# Cross-platform integration and Differential Expression analysis of T2DM and HTN transcriptomic profiles

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```
library(GEOquery)
## Loading required package: Biobase
## Loading required package: BiocGenerics
##
## Attaching package: 'BiocGenerics'
## The following objects are masked from 'package:stats':
##
       IQR, mad, sd, var, xtabs
##
## The following objects are masked from 'package:base':
##
##
       anyDuplicated, aperm, append, as.data.frame, basename, cbind,
##
       colnames, dirname, do.call, duplicated, eval, evalq, Filter, Find,
       get, grep, grepl, intersect, is.unsorted, lapply, Map, mapply,
##
       match, mget, order, paste, pmax, pmax.int, pmin, pmin.int,
       Position, rank, rbind, Reduce, rownames, sapply, setdiff, sort,
##
       table, tapply, union, unique, unsplit, which.max, which.min
##
## Welcome to Bioconductor
##
##
       Vignettes contain introductory material; view with
##
       'browseVignettes()'. To cite Bioconductor, see
       'citation("Biobase")', and for packages 'citation("pkgname")'.
## Setting options('download.file.method.GEOquery'='auto')
## Setting options('GEOquery.inmemory.gpl'=FALSE)
library(affy)
library(limma)
##
## Attaching package: 'limma'
```

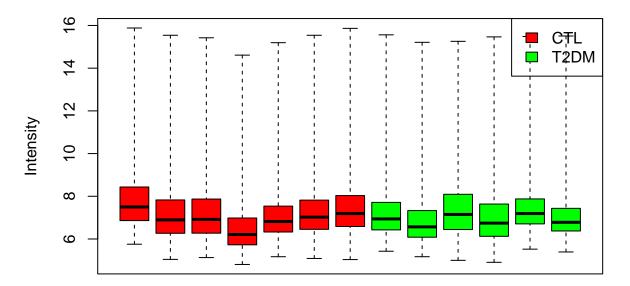
```
## The following object is masked from 'package:BiocGenerics':
##
##
       plotMA
library(sva)
## Loading required package: mgcv
## Loading required package: nlme
## This is mgcv 1.9-1. For overview type 'help("mgcv-package")'.
## Loading required package: genefilter
## Loading required package: BiocParallel
library(WGCNA)
## Loading required package: dynamicTreeCut
## Loading required package: fastcluster
## Attaching package: 'fastcluster'
## The following object is masked from 'package:stats':
##
##
       hclust
##
##
## Attaching package: 'WGCNA'
## The following object is masked from 'package:stats':
##
##
       cor
library(ensembldb)
## Loading required package: GenomicRanges
## Loading required package: stats4
## Loading required package: S4Vectors
##
## Attaching package: 'S4Vectors'
```

```
## The following object is masked from 'package:utils':
##
       findMatches
##
## The following objects are masked from 'package:base':
##
##
       expand.grid, I, unname
## Loading required package: IRanges
## Attaching package: 'IRanges'
## The following object is masked from 'package:nlme':
##
##
       collapse
## Loading required package: GenomeInfoDb
## Loading required package: GenomicFeatures
## Loading required package: AnnotationDbi
## Loading required package: AnnotationFilter
##
## Attaching package: 'ensembldb'
## The following object is masked from 'package:stats':
##
       filter
library(biomaRt)
# Download GSE25724 data (T2DM)
gse_T2DM <- getGEO("GSE25724", GSEMatrix =TRUE,getGPL=FALSE)</pre>
## Found 1 file(s)
## GSE25724_series_matrix.txt.gz
datMeta_T2DM <- pData(gse_T2DM[[1]])</pre>
rownames(datMeta_T2DM) <- datMeta_T2DM$geo_accession</pre>
# Read GSE25724 data
setwd("/Users/lorandacalderonzamora/GSE25724/")
data.affy_T2DM <- ReadAffy(celfile.path = "./")</pre>
datExpr_T2DM <- exprs(data.affy_T2DM)</pre>
```

```
# Align datMeta_T2DM and datExpr_T2DM by sample identifiers
GSM_T2DM <- rownames(pData(data.affy_T2DM))</pre>
GSM_T2DM <- substr(GSM_T2DM,1,9)</pre>
idx <- match(GSM_T2DM, datMeta_T2DM$geo_accession)</pre>
datMeta_T2DM <- datMeta_T2DM[idx,]</pre>
colnames(datExpr_T2DM)=rownames(datMeta_T2DM)
\# Cleaning and formatting of GSE25724 metadata
datMeta_T2DM <- datMeta_T2DM[,-c(3:7,14:36)]</pre>
colnames(datMeta_T2DM)[2] <- c("Dx")</pre>
datMeta_T2DM$Dx[rownames(datMeta_T2DM) %in% c("GSM631755", "GSM631756", "GSM631757", "GSM631758", "GSM6
datMeta_T2DM$Dx[rownames(datMeta_T2DM) %in% c("GSM631762", "GSM631763", "GSM631764", "GSM631765", "GSM6
datMeta_T2DM$Dx <- as.factor(datMeta_T2DM$Dx)</pre>
# Preprocessing and quality assessment of GSE25724 raw expression data
datExpr_T2DM <- log2(datExpr_T2DM)</pre>
dim(datExpr_T2DM)
## [1] 506944
```

```
# Exploratory visualization of GSE25724 raw data
boxplot(datExpr_T2DM,range=0, col=c('red', 'green')[as.numeric(datMeta_T2DM$Dx)], xaxt='n', xlab = "Arr
legend("topright",legend = levels(datMeta_T2DM$Dx),fill = c('red', 'green')[as.numeric(as.factor(levels))]
```

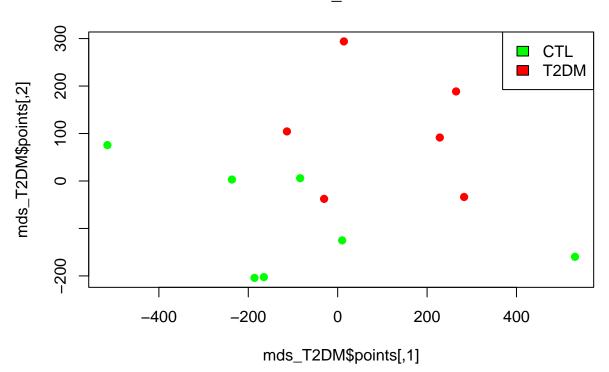
# **Boxplot**



Array

```
mds_T2DM = cmdscale(dist(t(datExpr_T2DM)),eig=TRUE)
plot(mds_T2DM$points,col=c('green', 'red')[as.numeric(datMeta_T2DM$Dx)],pch=19,main="MDS_T2DM")
legend("topright",legend = levels(datMeta_T2DM$Dx),fill =c('green', 'red')[as.numeric(as.factor(levels())]
```

## MDS\_T2DM



```
# Normalization using RMA
datExpr_T2DM <- rma(data.affy_T2DM, background=T, normalize=T, verbose=T)

## Warning: replacing previous import 'AnnotationDbi::tail' by 'utils::tail' when
## loading 'hgu133acdf'

## Warning: replacing previous import 'AnnotationDbi::head' by 'utils::head' when
## loading 'hgu133acdf'

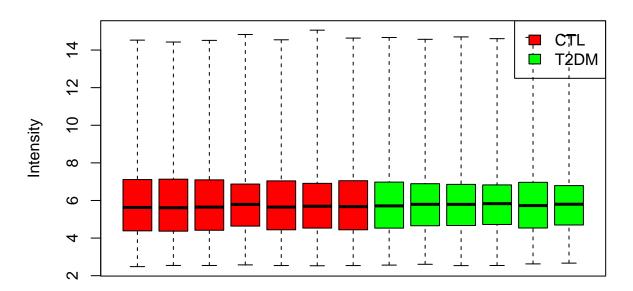
##

## Background correcting
## Normalizing
## Calculating Expression

datExpr_T2DM <- exprs(datExpr_T2DM)</pre>
```

```
# Exploratory visualization of GSE25724 normalized data
boxplot(datExpr_T2DM,range=0, col=c('red', 'green')[as.numeric(datMeta_T2DM$Dx)], xaxt='n', xlab = "Arr
legend("topright",legend = levels(datMeta_T2DM$Dx),fill = c('red', 'green')[as.numeric(as.factor(levels))]
```

# **Boxplot**



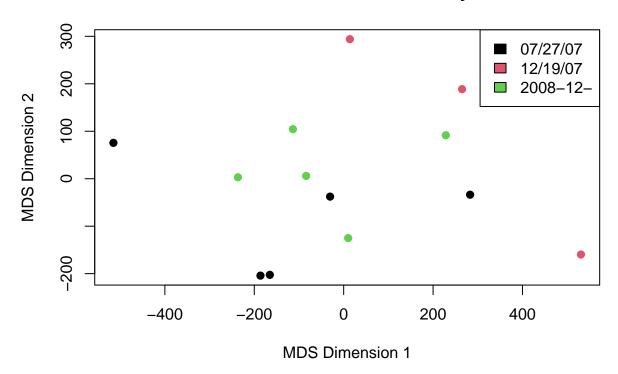
#### Array

```
# Extract ScanDate from GSE25724 for batch effect correction
batch_T2DM <- protocolData(data.affy_T2DM)$ScanDate
batch_T2DM <- substr(batch_T2DM,1,8)
batch_T2DM <- as.factor(batch_T2DM)
table(batch_T2DM)

## batch_T2DM
## 07/27/07 12/19/07 2008-12-
## 5 3 5</pre>
datMeta_T2DM$Batch <- batch_T2DM
```

# Visualization of ScanDate metadata from GSE25724 to identify potential batch effects
plot(mds\_T2DM\$points,col = as.numeric(datMeta\_T2DM\$Batch),pch=19,main="MDS Plot of GSE25724 Colored by legend("topright",legend = levels(datMeta\_T2DM\$Batch),fill = as.numeric(as.factor(levels(datMeta\_T2DM\$Batch))

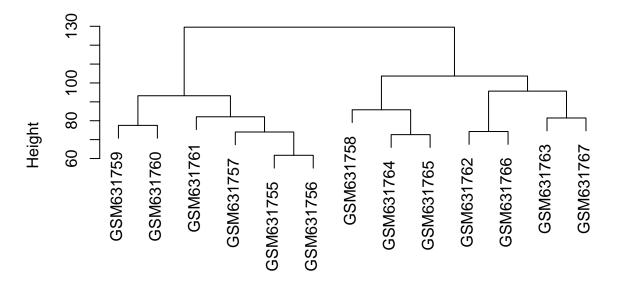
## MDS Plot of GSE25724 Colored by Batch



```
# Create ExpressionSet object after Batch effect assessment
datMeta_T2DM$Batch <- batch_T2DM
datMeta_proc_T2DM <- new("AnnotatedDataFrame", data = datMeta_T2DM)
colnames(datExpr_T2DM) <- rownames(datMeta_T2DM)
datAll_T2DM <- new("ExpressionSet", exprs = datExpr_T2DM, phenoData = datMeta_proc_T2DM)
# No singular batch was detected in the GSE25724 dataset.
# Therefore, batch correction with ComBat is technically feasible.
# However, as no evident batch effect was observed in exploratory analyses (MDS),
# ComBat was not applied, and no batch removal was necessary.</pre>
```

```
# Sample Clustering and outlier detection
tree_T2DM <- hclust(dist(t(exprs(datAll_T2DM))), method = "average")
plot(tree_T2DM, main = "Hierarchical clustering of GSE25724 samples", xlab = "", sub = "")</pre>
```

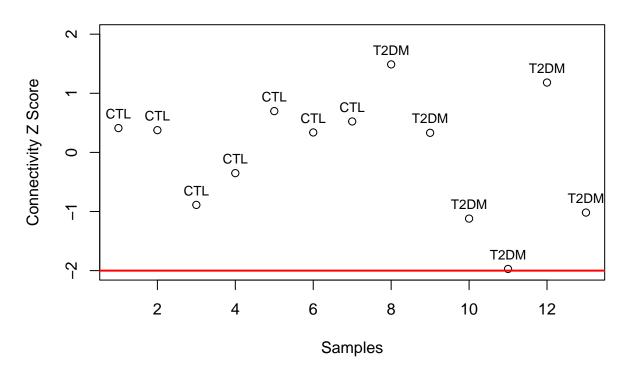
# **Hierarchical clustering of GSE25724 samples**



```
normadj_T2DM <- (0.5 + 0.5*bicor(exprs(datAll_T2DM)))^2
netsummary_T2DM <- fundamentalNetworkConcepts(normadj_T2DM)
C_T2DM <- netsummary_T2DM$Connectivity
Z.C_T2DM <- (C_T2DM - mean(C_T2DM)) / sqrt(var(C_T2DM))

datLabel_T2DM <- pData(datAll_T2DM)$Dx
plot(1:length(Z.C_T2DM),Z.C_T2DM,main="Outlier plot of GSE25724 samples ",xlab = "Samples",ylab="Connectext(1:length(Z.C_T2DM),Z.C_T2DM,label=datLabel_T2DM,pos=3,cex=0.8)
abline(h= -2, col="red", lwd = 2)</pre>
```

## **Outlier plot of GSE25724 samples**

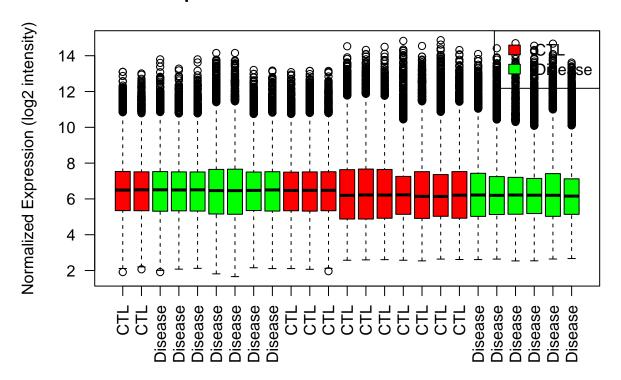


```
# Identify and remove potential outlier from GSE25724 samples based on connectivity Z-score
# No samples exceeded the threshold (Z < -2), so none were removed
to_keep_T2DM <- abs(Z.C_T2DM) < 2</pre>
table(to_keep_T2DM)
## to_keep_T2DM
## TRUE
##
     13
colnames(exprs(datAll_T2DM))[!to_keep_T2DM]
## character(0)
datAll_T2DM <- datAll_T2DM[, to_keep_T2DM]</pre>
# Annotating Probes using Ensembl
ensembl <- useEnsembl(biomart = "ensembl", dataset = "hsapiens_gene_ensembl")</pre>
# Annotating Probes for GSE25724 dataset
identifier <- "affy_hg_u133a_2"</pre>
getinfo <- c("affy_hg_u133a_2", "ensembl_gene_id", "entrezgene_id", "external_gene_name")</pre>
geneDat_T2DM <- getBM(attributes = getinfo,</pre>
                       filters = identifier,
```

```
values = rownames(exprs(datAll_T2DM)),
                       mart = ensembl)
idx_T2DM <- match(rownames(exprs(datA11_T2DM)), geneDat_T2DM$affy_hg_u133a_2)
geneDat_T2DM <- geneDat_T2DM[idx_T2DM, ]</pre>
table(is.na(geneDat_T2DM$ensembl_gene_id))
##
## FALSE TRUE
## 20259 2024
to_keep_T2DM <- !is.na(geneDat_T2DM$ensembl_gene_id)</pre>
geneDat_T2DM <- geneDat_T2DM[to_keep_T2DM, ]</pre>
datAll_T2DM <- datAll_T2DM[to_keep_T2DM, ]</pre>
# Collapse Rows for GSE25724 by Ensembl Gene ID
table(duplicated(geneDat T2DM$affy hg u133a 2))
##
## FALSE
## 20259
table(duplicated(geneDat T2DM$ensembl gene id))
##
## FALSE TRUE
## 13366 6893
CR_T2DM <- collapseRows(exprs(datAll_T2DM),</pre>
                         rowGroup = geneDat_T2DM$ensembl_gene_id,
                         rowID = geneDat_T2DM$affy_hg_u133a_2)
CRdata_T2DM <- CR_T2DM$datETcollapsed
idx_T2DM <- match(CR_T2DM$group2row[,"selectedRowID"], geneDat_T2DM$affy_hg_u133a_2)
geneDat_T2DM <- geneDat_T2DM[idx_T2DM, ]</pre>
rownames(geneDat_T2DM) <- geneDat_T2DM$ensembl_gene_id</pre>
# Differential Expression Analysis from GSE25724
mod_T2DM <- model.matrix(~pData(datAll_T2DM)$Dx)</pre>
fit_T2DM <- lmFit(CR_T2DM$datETcollapsed,mod_T2DM)</pre>
fit_T2DM <- eBayes(fit_T2DM)</pre>
tt_T2DM <- topTable(fit_T2DM,coef = 2,n = Inf,genelist = geneDat_T2DM)</pre>
head(tt_T2DM)
##
                   affy_hg_u133a_2 ensembl_gene_id entrezgene_id
## ENSG0000147642
                          218692_at ENSG00000147642
                                                             55638
## ENSG0000171109
                        207098 s at ENSG00000171109
                                                             55669
## ENSG00000156413
                        211465_x_at ENSG00000156413
                                                              2528
## ENSG0000143575
                          201145 at ENSG00000143575
                                                             10456
## ENSG0000187735
                        216241_s_at ENSG00000187735
                                                             6917
## ENSG00000086619
                          220012_at ENSG00000086619
                                                             56605
                                            logFC AveExpr
##
                   external_gene_name
                                                                            P. Value
                                                                   t
```

```
SYBU -1.7852973 7.456118 -8.898821 3.975337e-07
## ENSG0000147642
## ENSG0000171109
                                 MFN1 -2.0197137 5.854147 -8.708803 5.146010e-07
## ENSG00000156413
                                 FUT6 0.9909830 8.111291 7.684775 2.221989e-06
## ENSG0000143575
                                 HAX1 -0.9649762 8.280549 -7.464098 3.096722e-06
## ENSG0000187735
                                 TCEA1 -1.9923078 8.559339 -7.444819 3.188778e-06
## ENSG00000086619
                                 ER01B -2.4723116 7.598684 -7.394610 3.442341e-06
                     adj.P.Val
## ENSG00000147642 0.003439078 6.555426
## ENSG00000171109 0.003439078 6.335236
## ENSG00000156413 0.007096584 5.061263
## ENSG00000143575 0.007096584 4.766488
## ENSG00000187735 0.007096584 4.740382
## ENSG00000086619 0.007096584 4.672123
# Load a Normalized_expression_matrix_from_HTN csv file
data_NEM_HTN <- read.csv("/Users/lorandacalderonzamora/Downloads/combat_expr", row.names = 1)</pre>
# Preprocessing sample identifiers and group assignment
colnames(data_NEM_HTN) <- gsub("\\..*", "", colnames(data_NEM_HTN))</pre>
group <- factor(colnames(data NEM HTN))</pre>
# Merging T2DM and HTN Expression matrices by common genes
common_genes <- intersect(rownames(data_NEM_HTN), rownames(CRdata_T2DM))</pre>
data_NEM_HTN_common <- data_NEM_HTN[common_genes, ]</pre>
CRdata_T2DM_common <- CRdata_T2DM[common_genes, ]</pre>
unificated expr matrix <- cbind(data NEM HTN common, CRdata T2DM common)
unificated_expr_matrix <- as.data.frame(unificated_expr_matrix)</pre>
# Relabeling sample identifiers with group labels for Differential Expression analysis
sample_ids <- colnames(unificated_expr_matrix)</pre>
group_labels <- sample_ids</pre>
group_labels [group_labels %in% c("GSM631755", "GSM631756", "GSM631757", "GSM631758", "GSM631759", "GSM6
                                  "GSM609530")] <- "CTL"
group_labels [group_labels %in% c("GSM631762", "GSM631763", "GSM631764", "GSM631765", "GSM631766", "GSM6
                                  "GSM609526", "GSM609527")] <- "Disease"
colnames(unificated_expr_matrix) <- group_labels</pre>
# Boxplot of the merged Expression matrix from T2DM and HTN prior to Batch correction
boxplot(unificated_expr_matrix, main = "Gene Expression distribution before Batch correction", col = c(
legend("topright", legend = levels(factor(group_labels)), fill = c("red", "green"))
```

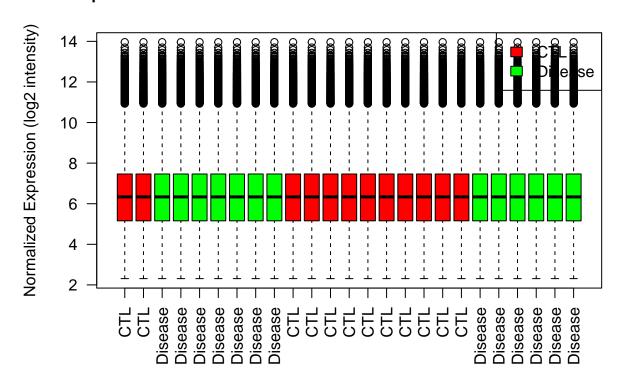
### **Gene Expression distribution before Batch correction**



```
# Quantile normalization of merged Expression matrix
library(preprocessCore)
expr_matrix_qn <- normalize.quantiles(as.matrix(unificated_expr_matrix))</pre>
rownames(expr_matrix_qn) <- rownames(unificated_expr_matrix)</pre>
colnames(expr_matrix_qn) <- colnames(unificated_expr_matrix)</pre>
# Batch effect correction between T2DM and HTN Datasets using ComBat
n_HTN <- ncol(data_NEM_HTN_common)</pre>
n_T2DM <- ncol(CRdata_T2DM_common)</pre>
batch <- c(rep("HTN", n_HTN), rep("T2DM", n_T2DM))</pre>
combat_expr <- ComBat(dat = expr_matrix_qn, batch = batch, par.prior = TRUE, prior.plots = FALSE)</pre>
## Found2batches
## Adjusting forOcovariate(s) or covariate level(s)
## Standardizing Data across genes
## Fitting L/S model and finding priors
## Finding parametric adjustments
## Adjusting the Data
```

```
# Boxplot of the merged Expression matrix from T2DM and HTN after to Batch correction
boxplot(expr_matrix_qn, main = "Gene Expression distribution unificated matriz T2DM and HTN after Batch
legend("topright", legend = levels(factor(group_labels)), fill = c("red", "green"))
```

#### Gene Expression distribution unificated matriz T2DM and HTN after Batch correc



```
# Define group factor for Differential Expression analysis
sample_ids <- colnames(combat_expr)
group <- as.factor(colnames(combat_expr))</pre>
```

# Differential Expression Analysis between Control and Disease samples from integrated T2DM and HTN Dat
design <- model.matrix(~ group)
fit\_merge <- lmFit(combat\_expr, design)
fit\_merge <- eBayes(fit\_merge)
tt\_merge <- topTable(fit\_merge, coef = 2, number = Inf)
head(tt\_merge)</pre>

```
## ENSG0000133226 -0.9124525 7.509201 -5.620966 6.025814e-06 0.04549518 3.822115
## ENSG0000018495 -1.0503591 5.976003 -5.123157 2.263210e-05 0.04549518 2.658900
## ENSG0000167633 0.4355697 4.519460 5.080068 2.539053e-05 0.04549518 2.557451
## ENSG00000134884 -0.8300671 7.907104 -5.022713 2.959263e-05 0.04549518 2.422281
## ENSG00000178078 0.7516481 7.001916 4.990679 3.223628e-05 0.04549518 2.338471
## ENSG00000126804 -1.0117875 6.690346 -4.987182 3.253889e-05 0.04549518 2.338471
```

```
ensembl gene id external gene name entrezgene id
##
                                                            logFC AveExpr
## 1 ENSG00000000003
                                TSPAN6
                                                7105 0.06385849 6.545041
## 2 ENSG0000000005
                                   TNMD
                                                64102 0.18128788 3.845240
## 3 ENSG00000000419
                                   DPM1
                                                 8813 -0.69726956 8.056324
## 4 ENSG0000000457
                                  SCYL3
                                                57147 -0.16583302 6.040446
## 5 ENSG0000000460
                                  FIRRM
                                                55732 0.21194351 3.730130
## 6 ENSG0000000938
                                                 2268 0.00536728 6.778804
                                    FGR
##
               t
                     P. Value adj. P. Val
                                                В
## 1 0.17129185 0.865287307 0.93522714 -5.946200
## 2 1.96263019 0.060199360 0.23944823 -4.223632
## 3 -3.09296738 0.004608104 0.08138296 -2.039369
## 4 -1.24559146 0.223746720 0.47454200 -5.230485
## 5 2.20787740 0.036043337 0.18504539 -3.801831
## 6 0.01988833 0.984280506 0.99344443 -5.960229
```