Lab Book 20 11 15

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Monday, Tuesday, Wednesday

In the first part of the week I started setting up the GSEA using the RGSEA script. This turned out to be more complicated than I first thought because 1) I have never worked off an imported script file before and 2) the script John gave me turned out to need a bit of modifying.

The first problem I had was getting the import of the ensembl database information appropriate for my data. I was using Biomart to do this, and as it turns out, Biomart are having some issues and the info is now being managed by ensembl.org. So the first part is:

```
setwd ("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/") #set working directory
CHMP2B_exp <- read.csv("eset_CHMP2B_250615_exprs.csv") #import expression file
row.names(CHMP2B_exp) <- CHMP2B_exp[,1] #make Probe IDs row names if not already done
CHMP2B_exp[,1] <- NULL #remove column containing probe IDs</pre>
library (WGCNA)
options(stringsAsFactors = FALSE)
DATA <- CHMP2B_exp
DATA <- t(DATA) #transpose data (this is required for Biomart)
####GSEA
### connect to the biomart database that is correct for your data ###
mart <- useMart("ENSEMBL_MART_ENSEMBL",dataset="hsapiens_gene_ensembl", host="www.ensembl.org")
x <- colnames(DATA) #create vector containing probe IDs
### Retrive the specified attributes from the above database ###
# mart_attribute <- listAttributes(mart)</pre>
# mart_filter <- listFilters(mart)</pre>
\# mart_back <- getBM(attributes =c("hgnc_symbol", "entrezgene", "ensembl_transcript_id"),
                     filters="ensembl_transcript_id", values=x, mart=mart)
### Create array of attributes required for your data set ###
mart_back <- getBM(attributes=c("hgnc_symbol", "entrezgene", "ensembl_transcript_id", "affy_hg_u133_plus
                   filters = "affy hg u133 plus 2", values = x, mart = mart)
e2 <- mart_back[,1] #take hgnc label column
e3 <- which(!(e2 == "")) #remove any rows with blank cells
mart_back1 <- mart_back[c(e3),] #apply to data frame</pre>
```

Now you have a list of hgnc labels for your genes. The next stage is to replace the probe IDs in your data set with the gene labels

```
### Assign hgnc symbol as column names of expression data ###
#This looks at the ID probes from the expression file and corresponds them to the list
#ensembl_transcript_ID. The expression file column names are then changed to the corresponding
#hgnc symbol

#This will take a decent amount of time, so don't worry if it runs for 30 minutes+

DATA1 <- DATA

for (i in 1:length(mart_back1[,4]))
{
    c1 <- which (x %in% mart_back1[i,4])
    c1 <- c1[i]
#colnames (DATA1)[c(c1)] <- mart_back1[i,2]
    colnames (DATA1)[c(c1)] <- mart_back1[i,1]
}
#t <- array (dim =c(length(DATA[,1]), length (DATA[1,]), d))
DATA1b <- t(DATA1) #transpose back
write.csv(x = DATA1b, file = "/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/CHMP2B_ex</pre>
```

Thursday

So far I had been managing by creating my gct files externally in excel and then using them in the GSEA code. However, a problem came about when I realised that when you assign gene names to probes you get multiple probes per gene. This meant that I had no good way of either normalising across the probes or picking a suitable representative probe. I decided then to try and understand a section of code given to me by John that creates the gct file in the correct format.

```
t2 <- DATA1 #t2 needs samples as row names and gene IDs as column names

# t1 <- grep("ENST", colnames(t)) #This is for when you are using data with ensembl transcript IDs

# t2 <- t [,-c(t1)]

t1a <- which (colnames(t2) %in% "") #take column names in t2 that contain blanks

if (length (t1a)>0) {t2 <- t2 [,-c(t1a)]} #if there are any blanks, remove them (this will be 0 if you

t2.colnames <- toupper(colnames(t2)) #convert all column names of t2 to uppercase

#create an empty matrix with same rows as t2, but with only the same number of columns as unique names

t2.agg <- matrix (nrow=length(t2[,1]), ncol=length(unique(t2.colnames)))

row.names (t2.agg) <- row.names (t2) #make the row names the same as t2

colnames (t2.agg) <- sort(unique(t2.colnames)) #make the colum names = to unique t2 column names and so

#I think this aggregates the t2 columns by name, takes the mean, and then applies it back to the data f

for (i in 1:length (t2[,1]))

{
agg <- aggregate(t2[i,],by=list(t2.colnames),mean)

t2.agg[i,] <- agg[,2]
}
```

You then have to bind this information into a .gct file containing your expression data and the corresponding hgnc gene labels

```
j="All."
t.agg1 <- t(t2.agg)
t.agg2 <- cbind (row.names(t.agg1), rep(NA,length(t.agg1[,1])), t.agg1)
colnames(t.agg2)[1:2] <- c("NAME", "DESCRIPTION")
#t.agg2 <- data.matrix(t.agg2)
#mode(t.agg2) <- "numeric"
write("#1.2", file = paste("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/", j, "CHMP write(paste(dim(t.agg2)[1], dim(t.agg1)[2], sep = "\t"),file = paste("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/", j, "write.table (t.agg2, file= paste("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/", j,</pre>
```

Next you have to make your cls file - this contains meta-data on the phenotype groups you have in your data set.

```
#Create cls file
write(paste("10", "2", "1", sep = " "), file = "/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Set
write(paste("#", "PAT", "CON", sep = " "),file = "/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data S
write(paste("0","0","0","1","1","1","1","1", sep = " "),file = "/Users/clairegreen/Documents/PhD/TD
```

Finally, you load the GSEA script file, and run the analysis

use.fast.enrichment.routine = T

```
GSEA.program.location <- ("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/GSEA-P-R/GSEA.1.0.R
source(GSEA.program.location, verbose=T, max.deparse.length=9999)
# dir.create(paste("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/C9orf72_LCM/GSEA_output"))
GSEA (
                                                                       # Input/Output Files :-----
input.ds = paste("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/All.CHMP2B.gct"), #
input.cls = "/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/CHMP2B_pheno.cls", # Inpu
           "/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/GSEA-P-R/GeneSetDatabases/c5.bp.
                      = paste("/Users/clairegreen/Documents/PhD/TDP-43/TDP-43 Data Sets/CHMP2B/GSEA2/"
output.directory
# Program parameters :-----
                    = "CHMP2B",  # Documentation string used as a prefix to name result files (defa
doc.string
non.interactive.run = F,
                                         # Run in interactive (i.e. R GUI) or batch (R command line) m
reshuffling.type = "sample.labels", # Type of permutation reshuffling: "sample.labels" or "gene.l
                                      # Number of random permutations (default: 1000)
nperm
                      = 1000,
weighted.score.type = 1,
                                        # Enrichment correlation-based weighting: O=no weight (KS), 1
nom.p.val.threshold = -1,
                                        # Significance threshold for nominal p-vals for gene sets (de
                                       # Significance threshold for FWER p-vals for gene sets (defau
fwer.p.val.threshold = -1,
                                       # Significance threshold for FDR q-vals for gene sets (defaul
fdr.q.val.threshold = 0.25,
                      = 20,
                                       # Besides those passing test, number of top scoring gene sets
topgs
adjust.FDR.q.val
                                       # Adjust the FDR q-vals (default: F)
gs.size.threshold.min = 15,
                                       # Minimum size (in genes) for database gene sets to be consid
gs.size.threshold.max = 500,
                                       # Maximum size (in genes) for database gene sets to be consid
                = F,
                                       # Reverse direction of gene list (pos. enrichment becomes neg
reverse.sign
preproc.type
                      = 0,
                                        # Preproc.normalization: O=none, 1=col(z-score)., 2=col(rank)
random.seed
                     = 3338,
                                      # Random number generator seed. (default: 123456)
                     = 0,
                                       # For experts only. Permutation type: 0 = unbalanced, 1 = bal
perm.type
                                       # For experts only. Subsampling fraction. Set to 1.0 (no resa
                      = 1.0,
fraction
                      = F,
                                        # For experts only, Resampling mode (replacement or not repla
replace
                                     # For experts only, save intermediate results (e.g. matrix of # Use original (old) version of GSEA (default: F)
save.intermediate.results = F,
```

Use faster routine to compute enrichment for random permuta