X" "60PY_A" "4Z4X_A" "60PZ_A"
[217] "2JE4_A" "1DAZ_C" "7MAP_A" "7MAQ_A" "1K1U_A" "2B7Z_A" "3MWS_A" "1K1T_A"
[225] "8DCH_A" "3I2L_A" "6P9A_A" "2FXD_A" "2J9J_A" "3DCK_A" "2J9J_B" "3NXE_A"
[233] "2040_A" "2040_A" "3NXE_A" "3KA2_A" "3HLO_A" "5B18_A" "1SIP_A" "2SAM_A"
[241] "1AZ5_A" "1SIV_A" "1HII_A" "1IVP_A"
```

## Prediction of functional motions

We can run a Normal Mode Analysis (NMA) to predict large scale motions/flexibility/dynamics of any biomolecule that we can read into R.

Let's look at ADK

```
adk <- read.pdb("1ake")
```

```
Note: Accessing on-line PDB file
PDB has ALT records, taking A only, rm.alt=TRUE
```

```
adk_A <- trim(adk, chain="A")
adk_A
```

```
Call: trim.pdb(pdb = adk, chain = "A")

Total Models#: 1
Total Atoms#: 1954, XYZs#: 5862 Chains#: 1 (values: A)

Protein Atoms#: 1656 (residues/Calpha atoms#: 214)
Nucleic acid Atoms#: 0 (residues/phosphate atoms#: 0)

Non-protein/nucleic Atoms#: 298 (residues: 242)
Non-protein/nucleic resid values: [ AP5 (1), HOH (241) ]

Protein sequence:
MRIILLGAPGAGKGTQAQFIMEKYGIPQISTGDMRLRAAVKSGSELGKQAKDIMDAGKLVT
```

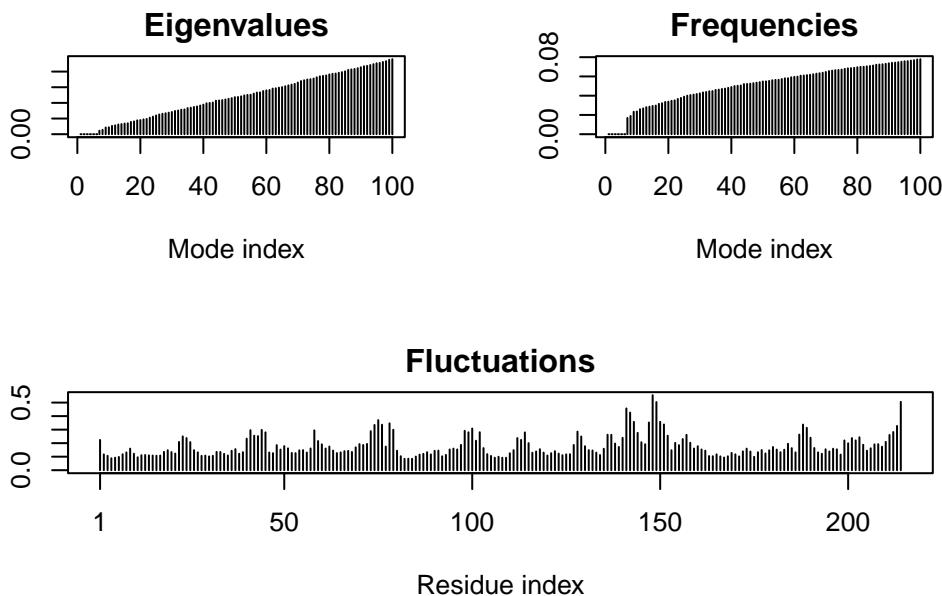
```
DELVIALVKERIAQEDCRNGFLLDGFPR TIPQADAMKEAGINV D YVLEFDVPDELIVDRI  
VGRRVHAPSGRVYHVKF NPPKVEGKDDVTGEELTTRKDDQEETVRKRLVEYHQMTAPLIG  
YYSKEAEAGNTKYAKVDGTPVAEVRA DLEKILG
```

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

```
m <- nma(adk_A)
```

```
Building Hessian...           Done in 0.012 seconds.  
Diagonalizing Hessian...     Done in 0.268 seconds.
```

```
plot(m)
```



Let's write out a “trajectory” of predicted motion

```
mktrj(m, file="adk_nma.pdb")
```

You can load this into molstar and play the movie!

## Play with 3D viewing in R

We can use the new **bio3dview** package, which is not yet on CRAN, to render interactive 3D views in R and HTML quarto output reports.

To install from github, we can use the **pak** package.

```
# install.packages("pak")
#
# pak::pak("bioboot/bio3dview")
```