

# Lab 6: R Functions

Vivian Cai

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## Improving analysis code by writing functions

A.

1) Original code:

```
# (A. Can you improve this analysis code?)
df <- data.frame(a=1:10, b=seq(200,400,length=10),c=11:20,d=NA)
df$a <- (df$a - min(df$a)) / (max(df$a) - min(df$a))
df$b <- (df$b - min(df$a)) / (max(df$b) - min(df$b))
df$c <- (df$c - min(df$c)) / (max(df$c) - min(df$c))
df$d <- (df$d - min(df$d)) / (max(df$a) - min(df$d))
```

2) Fix copy/paste errors:

```
df <- data.frame(a=1:10, b=seq(200,400,length=10),c=11:20,d=NA)
df$a <- (df$a - min(df$a)) / (max(df$a) - min(df$a))
df$b <- (df$b - min(df$b)) / (max(df$b) - min(df$b))
df$c <- (df$c - min(df$c)) / (max(df$c) - min(df$c))
df$d <- (df$d - min(df$d)) / (max(df$d) - min(df$d))
```

3) Write a function to clean it up:

```
df <- data.frame(a=1:10, b=seq(200,400,length=10),c=11:20,d=NA)
scaling <- function(x, na.rm = TRUE) {
  rng <- range(x, na.rm = na.rm)
  (x - rng[1]) / (rng[2] - rng[1])
}

scaling(df$a)
```

```
## [1] 0.0000000 0.1111111 0.2222222 0.3333333 0.4444444 0.5555556 0.6666667
## [8] 0.7777778 0.8888889 1.0000000
```

B.

1) Original code:

```

# Can you improve this analysis code?
#install.packages("bio3d")
library(bio3d)
s1 <- read.pdb("4AKE") # kinase with drug

## Note: Accessing on-line PDB file

s2 <- read.pdb("1AKE") # kinase no drug

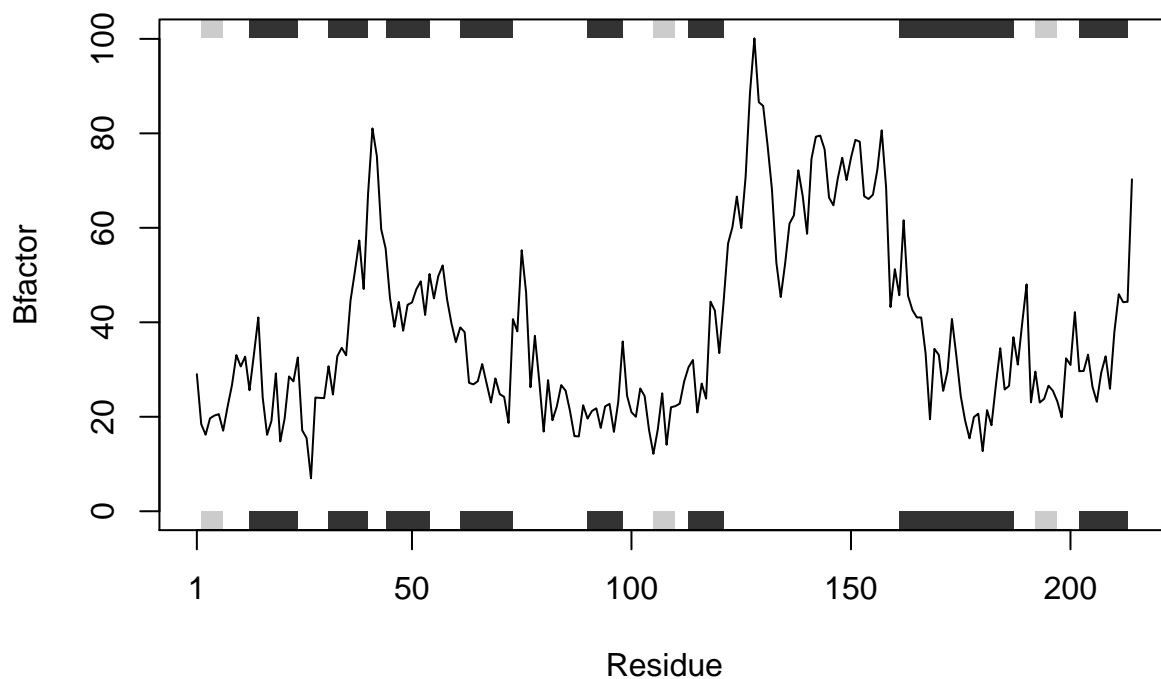
## Note: Accessing on-line PDB file
## PDB has ALT records, taking A only, rm.alt=TRUE

s3 <- read.pdb("1E4Y") # kinase with drug

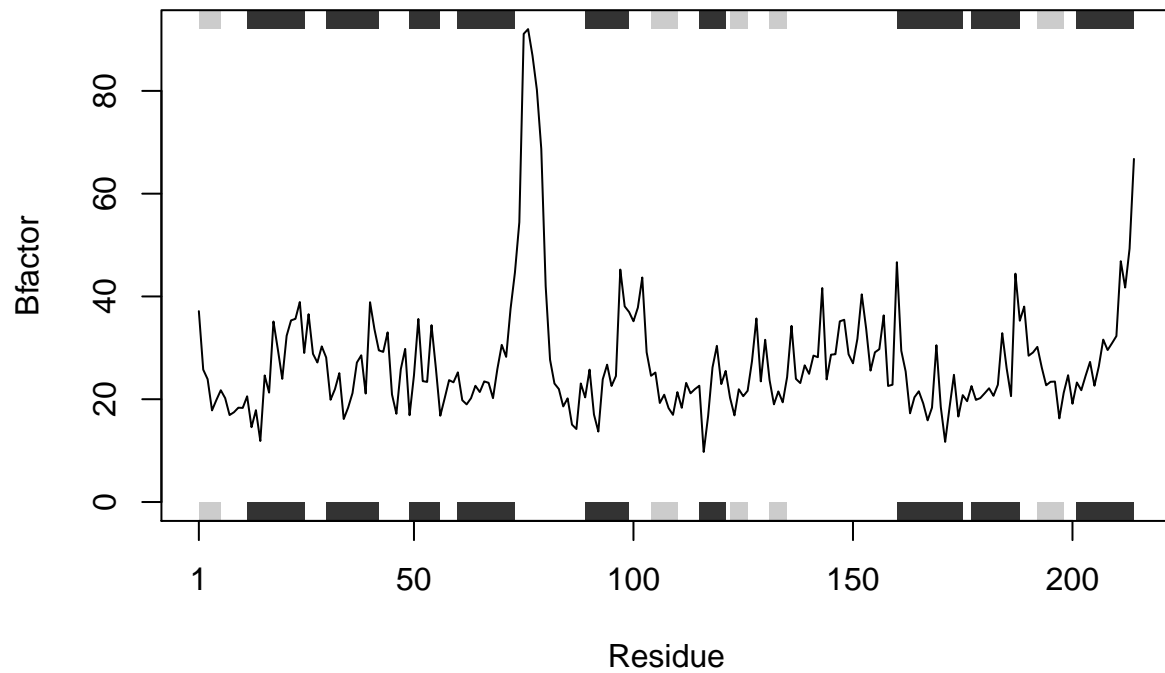
## Note: Accessing on-line PDB file

s1.chainA <- trim.pdb(s1, chain="A", elety="CA")
s2.chainA <- trim.pdb(s2, chain="A", elety="CA")
s3.chainA <- trim.pdb(s1, chain="A", elety="CA")
s1.b <- s1.chainA$atom$b
s2.b <- s2.chainA$atom$b
s3.b <- s3.chainA$atom$b
plotb3(s1.b, sse=s1.chainA, typ="l", ylab="Bfactor")

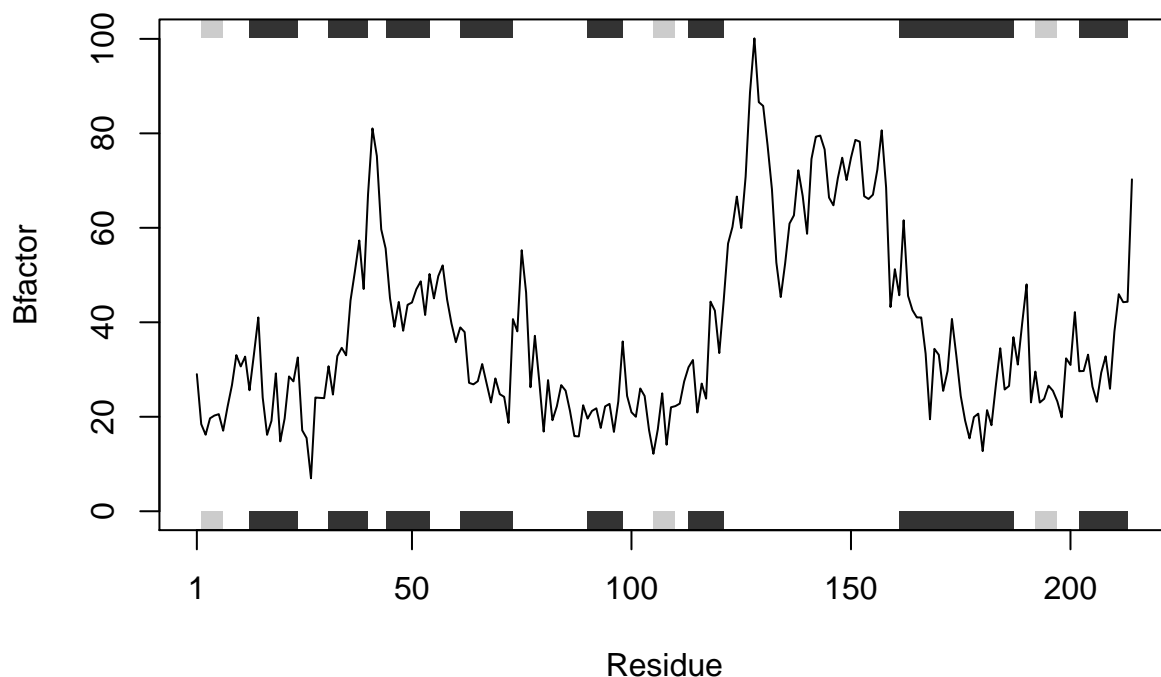
```



```
plotb3(s2.b, sse=s2.chainA, typ="l", ylab="Bfactor")
```



```
plotb3(s3.b, sse=s3.chainA, typ="l", ylab="Bfactor")
```



2) Copy/Paste error gone:

```
# Can you improve this analysis code?
install.packages("bio3d")
library(bio3d)
s1 <- read.pdb("4AKE") # kinase with drug
```

```
## Note: Accessing on-line PDB file
```

```
## Warning in get.pdb(file, path = tempdir(), verbose = FALSE): /var/folders/8g/
## c7740b013vd3vszd57ym1z3r0000gn/T//Rtmpz8nUcu/4AKE.pdb exists. Skipping download
```

```
s2 <- read.pdb("1AKE") # kinase no drug
```

```
## Note: Accessing on-line PDB file
```

```
## Warning in get.pdb(file, path = tempdir(), verbose = FALSE): /var/folders/8g/
## c7740b013vd3vszd57ym1z3r0000gn/T//Rtmpz8nUcu/1AKE.pdb exists. Skipping download
```

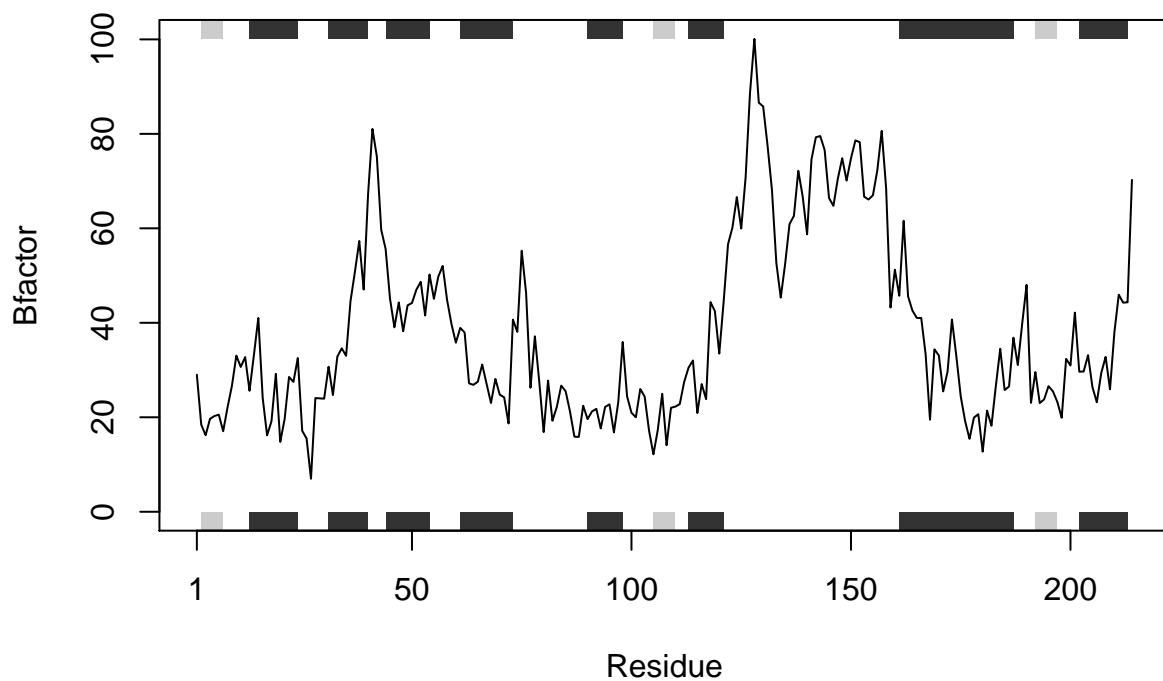
```
## PDB has ALT records, taking A only, rm.alt=TRUE
```

```
s3 <- read.pdb("1E4Y") # kinase with drug
```

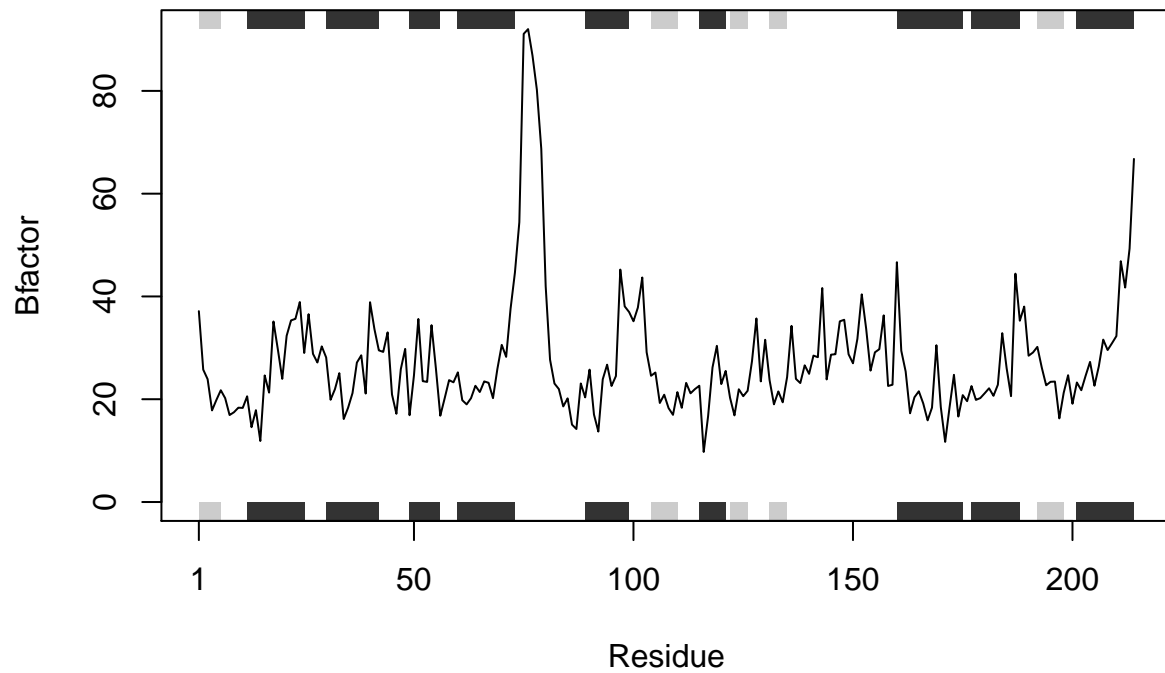
```
## Note: Accessing on-line PDB file
```

```
## Warning in get.pdb(file, path = tempdir(), verbose = FALSE): /var/folders/8g/  
## c7740b013vd3vszd57ym1z3r0000gn/T//Rtmpz8nUcu/1E4Y.pdb exists. Skipping download
```

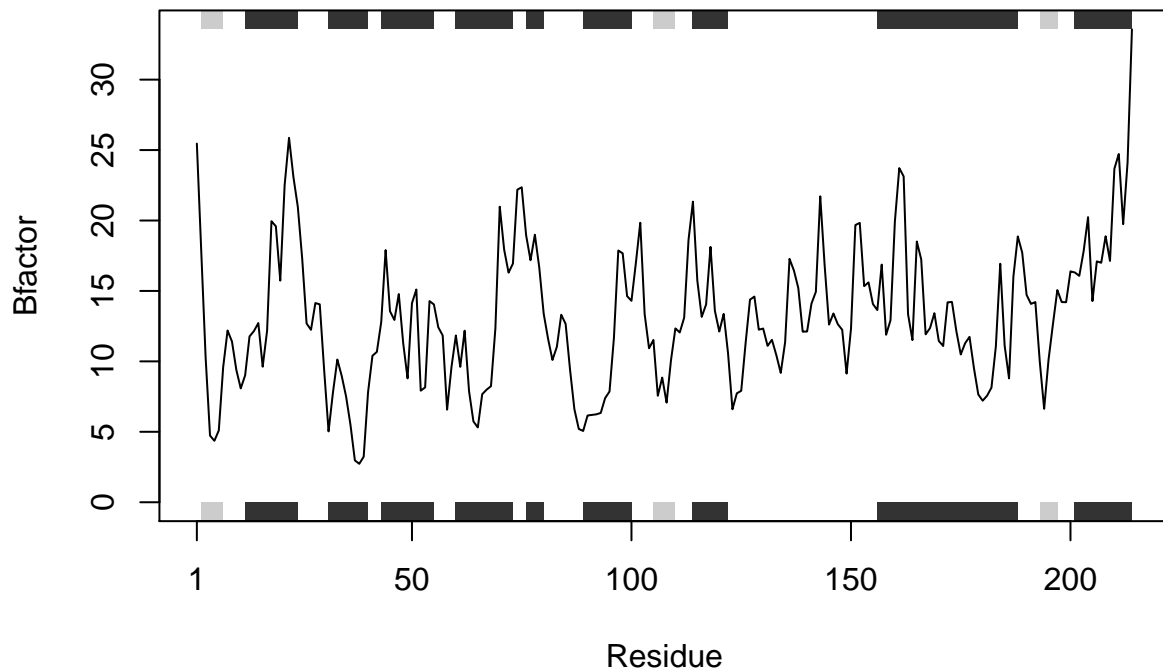
```
s1.chainA <- trim.pdb(s1, chain="A", elety="CA")  
s2.chainA <- trim.pdb(s2, chain="A", elety="CA")  
s3.chainA <- trim.pdb(s3, chain="A", elety="CA")  
s1.b <- s1.chainA$atom$b  
s2.b <- s2.chainA$atom$b  
s3.b <- s3.chainA$atom$b  
plotb3(s1.b, sse=s1.chainA, typ="l", ylab="Bfactor")
```



```
plotb3(s2.b, sse=s2.chainA, typ="l", ylab="Bfactor")
```



```
plotb3(s3.b, sse=s3.chainA, typ="l", ylab="Bfactor")
```



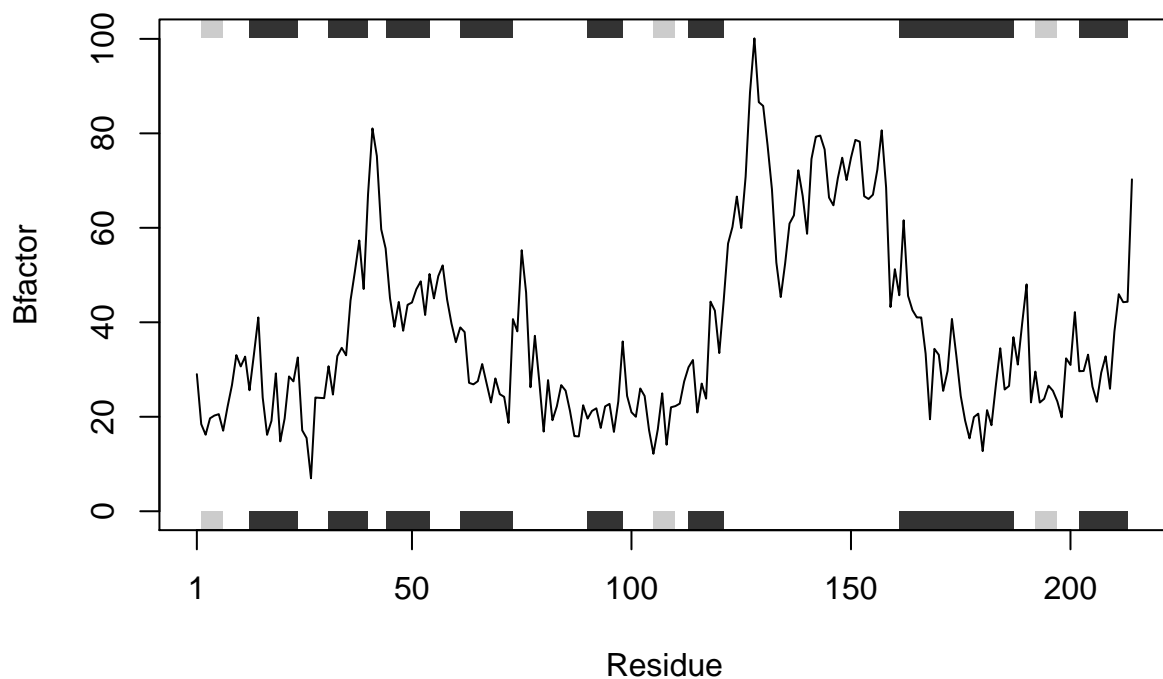
3) Clean up with a new function:

```
PDB_Bfac_plt <- function(x, chain = "A", elety = "CA", typ = "l", ylab = "Bfactor") {
  PDB <- read.pdb(x)
  Chain_PDB <- trim.pdb(PDB, chain = chain, elety = elety)
  Chain_atomb <- Chain_PDB$atom$b
  plotb3(Chain_atomb, sse = Chain_PDB, typ = typ, ylab = ylab)
}

PDB_Bfac_plt("4AKE")
```

## Note: Accessing on-line PDB file

```
## Warning in get.pdb(file, path = tempdir(), verbose = FALSE): /var/folders/8g/
## c7740b013vd3vszd57ym1z3r0000gn/T//Rtmpz8nUcu/4AKE.pdb exists. Skipping download
```



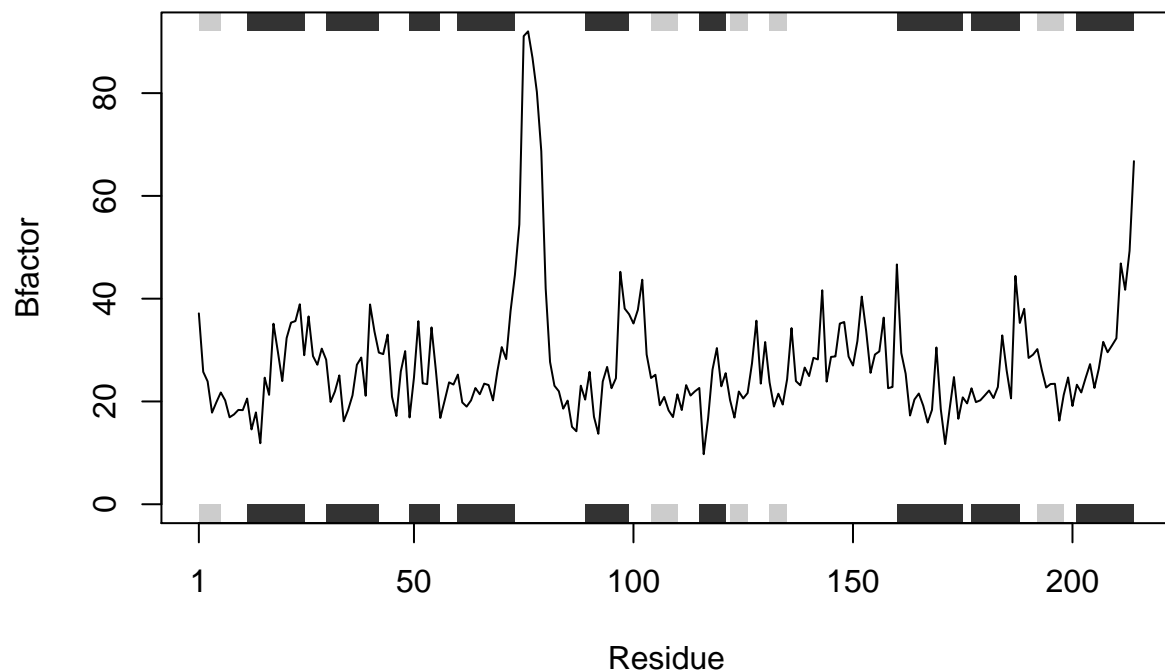
```
PDB_Bfac_plt("1AKE")
```

```
## Note: Accessing on-line PDB file
```

```
## Warning in get.pdb(file, path = tempdir(), verbose = FALSE): /var/folders/8g/
## c7740b013vd3vszd57ym1z3r0000gn/T//Rtmpz8nUcu/1AKE.pdb exists. Skipping download
```

```
## PDB has ALT records, taking A only, rm.alt=TRUE
```

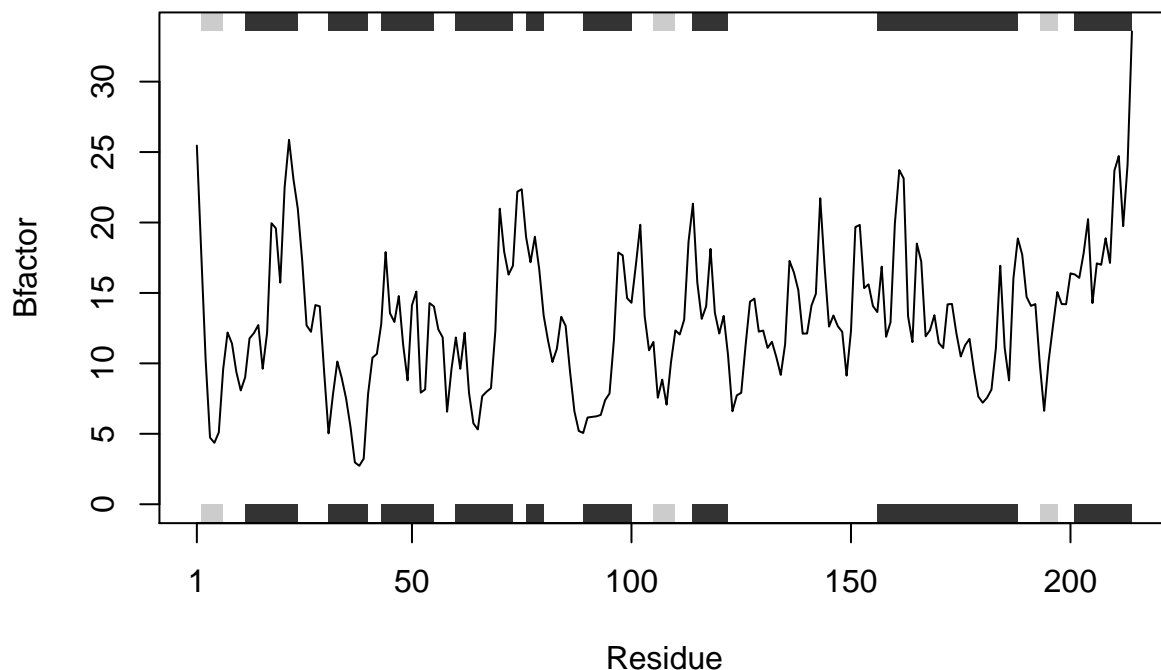




```
PDB_Bfac_plt("1E4Y")
```

```
## Note: Accessing on-line PDB file
```

```
## Warning in get.pdb(file, path = tempdir(), verbose = FALSE): /var/folders/8g/  
## c7740b013vd3vszd57ym1z3r0000gn/T//Rtmpz8nUcu/1E4Y.pdb exists. Skipping download
```



4) Questions: **Q1. What type of object is returned from the read.pdb() function?**

```
str(read.pdb("4AKE"))
```

```
## Note: Accessing on-line PDB file
```

```
## Warning in get.pdb(file, path = tempdir(), verbose = FALSE): /var/folders/8g/
## c7740b013vd3vszd57ym1z3r0000gn/T//Rtmpz8nUcu/4AKE.pdb exists. Skipping download
```

```
## List of 8
```

```
## $ atom : 'data.frame': 3459 obs. of 16 variables:
## ..$ type : chr [1:3459] "ATOM" "ATOM" "ATOM" "ATOM" ...
## ..$ eleno : int [1:3459] 1 2 3 4 5 6 7 8 9 10 ...
## ..$ elety : chr [1:3459] "N" "CA" "C" "O" ...
## ..$ alt : chr [1:3459] NA NA NA NA ...
## ..$ resid : chr [1:3459] "MET" "MET" "MET" "MET" ...
## ..$ chain : chr [1:3459] "A" "A" "A" "A" ...
## ..$ resno : int [1:3459] 1 1 1 1 1 1 1 1 2 2 ...
## ..$ insert: chr [1:3459] NA NA NA NA ...
## ..$ x : num [1:3459] -10.93 -9.9 -9.17 -9.8 -10.59 ...
## ..$ y : num [1:3459] -24.9 -24.4 -23.3 -22.3 -24 ...
## ..$ z : num [1:3459] -9.52 -10.48 -9.81 -9.35 -11.77 ...
## ..$ o : num [1:3459] 1 1 1 1 1 1 1 1 1 1 ...
## ..$ b : num [1:3459] 41.5 29 27.9 26.4 34.2 ...
## ..$ segid : chr [1:3459] NA NA NA NA ...
## ..$ elesy : chr [1:3459] "N" "C" "C" "O" ...
```

```
## ..$ charge: chr [1:3459] NA NA NA NA ...
## $ xyz : 'xyz' num [1, 1:10377] -10.93 -24.89 -9.52 -9.9 -24.42 ...
## $ seqres: Named chr [1:428] "MET" "ARG" "ILE" "ILE" ...
## ..- attr(*, "names")= chr [1:428] "A" "A" "A" "A" ...
## $ helix :List of 4
## ..$ start: Named num [1:19] 13 31 44 61 75 90 113 161 202 13 ...
## .. ..- attr(*, "names")= chr [1:19] "" "" "" "" ...
## ..$ end : Named num [1:19] 24 40 54 73 77 98 121 187 213 24 ...
## .. ..- attr(*, "names")= chr [1:19] "" "" "" "" ...
## ..$ chain: chr [1:19] "A" "A" "A" "A" ...
## ..$ type : chr [1:19] "5" "1" "1" "1" ...
## $ sheet :List of 4
## ..$ start: Named num [1:14] 192 105 2 81 27 123 131 192 105 2 ...
## .. ..- attr(*, "names")= chr [1:14] "" "" "" "" ...
## ..$ end : Named num [1:14] 197 110 7 84 29 126 134 197 110 7 ...
## .. ..- attr(*, "names")= chr [1:14] "" "" "" "" ...
## ..$ chain: chr [1:14] "A" "A" "A" "A" ...
## ..$ sense: chr [1:14] "0" "1" "1" "1" ...
## $ calpha: logi [1:3459] FALSE TRUE FALSE FALSE FALSE FALSE ...
## $ remark:List of 1
## ..$ biomat:List of 4
## .. ..$ num : int 1
## .. ..$ chain :List of 1
## .. .. ..$ : chr [1:2] "A" "B"
## .. ..$ mat :List of 1
## .. .. ..$ :List of 1
## .. .. .. ..$ A B: num [1:3, 1:4] 1 0 0 0 1 0 0 0 1 0 ...
## .. ..$ method: chr "AUTHOR"
## $ call : language read.pdb(file = "4AKE")
## - attr(*, "class")= chr [1:2] "pdb" "sse"
```

A data frame is returned.

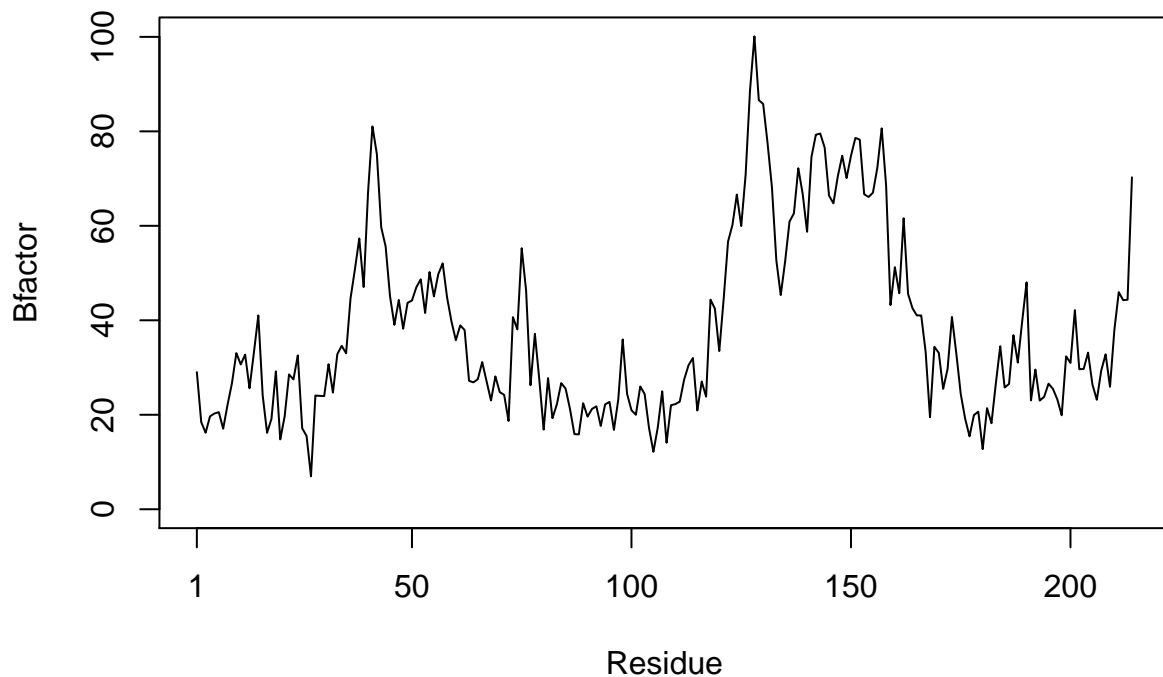
**Q2. What does the trim.pdb() function do?**

```
?trim.pdb
```

It creates a smaller PDB from a given PDB object, containing a subset of atoms from the original PDB.

**Q3. What input parameter would turn off the marginal black and grey rectangles in the plots and what do they represent in this case?**

```
plotb3(s1.b, sse=NULL, typ="l", ylab="Bfactor")
```



As shown, turning off the sse argument gets rid of the bars, which are indications of secondary structures as returned from dssp, stride or in certain cases read.pdb.

**Q4. What would be a better plot to compare across the different proteins?** I think the same line plot with 3 Bfactor lines showing on the same plot would be good for comparison, since all 3 proteins of interest are the same length.

**Q5. Which proteins are more similar to each other in their B-factor trends. How could you quantify this? HINT: try the rbind(), dist() and hclust() functions together with a resulting dendrogram plot. Look up the documentation to see what each of these functions does.**

```
hc <- hclust( dist( rbind(s1.b, s2.b, s3.b) ) )
plot(hc)
```

## Cluster Dendrogram



```
dist(rbind(s1.b, s2.b, s3.b))
hclust (*, "complete")
```

rbind combines the rows of a chosen data.frame, in this case, all the b-factor values for each protein we are looking at are combined into the same dataframe.(As shown below)

```
head(rbind(s1.b, s2.b, s3.b))
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12]
## s1.b 29.02 18.44 16.20 19.67 20.26 20.55 17.05 22.13 26.71 33.05 30.66 32.73
## s2.b 37.14 25.76 23.90 17.83 19.86 21.75 20.21 16.92 17.47 18.35 18.31 20.57
## s3.b 25.46 17.86 10.28 4.73 4.36 5.10 9.59 12.19 11.41 9.39 8.08 9.01
##      [,13] [,14] [,15] [,16] [,17] [,18] [,19] [,20] [,21] [,22] [,23] [,24]
## s1.b 25.61 33.19 41.03 24.09 16.18 19.14 29.19 14.79 19.63 28.54 27.49 32.56
## s2.b 14.56 17.87 11.87 24.63 21.29 35.13 29.68 23.96 32.34 35.34 35.64 38.91
## s3.b 11.77 12.15 12.72 9.62 12.18 19.95 19.59 15.73 22.51 25.87 23.08 20.97
##      [,25] [,26] [,27] [,28] [,29] [,30] [,31] [,32] [,33] [,34] [,35] [,36]
## s1.b 17.13 15.50 6.98 24.07 24.00 23.94 30.70 24.70 32.84 34.60 33.01 44.60
## s2.b 29.00 36.55 28.83 27.15 30.28 28.13 19.90 21.95 25.07 16.15 18.35 21.19
## s3.b 17.28 12.69 12.24 14.14 14.05 9.38 5.03 7.78 10.13 8.96 7.50 5.48
##      [,37] [,38] [,39] [,40] [,41] [,42] [,43] [,44] [,45] [,46] [,47] [,48]
## s1.b 50.74 57.32 47.04 67.13 81.04 75.20 59.68 55.63 45.12 39.04 44.31 38.21
## s2.b 27.13 28.55 21.10 38.88 33.63 29.51 29.21 33.01 20.92 17.17 25.84 29.80
## s3.b 2.97 2.73 3.23 7.81 10.40 10.67 12.79 17.90 13.56 12.94 14.78 11.31
##      [,49] [,50] [,51] [,52] [,53] [,54] [,55] [,56] [,57] [,58] [,59] [,60]
## s1.b 43.70 44.19 47.00 48.67 41.54 50.22 45.07 49.77 52.04 44.82 39.75 35.79
## s2.b 16.89 24.66 35.62 23.52 23.37 34.41 25.96 16.79 20.20 23.72 23.29 25.23
## s3.b 8.79 14.13 15.10 7.92 8.15 14.28 14.04 12.42 11.84 6.57 9.59 11.84
##      [,61] [,62] [,63] [,64] [,65] [,66] [,67] [,68] [,69] [,70] [,71] [,72]
```

```

## s1.b 38.92 37.93 27.18 26.86 27.53 31.16 27.08 23.03 28.12 24.78 24.22 18.69
## s2.b 19.81 19.00 20.21 22.62 21.40 23.47 23.20 20.21 25.90 30.58 28.25 37.60
## s3.b 9.61 12.18 7.89 5.74 5.31 7.67 7.99 8.24 12.34 20.98 17.93 16.30
##      [,73] [,74] [,75] [,76] [,77] [,78] [,79] [,80] [,81] [,82] [,83] [,84]
## s1.b 40.67 38.08 55.26 46.29 26.25 37.14 27.50 16.86 27.76 19.27 22.22 26.70
## s2.b 44.66 54.46 91.10 92.02 86.85 80.21 68.72 42.01 27.69 23.06 21.98 18.60
## s3.b 16.94 22.19 22.36 18.96 17.18 18.99 16.65 13.39 11.61 10.10 11.03 13.31
##      [,85] [,86] [,87] [,88] [,89] [,90] [,91] [,92] [,93] [,94] [,95] [,96]
## s1.b 25.52 21.22 15.9 15.84 22.44 19.61 21.23 21.79 17.64 22.19 22.73 16.80
## s2.b 20.17 15.06 14.2 23.07 20.36 25.76 17.02 13.71 23.88 26.72 22.58 24.51
## s3.b 12.66 9.44 6.6 5.20 5.06 6.16 6.20 6.24 6.34 7.39 7.86 11.66
##      [,97] [,98] [,99] [,100] [,101] [,102] [,103] [,104] [,105] [,106] [,107]
## s1.b 23.25 35.95 24.42 20.96 20.00 25.99 24.39 17.19 12.16 17.35 24.97
## s2.b 45.23 38.07 36.97 35.17 37.83 43.69 29.14 24.56 25.20 19.27 20.88
## s3.b 17.87 17.67 14.63 14.30 16.98 19.84 13.36 10.93 11.52 7.56 8.85
##      [,108] [,109] [,110] [,111] [,112] [,113] [,114] [,115] [,116] [,117]
## s1.b 14.08 22.01 22.26 22.78 27.47 30.49 32.02 20.90 27.03 23.84
## s2.b 18.27 16.96 21.38 18.33 23.18 21.15 21.97 22.63 9.74 16.71
## s3.b 7.07 10.08 12.34 12.05 13.10 18.63 21.34 15.73 13.16 14.04
##      [,118] [,119] [,120] [,121] [,122] [,123] [,124] [,125] [,126] [,127]
## s1.b 44.37 42.47 33.48 44.56 56.67 60.18 66.62 59.95 70.81 88.63
## s2.b 26.18 30.39 22.95 25.51 20.28 16.86 21.94 20.59 21.64 27.42
## s3.b 18.13 13.59 12.12 13.37 10.57 6.60 7.73 7.91 11.31 14.38
##      [,128] [,129] [,130] [,131] [,132] [,133] [,134] [,135] [,136] [,137]
## s1.b 100.11 86.60 85.80 77.48 68.13 52.66 45.34 52.43 60.90 62.64
## s2.b 35.72 23.47 31.57 23.71 19.01 21.52 19.40 24.32 34.28 23.96
## s3.b 14.60 12.25 12.33 11.10 11.53 10.44 9.18 11.36 17.28 16.45
##      [,138] [,139] [,140] [,141] [,142] [,143] [,144] [,145] [,146] [,147]
## s1.b 72.19 66.75 58.73 74.57 79.29 79.53 76.58 66.40 64.76 70.48
## s2.b 23.14 26.60 24.94 28.49 28.18 41.64 23.85 28.67 28.76 35.16
## s3.b 15.21 12.11 12.12 14.10 14.94 21.72 16.82 12.61 13.40 12.64
##      [,148] [,149] [,150] [,151] [,152] [,153] [,154] [,155] [,156] [,157]
## s1.b 74.84 70.11 74.82 78.61 78.24 66.70 66.10 67.01 72.28 80.64
## s2.b 35.46 28.74 26.99 31.74 40.41 33.73 25.57 29.13 29.74 36.32
## s3.b 12.24 9.13 12.31 19.68 19.83 15.34 15.61 14.07 13.64 16.87
##      [,158] [,159] [,160] [,161] [,162] [,163] [,164] [,165] [,166] [,167]
## s1.b 68.54 43.23 51.24 45.72 61.60 45.61 42.57 41.03 41.02 33.34
## s2.b 22.58 22.82 46.67 29.44 25.40 17.27 20.38 21.55 19.19 15.89
## s3.b 11.89 12.92 19.93 23.72 23.13 13.35 11.51 18.51 17.24 11.92
##      [,168] [,169] [,170] [,171] [,172] [,173] [,174] [,175] [,176] [,177]
## s1.b 19.48 34.38 33.11 25.48 29.68 40.71 32.91 24.41 19.20 15.43
## s2.b 18.37 30.51 18.47 11.70 18.45 24.75 16.63 20.80 19.62 22.56
## s3.b 12.36 13.42 11.45 11.09 14.19 14.22 12.15 10.49 11.29 11.74
##      [,178] [,179] [,180] [,181] [,182] [,183] [,184] [,185] [,186] [,187]
## s1.b 19.93 20.66 12.72 21.40 18.21 26.68 34.50 25.77 26.52 36.85
## s2.b 19.87 20.22 21.16 22.13 20.66 22.82 32.86 26.04 20.60 44.44
## s3.b 9.53 7.65 7.21 7.56 8.14 11.07 16.93 11.12 8.79 16.03
##      [,188] [,189] [,190] [,191] [,192] [,193] [,194] [,195] [,196] [,197]
## s1.b 31.05 39.84 48.03 23.04 29.57 23.00 23.80 26.59 25.49 23.25
## s2.b 35.28 38.03 28.46 29.10 30.19 26.17 22.71 23.39 23.44 16.27
## s3.b 18.87 17.72 14.72 14.08 14.21 9.99 6.63 10.11 12.64 15.06
##      [,198] [,199] [,200] [,201] [,202] [,203] [,204] [,205] [,206] [,207]
## s1.b 19.89 32.37 30.97 42.16 29.64 29.69 33.15 26.38 23.17 29.35
## s2.b 21.26 24.67 19.12 23.26 21.75 24.59 27.26 22.63 26.40 31.60

```

```
## s3.b  14.21  14.20  16.39  16.31  16.07  17.83  20.24  14.28  17.10  17.00
##      [,208] [,209] [,210] [,211] [,212] [,213] [,214]
## s1.b  32.80  25.92  38.01  45.95  44.26  44.35  70.26
## s2.b  29.57  30.90  32.29  46.86  41.73  49.31  66.76
## s3.b  18.88  17.13  23.68  24.72  19.74  24.12  33.57
```

Then, `dist` computes the distance between rows of the given `data.frame`. Therefore, each element in each row is considered and compared with the other rows. The `dist` result calculated is then used to make the dendrogram. In our case, `s1` and `s3` are more similar to each other than they are to `s2`.

**Q6. How would you generalize the original code above to work with any set of input protein structures?** This is already done using my `PDB_Bfac_plt()` function.