```
rm(list=ls())
######################
rm(list=ls())
load(file ="0-gset GEO.RData")
load(file ="0-NoSe.RData")
load(file ="0-NoMi.RData")
# The negative Fc = expression is higher in Normal, Threshod=1.5
Filter NS <- row.names(subset(NoSe, abs(logFC) > 0.18 & adj.P.Val
0.05))
Filter NM <- row.names(subset(NoMi, abs(logFC) > 0.18 & adj.P.Val
                                                                         <
0.05))
shared =as.data.frame(merge(Filter NM, Filter NS, by=1, sort=F)[,1])
Or shared=intersect(Filter NM, Filter NS)
row.names(shared) <- shared[,1]</pre>
expression= as.data.frame(exprs(gset))
myData Normalized Expr Filt =merge(expression, shared, by = "row.names",
row.names(myData Normalized Expr Filt) = myData Normalized Expr Filt[,1]
rem <- c("Row.names", "merge(Filter NM, Filter NS, by = 1, sort = F)[,</pre>
11")
myData Normalized Expr Filt = myData Normalized Expr Filt[ ,
!(names(myData Normalized Expr Filt) %in% rem)]
dim(myData Normalized Expr Filt)
#####################
############################### Convert Affymetrix ID to gene SYMBOL or ENTREZID or
GENENAME. select(hgu133plus2.db, PROBES, c("SYMBOL", "ENTREZID",
"GENENAME"))
library("hqu133plus2.db")
SYMBOL <- data.frame(SYMBOL=sapply(contents(hgu133plus2SYMBOL), paste,
collapse=", ") )
myData Normalized Expr Filt Symbol <- merge(SYMBOL,
myData Normalized Expr Filt, by.x=0, by.y=0, all.y=T)
row.names(myData Normalized Expr Filt Symbol) <-</pre>
myData Normalized Expr Filt Symbol[,1]
myData Normalized Expr Filt_Symbol <-</pre>
myData Normalized Expr Filt Symbol[,-1]
#####################
```

```
################## Averge genes with several probset by collapseRows of
WGCNA
library (WGCNA)
datET <- myData Normalized Expr Filt Symbol[,-1]</pre>
rowGroup <- myData Normalized Expr Filt Symbol[,1]</pre>
rowID <- rownames(datET)</pre>
collapse.object=collapseRows(datET=datET, rowGroup=rowGroup,
rowID=rowID, method="MaxMean") # method="maxRowVariance" or MaxMean
myData Normalized Expr Filt Symbol Coll=data.frame(
collapse.object$group2row, collapse.object$datETcollapsed)
myData Normalized Expr Filt Symbol Coll=myData Normalized Expr Filt Symbo
1 \text{ Coll}[,-c(1,2)]
######################
########################## WGCNA
allowWGCNAThreads()
suppressMessages(library(cluster))
options(stringsAsFactors = FALSE);
drops=c("GSM1256795", "GSM1256794", "GSM1256792", "GSM1256790", "GSM1256789",
"GSM1256783", "GSM1256765", "GSM1256764", "GSM1256763", "GSM1256762", "GSM1256
761", "GSM1256760", "GSM1256759", "GSM1256758", "GSM1256757", "GSM1256756", "GS
M1256755", "GSM1256754", "GSM1256753", "GSM1256752", "GSM1256751", "GSM1256750
", "GSM1256749", "GSM1256748", "GSM1256747", "GSM1256746", "GSM1256745", "GSM12
56744", "GSM1256743", "GSM1256742", "GSM1256741", "GSM1256740", "GSM1256739", "
GSM1256738", "GSM1256737", "GSM1256736", "GSM1256735.CEL", "GSM1256696", "GSM1
256702")
df =
myData Normalized Expr Filt Symbol Coll[,!(names(myData Normalized Expr F
ilt Symbol Coll) %in% drops)]
datExpr = as.data.frame(t(df))
################################ Outlier detection
A = adjacency(t(datExpr), type = "distance")
k = as.numeric(apply(A, 2, sum)) - 1
Z.k = scale(k)
thresholdZ.k = -2.5 # often -2.5
outlierColor = ifelse(Z.k < thresholdZ.k, "red", "black")</pre>
sampleTree = hclust(as.dist(1 - A), method = "average")
datColors = data.frame(outlierC = outlierColor)
############ Remove outlying samples from expression data
remove.samples = Z.k < thresholdZ.k | is.na(Z.k)</pre>
datExpr = datExpr[!remove.samples, ]
```

```
A = adjacency(t(datExpr), type = "distance")
k = as.numeric(apply(A, 2, sum)) - 1
Z.k = scale(k)
#############
###############################
######################### Divide disease and normal samples
load(file ="4-datExpr RemSam.RData")
# Remove "GSM1256685" "GSM1256686"
datExpr m =
as.data_frame(datExpr[c("GSM1256778","GSM1256776","GSM1256775","GSM125677
4", "GSM1256718", "GSM1256715", "GSM1256713", "GSM1256708", "GSM1256706", "GSM1
256705", "GSM1256700", "GSM1256698", "GSM1256694", "GSM1256693", "GSM1256692",
"GSM1256691", "GSM1256690", "GSM1256689", "GSM1256688", "GSM1256684", "GSM1256
683", "GSM1256682", "GSM1256677", "GSM1256674", "GSM1256672"),])
# Remove "GSM1256678" "GSM1256679"
datExpr s = as.data.frame(datExpr[c(
"GSM1256782", "GSM1256781", "GSM1256780", "GSM1256779", "GSM1256777", "GSM1256
773", "GSM1256719", "GSM1256717", "GSM1256716", "GSM1256714", "GSM1256712", "GS
M1256711", "GSM1256710", "GSM1256709", "GSM1256707", "GSM1256704", "GSM1256703
", "GSM1256701", "GSM1256699", "GSM1256697", "GSM1256695", "GSM1256687", "GSM12
56681", "GSM1256680", "GSM1256676", "GSM1256675", "GSM1256673", "GSM1256671", "
GSM1256670", "GSM1256669", "GSM1256668", "GSM1256667", "GSM1256666", "GSM12566
65", "GSM1256664", "GSM1256663", "GSM1256662", "GSM1256661", "GSM1256660", "GSM
1256659", "GSM1256658", "GSM1256657", "GSM1256656", "GSM1256655", "GSM1256654"
,"GSM1256653"),])
datExpr n =
as.data.frame(datExpr[c("GSM1256800","GSM1256799","GSM1256798","GSM125679
7", "GSM1256796", "GSM1256793", "GSM1256791", "GSM1256788", "GSM1256787", "GSM1
256786", "GSM1256785", "GSM1256784", "GSM1256772", "GSM1256771", "GSM1256770",
"GSM1256769", "GSM1256768", "GSM1256767", "GSM1256766", "GSM1256734", "GSM1256
733", "GSM1256732", "GSM1256731", "GSM1256730", "GSM1256729", "GSM1256728", "GS
M1256727", "GSM1256726", "GSM1256725", "GSM1256724", "GSM1256723", "GSM1256722
", "GSM1256721", "GSM1256720"), 1)
# Check the matrix
multi=list(Data1=list(data=datExpr n),
Data2=list(data=datExpr s), Data3=list(data=datExpr m))
multi g=goodSamplesGenesMS(multi)
datExpr n=datExpr n[,multi g$goodGenes]
goodSamplesGenes(datExpr n)
datExpr s=datExpr s[,multi g$goodGenes]
goodSamplesGenes(datExpr s)
```

```
datExpr m=datExpr m[, multi q$qoodGenes]
goodSamplesGenes(datExpr m)
############ Choose a set of soft-thresholding powers
load(file ="4-datExpr n.RData")
load(file ="4-datExpr s.RData")
load(file ="4-datExpr m.RData")
library(WGCNA)
allowWGCNAThreads()
suppressMessages(library(cluster))
options(stringsAsFactors = FALSE);
powers = c(c(1:10), seq(from = 12, to=30, by=1))
sft = pickSoftThreshold(datExpr n, powerVector = powers,
networkType="signed", corFnc = "bicor", verbose = 5,blockSize=17000)
corFnc = "bicor" Or corFnc = cor, corOptions = list(use = 'p')
sizeGrWindow(9, 5)
                      #c(bottom, left, top, right)
par(mar=c(6,8,4,4))
par(mfrow = c(1,2));
cex1 = 0.9;
plot(sft$fitIndices[,1], -
sign(sft$fitIndices[,3])*sft$fitIndices[,2],xlab="Soft Threshold
(power)", ylab="Scale Free Topology Model Fit, signed R^2", type="n", main =
paste("Scale independence"));
text(sft$fitIndices[,1], -
sign(sft$fitIndices[,3])*sft$fitIndices[,2],labels=powers,cex=cex1,col="r
ed");
abline (h=0.80, col="red")
plot(sft$fitIndices[,1], sft$fitIndices[,5], xlab="Soft Threshold
(power) ", ylab="Mean Connectivity", type="n", main = paste("Mean
connectivity"))
text(sft$fitIndices[,1], sft$fitIndices[,5], labels=powers,
cex=cex1, col="red")
##############
############ Comparison of mean expression level and connectivity
between two data
Data1 mean = as.data.frame((t(datExpr n)[, -c(1:1)]));
rownames(Data1 mean) = names(datExpr n);
Data2 mean = as.data.frame((t(datExpr s)[, -c(1:1)]));
```

```
rownames(Data2 mean) = names(datExpr s);
Data3 mean = as.data.frame((t(datExpr m)[, -c(1:1)]));
rownames(Data3 mean) = names(datExpr m);
par(mfrow=c(1,3))
par(mar=c(6,6,4,4))
                      #c(bottom, left, top, right)
mean.Data1=apply(Data1 mean, 1, mean)
mean.Data2=apply(Data2 mean, 1, mean)
mean.Data3=apply(Data3 mean,1,mean)
#plot the result between Data1 and Data2
verboseScatterplot(mean.Data1, mean.Data3, corFnc = "bicor", xlab="Normal",
ylab="Mild", abline =TRUE, abline.color=2, abline.lty =5)
verboseScatterplot(mean.Data1, mean.Data2, corFnc =
"bicor", xlab="Normal", ylab="Severe", abline
=TRUE, abline.color=2, abline.lty =5)
verboseScatterplot(mean.Data3,mean.Data2,corFnc =
"bicor", xlab="Mild", ylab="Severe", abline =TRUE, abline.color=2, abline.lty
title("Mean expression comparison", outer=TRUE, line = -1)
###########
########### Explores the preservation of connectivity between two data
par(mfrow=c(1,3))
par(mar=c(6,6,4,4)) #c(bottom, left, top, right)
sftDatal=softConnectivity(datExpr n,corFnc = "bicor",type = "signed",
blockSize =17000, minNSamples=5, power=13) # nedd power by
pickSoftThreshold
sftData2=softConnectivity(datExpr s,corFnc = "bicor",type = "signed",
blockSize =17000, minNSamples=5, power=16)
sftData3=softConnectivity(datExpr m,corFnc = "bicor",type = "signed",
blockSize =17000, minNSamples=5, power=30)
#plot the result between Data1 and Data2
verboseScatterplot(sftData1,sftData3,type = "signed", corFnc = "bicor",
blockSize = 15000, xlab="Normal", ylab="Mild", abline
=TRUE, abline.color=2, abline.lty =5)
verboseScatterplot(sftData1,sftData2,type = "signed", corFnc =
"bicor", blockSize = 15000, xlab="Normal", ylab="Severe", abline
=TRUE, abline.color=2, abline.lty =5)
verboseScatterplot(sftData3,sftData2,type = "signed", corFnc =
"bicor", blockSize = 15000, xlab="Mild", ylab="Severe", abline
=TRUE, abline.color=2, abline.lty =5)
```

```
########### Module detection. Power and other things have to be set
net1 = blockwiseModules(datExpr n, power = 13,corType= "bicor",
networkType = "signed", TOMType = "signed", maxBlockSize=17000,
minModuleSize = 30, reassignThreshold = 0, mergeCutHeight = 0.25,
numericLabels = TRUE, pamRespectsDendro = FALSE, saveTOMs = TRUE,
saveTOMFileBase = "5-Data1TOM",nThreads = 6, verbose = 3)
table(net1$colors)
# open a graphics window
sizeGrWindow(12, 9)
# Convert labels to colors for plotting
mergedColors = labels2colors(net1$colors)
# Plot the dendrogram and the module colors underneath
plotDendroAndColors(net1$dendrograms[[1]],
mergedColors[net1$blockGenes[[1]]], "Module colors", dendroLabels =
FALSE, hang = 0.03, addGuide = TRUE, guideHang = 0.05,
                   main="Gene hierarchical clustering dendrogram
(Normal)")
##############
############ Save the modules Data1
probes=names(datExpr n)
moduleLabelsAutomatic1 = net1$colors
moduleColorsAutomatic1 = labels2colors(moduleLabelsAutomatic1)
moduleColorsAutomaticData1=moduleColorsAutomatic1
modules=paste(probes, net1$colors, moduleColorsAutomatic1, sep=",")
write.csv(modules, file = "6-Modules.csv", sep="," ,quote = FALSE,
row.names = FALSE)
###############
########### Normal vs Severe
multiColor=list(Data1=moduleColorsAutomaticData1)
setLabels = c("Data1", "Data2")
multiExpr=list(Data1=list(data=datExpr n), Data2=list(data=datExpr s))
nPermutations1=200
```

```
set.seed(1)
system.time({ mp s = modulePreservation(multiExpr,
multiColor,networkType= "signed",corFnc= "bicor", referenceNetworks = 1,
nPermutations = nPermutations1, randomSeed = 1, quickCor = 0,
maxModuleSize=5000, verbose = 3) })
save(mp s, file = "7-modulePreservation s.RData")
###########
############ Mmodule preservation results  ## Excel
# specify the reference and the test networks
ref=1; test = 2
statsObs= cbind(mp s$quality$observed[[ref]][[test]][,-
1],mp s$preservation$observed[[ref]][[test]][,-1])
statsZ= cbind(mp s$quality$Z[[ref]][[test]][,-
1],mp s$preservation$Z[[ref]][[test]][,-1]);
moduleSize=mp s$preservation$Z[[ref]][[test]]$moduleSize
log.p=mp s$preservation$log.p[[ref]][[test]][,-1]
QualityStats=print(cbind(moduleSize, statsObs[, c("medianRank.pres",
"medianRank.qual")], signif(statsZ[, c("Zsummary.pres",
"Zsummary.qual")],2),signif(log.p[,c("log.psummary.pres")],2)))
#############
############ Mmodule preservation results ## Plot
Obs.PreservationStats= mp s$preservation$observed[[ref]][[test]]
Z.PreservationStats=mp s$preservation$Z[[ref]][[test]]
modColors = rownames(Obs.PreservationStats)
moduleSize = Obs.PreservationStats$moduleSize
selectModules = !(modColors %in% c("grey", "gold"))
point.label = modColors[selectModules]
medianRank=Obs.PreservationStats$medianRank.pres
Zsummary=Z.PreservationStats$Zsummary.pres
par(mfrow=c(1,2), mar = c(4.5, 4.5, 2.5, 1))
plot (moduleSize[selectModules], medianRank[selectModules], col=1,
bg=modColors[selectModules],pch = 21,main="MedianRank Preservation", cex
= 2, ylab ="MedianRank", xlab="Module size", log="x")
labelPoints (moduleSize[selectModules], medianRank[selectModules], point.lab
el,cex=1,offs=0.03)
abline (h=8, col = "red", lty = 2);
```

```
plot (moduleSize[selectModules], Zsummary[selectModules], col = 1,
bq=modColors[selectModules],pch = 21,main="Zsummary preservation",
cex=2,ylab ="Zsummary", xlab = "Module size", log = "x")
labelPoints(moduleSize[selectModules], Zsummary[selectModules], point.label
,cex=1,offs=0.03)
abline (h=5, col = "red", lty = 2)
#############
########### Normal vs Mild
multiColor=list(Data1=moduleColorsAutomaticData1)
setLabels = c("Data1", "Data2")
multiExpr=list(Data1=list(data=datExpr n), Data2=list(data=datExpr m))
nPermutations1=200
set.seed(1)
system.time({ mp m = modulePreservation(multiExpr, multiColor,
networkType= "signed",corFnc= "bicor", referenceNetworks = 1,
nPermutations = nPermutations1, randomSeed = 1, quickCor =
0, maxModuleSize=5000, verbose = 3 ) })
save(mp m, file = "9-modulePreservation m.RData")
#############
############ Mmodule preservation results  ## Excel
ref=1; test = 2
statsObs= cbind(mp m$quality$observed[[ref]][[test]][,-
1],mp m$preservation$observed[[ref]][[test]][,-1])
statsZ= cbind(mp m$quality$Z[[ref]][[test]][,-
1],mp m$preservation$Z[[ref]][[test]][,-1]);
moduleSize=mp m$preservation$Z[[ref]][[test]]$moduleSize
log.p=mp m$preservation$log.p[[ref]][[test]][,-1]
QualityStats=print(cbind(moduleSize, statsObs[, c("medianRank.pres",
"medianRank.qual")], signif(statsZ[, c("Zsummary.pres",
"Zsummary.qual")],2),signif(log.p[,c("log.psummary.pres")],2)))
#############
############ Mmodule preservation results ## Plot
Obs.PreservationStats= mp m$preservation$observed[[ref]][[test]]
Z.PreservationStats=mp m$preservation$Z[[ref]][[test]]
modColors = rownames(Obs.PreservationStats)
moduleSize = Obs.PreservationStats$moduleSize
```

```
selectModules = !(modColors %in% c("grey", "gold"))
point.label = modColors[selectModules]
medianRank=Obs.PreservationStats$medianRank.pres
Zsummary=Z.PreservationStats$Zsummary.pres
par(mfrow=c(1,2), mar = c(4.5, 4.5, 2.5, 1))
plot (moduleSize[selectModules], medianRank[selectModules], col=1,
bg=modColors[selectModules],pch = 21,main="MedianRank Preservation", cex
= 2, ylab ="MedianRank", xlab="Module size", log="x")
labelPoints (moduleSize[selectModules], medianRank[selectModules], point.lab
el, cex=1, offs=0.03)
abline (h=8, col = "red", lty = 2);
plot (moduleSize[selectModules], Zsummary[selectModules], col = 1,
bg=modColors[selectModules],pch = 21,main="Zsummary Preservation",
cex=2,ylab ="Zsummary", xlab = "Module size", log = "x")
labelPoints(moduleSize[selectModules], Zsummary[selectModules], point.label
,cex=1,offs=0.03)
abline (h=5, col = "red", lty = 2)
###############
########### Module membership analysis, kME
moduleLabelsAutomatic1 = net1$colors
moduleColorsAutomatic1 = labels2colors(moduleLabelsAutomatic1)
moduleColorsAutomaticData1=moduleColorsAutomatic1
load(file ="4-datExpr n.RData")
load(file ="4-datExpr s.RData")
load(file ="4-datExpr_m.RData")
ME.Data1=moduleEigengenes(datExpr n, moduleColorsAutomaticData1) $eigengene
ME.Data2=moduleEigengenes(datExpr s, moduleColorsAutomaticData1) $eigengene
ME.Data3=moduleEigengenes(datExpr m, moduleColorsAutomaticData1) $eigengene
kME Normal=signedKME(datExpr n, ME.Data1, corFnc = "bicor")
kME Severe=signedKME(datExpr s, ME.Data2, corFnc = "bicor")
kME Mild=signedKME(datExpr m, ME.Data3, corFnc = "bicor")
genename = names(datExpr n)
```

```
kME red=as.data.frame(cbind(kME Normal$kMEred,kME Severe$kMEred,kME Mild$
kMEred))
colnames(kME red) = c("kME Normal", "kME Severe", "kME Mild")
row.names(kME red) = names(datExpr n)
inModule = (moduleColorsAutomaticData1=="red")
kME red=kME red[inModule,]
# Plot between different treatments
par(mfrow=c(5,6))
                       #c(bottom, left, top, right)
par(mar=c(2,4,3,2))
verboseScatterplot(kME red$kME Normal, kME red$kME Severe,corFnc =
"cor",xlab="Normal", ylab="Severe", abline
=TRUE, abline.color=2, abline.lty =5, font=1, font.lab=1, font.main=1)
verboseScatterplot(kME red$kME Normal, kME red$kME Mild,corFnc =
"cor", xlab="Normal", ylab="Mild", abline =TRUE, abline.color=2, abline.lty
=5, font=1, font.lab=1, font.main=1)
verboseScatterplot(kME red$kME Severe, kME red$kME Mild,corFnc =
"cor", xlab="Mild", ylab="Severe", abline =TRUE, abline.color=2, abline.lty
=5, font=1, font.lab=1, font.main=1)
title("Module membership for turquoise module", outer=TRUE, line = -1)
par (mfrow=c(1,1), mar=c(3.5, 3.5, 2, 1), mgp=c(2.4, 0.8, 0), las=1)
ME pink=as.data.frame(cbind(ME.Data1$MEpink,ME.Data2$MEpink,ME.Data3$MEpi
colnames (ME pink) = c("ME Normal", "ME Severe", "ME Mild")
##############
############ Intramodular connectivity (kIM)
kIM Normal=intramodularConnectivity.fromExpr(datExpr n,
moduleColorsAutomaticData1,corFnc = "bicor",networkType = "signed",
scaleByMax=TRUE, power=13) $kWithin
kIM Severe=intramodularConnectivity.fromExpr(datExpr s,
moduleColorsAutomaticData1,corFnc = "bicor",networkType = "signed",
scaleByMax=TRUE, power=16) $kWithin
kIM Mild=intramodularConnectivity.fromExpr(datExpr m,
moduleColorsAutomaticData1,corFnc = "bicor",networkType = "signed",
scaleByMax=TRUE, power=30) $kWithin
kIM=as.data.frame(cbind(kIM Normal, kIM Severe, kIM Mild))
```

```
colnames(kIM) = c("kIM Normal", "kIM Severe", "kIM Mild")
row.names(kIM) = names(datExpr n)
inModule = (moduleColorsAutomaticData1=="yellow")
kIM yellow=kIM[inModule,]
write.table(kIM yellow, "kIM yellow.txt")
par(mfrow=c(1,3))
par(mar=c(8,4,8,4))
                      #c(bottom, left, top, right)
verboseScatterplot(kIM green$kIM Normal,kIM green$kIM Severe,corFnc =
"bicor", xlab="Normal", ylab="Severe", abline
=TRUE, abline.color=2, abline.lty =5)
verboseScatterplot(kIM green$kIM Normal,kIM green$kIM Mild,corFnc =
"bicor", xlab="Normal", ylab="Mild", abline
=TRUE, abline.color=2, abline.lty =5)
verboseScatterplot(kIM green$kIM Mild,kIM green$kIM Severe,corFnc =
"bicor", xlab="Mild", ylab="Severe", abline
=TRUE, abline.color=2, abline.lty =5)
title("Intramodular connectivity for green module", outer=TRUE, line = -1)
```

##############