

# Shank3 sleep deprivation differential expression

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## Description

This is the report of the analysis made for the paper *Shank3 Modulates Sleep and Circadian Activity* by Ashley M. Ingiosi, Taylor Wintler, Hannah Schoch, Kristan G. Singletary, Dario Righelli, Leandro G. Roser, Davide Rizzo, Marcos G. Frank and Lucia Peixoto.

Autism Spectrum Disorder (ASD) is the most prevalent neurodevelopmental disorder in the US that often co-presents with sleep problems. Sleep impairments in ASD predict the severity of ASD core diagnostic symptoms and have a considerable impact on the quality of life of caregivers. However, little is known about the underlying molecular mechanism(s) of sleep impairments in ASD. In this study we investigated the role of Shank3, a high confidence ASD gene candidate, in the regulation of sleep. We show that Shank3 mutant mice have problems falling asleep despite accumulating sleep pressure. Using RNA-seq we show that sleep deprivation doubles the differences in gene expression between mutants and wild types and downregulates circadian transcription factors Per3, Dec2, and Rev-erba. Shank3 mutants also have trouble regulating

locomotor activity in the absence of light. Overall, our study shows that Shank3 is an important modulator of sleep and circadian activity.

## Importing data

Importing data and filtering out those genes with cpm lesser than 1. We use the *filtered.data* method in *NOISeq* package.

```
countMatrix <- ReadDataFrameFromTsv(file.name.path="./data/refSEQ_countMatrix.txt")

## ./data/refSEQ_countMatrix.txt read from disk!
# head(countMatrix)

designMatrix <- ReadDataFrameFromTsv(file.name.path="./design/all_samples_short_names.tsv")

## ./design/all_samples_short_names.tsv read from disk!
# head(designMatrix)

filteredCountsProp <- filterLowCounts(counts.dataframe=countMatrix,
                                     is.normalized=FALSE,
                                     design.dataframe=designMatrix,
                                     cond.col.name="gcondition",
                                     method.type="Proportion")

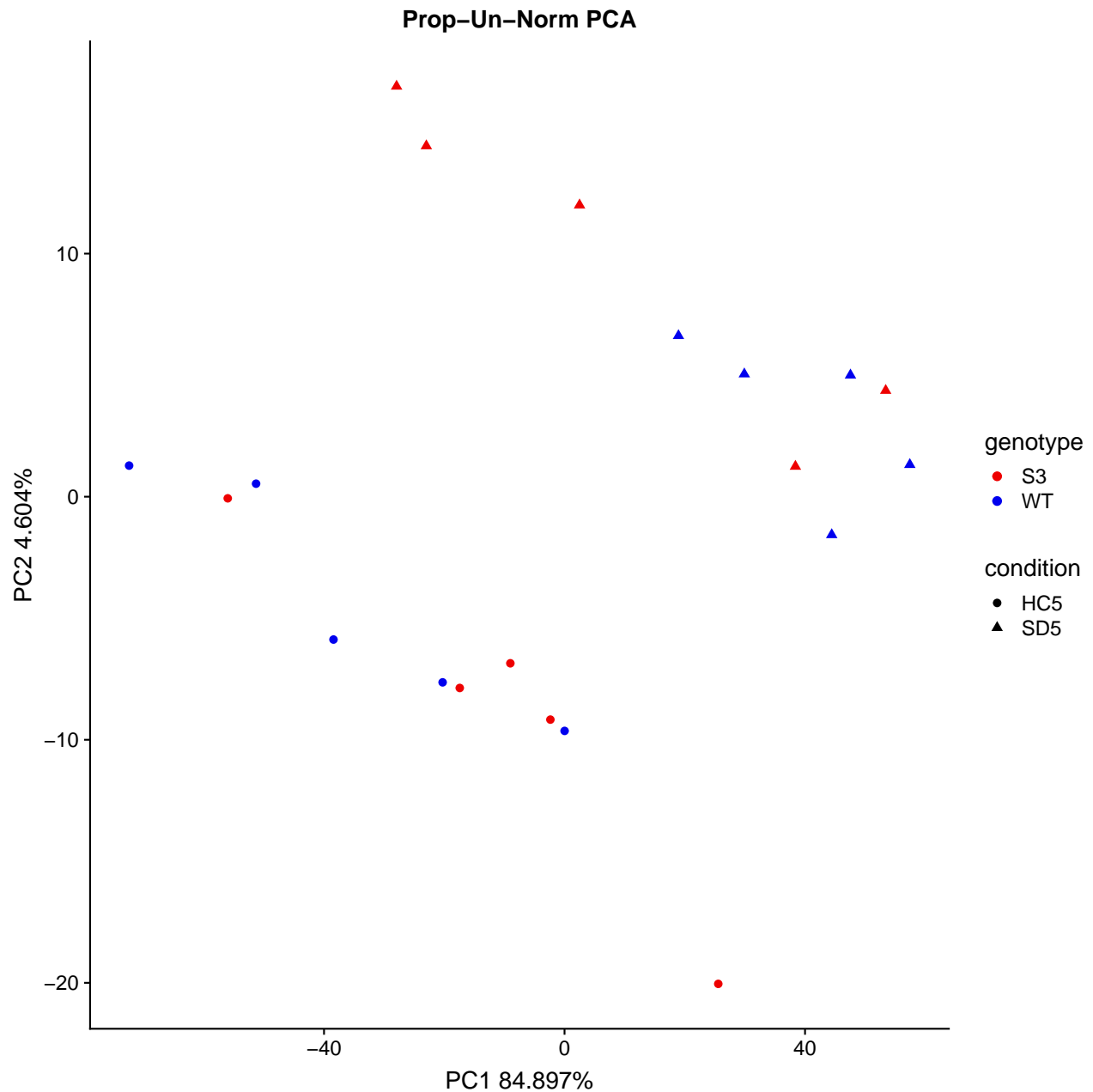
## features dimensions before normalization: 27179
## Filtering out low count features...
## 14454 features are to be kept for differential expression analysis with filtering method 3
```

## Plot PCA of log unnormalized data

PCA Plot of filtered not-normalized data.

```
PlotPCAPlotlyFunction(counts.data.frame=log1p(filteredCountsProp),
                      design.matrix=designMatrix,
                      shape.colname="condition", color.colname="genotype", xPCA="PC1", yPCA="PC2",
                      plotly.flag=FALSE, show.plot.flag=TRUE, prefix.plot="Prop-Un-Norm")

## [1] FALSE
```



## Control Genes

### Negative control genes

Loading Negative Control Genes to normalize data

```
library(readxl)
```

```
sd.ctrls <- read_excel(path="./data/controls/Additional File 4 full list of BMC genomics SD&RS2.xlsx", sheet="SD5")
sd.ctrls <- sd.ctrls[order(sd.ctrls$adj.P.Val),]
```

```
sd.neg.ctrls <- sd.ctrls[sd.ctrls$adj.P.Val > 0.9, ]
```

```

sd.neg.ctrls <- sd.neg.ctrls$`MGI Symbol`
sd.neg.ctrls <- sd.neg.ctrls[!which(is.na(sd.neg.ctrls))]

int.neg.ctrls <- sd.neg.ctrls
int.neg.ctrls <- unique(int.neg.ctrls)
neg.map <- convertGenesViaMouseDb(gene.list=int.neg.ctrls, fromType="SYMBOL",
                                "ENTREZID")
# sum(is.na(neg.map$ENTREZID))
neg.ctrls.entrez <- as.character(neg.map$ENTREZID)

ind.ctrls <- which(rownames(filteredCountsProp) %in% neg.ctrls.entrez)
counts.neg.ctrls <- filteredCountsProp[ind.ctrls,]

```

## positive control genes

Loading Positive Control Genes to detect them during the differential expression step.

```

## sleep deprivation
sd.lit.pos.ctrls <- read_excel("./data/controls/SD_RS_PosControls_final.xlsx",
                              sheet=1)
colnames(sd.lit.pos.ctrls) <- sd.lit.pos.ctrls[1,]
sd.lit.pos.ctrls <- sd.lit.pos.ctrls[-1,]

sd.est.pos.ctrls <- read_excel("./data/controls/SD_RS_PosControls_final.xlsx",
                              sheet=3)

sd.pos.ctrls <- cbind(sd.est.pos.ctrls$`MGI Symbol`, "est")
sd.pos.ctrls <- rbind(sd.pos.ctrls, cbind(sd.lit.pos.ctrls$Gene, "lit"))

sd.pos.ctrls <- sd.pos.ctrls[!which(duplicated(sd.pos.ctrls[,1])),]
sd.pos.ctrls <- sd.pos.ctrls[!which(is.na(sd.pos.ctrls[,1])),]

```

## Normalizations

### TMM Normalization

Normalizing data with *TMM*, as implemented in *edgeR* package, and plotting a PCA and an RLE plot of them.

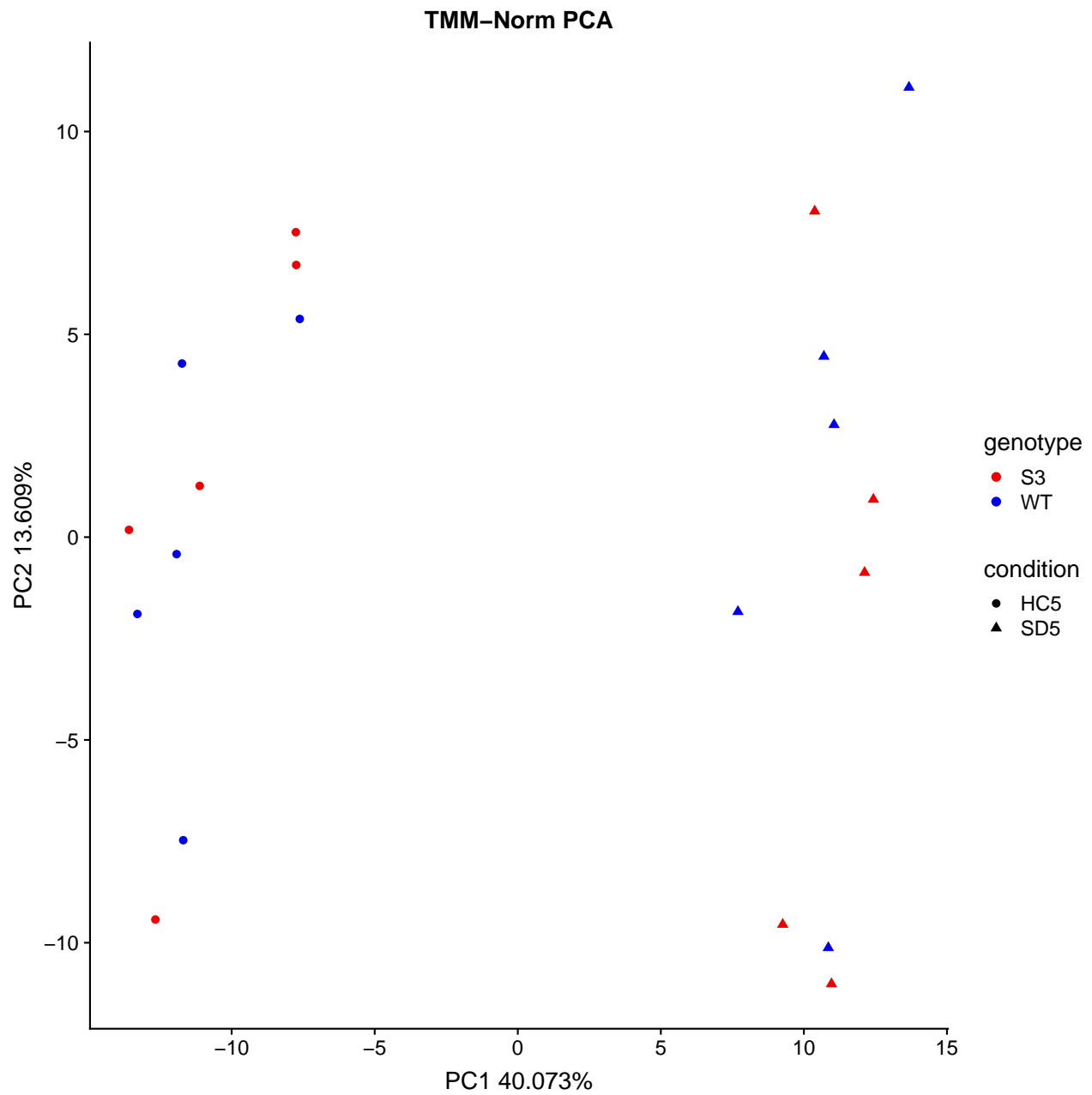
```

normPropCountsUqua <- NormalizeData(data.to.normalize=filteredCountsProp,
                                   norm.type="tmm",
                                   design.matrix=designMatrix)

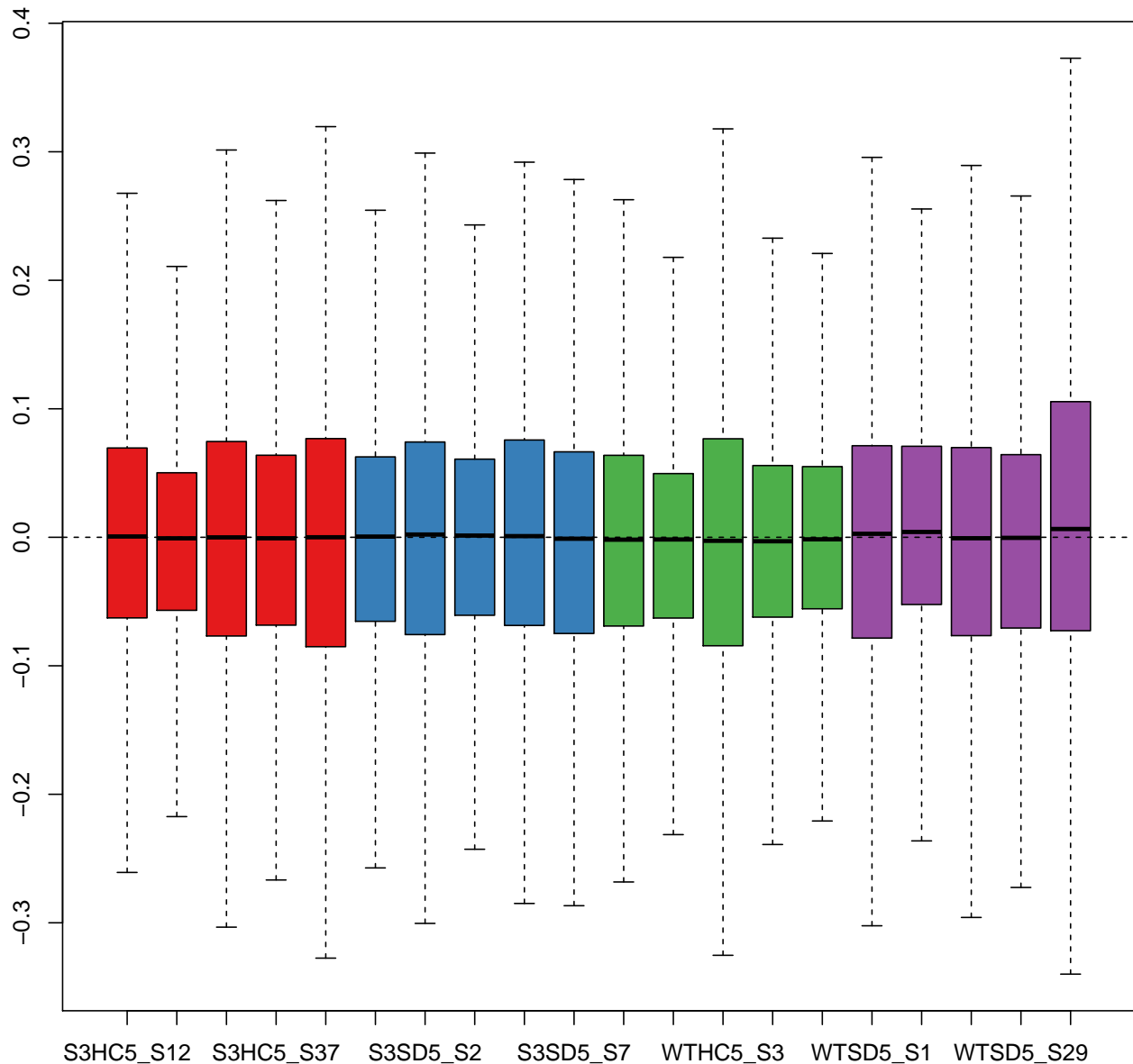
PlotPCAPlotlyFunction(counts.data.frame=log1p(normPropCountsUqua),
                      design.matrix=designMatrix, shapeColname="condition",
                      colorColname="genotype", xPCA="PC1", yPCA="PC2",
                      plotly.flag=FALSE, show.plot.flag=TRUE,
                      prefix.plot="TMM-Norm")

```

```
## [1] FALSE
```



```
pal <- RColorBrewer::brewer.pal(9, "Set1")  
plotRLE(as.matrix(normPropCountsUqua), outline=FALSE, col=pal[designMatrix$gcondition])
```



## TMM + RUVs Normalization

Applying a *RUVs* method of *RUVSeq* package on normalized data, in order to adjust the counts for the unwanted variation. And of course we plot a PCA and an RLE plot on these data.

```
library(RUVSeq)
neg.ctrl.list <- rownames(counts.neg.ctrls)
groups <- makeGroups(designMatrix$gcondition)
ruvedExprData <- RUVs(as.matrix(round(normPropCountsUqua)), cIdx=neg.ctrl.list,
                      scIdx=groups, k=5)

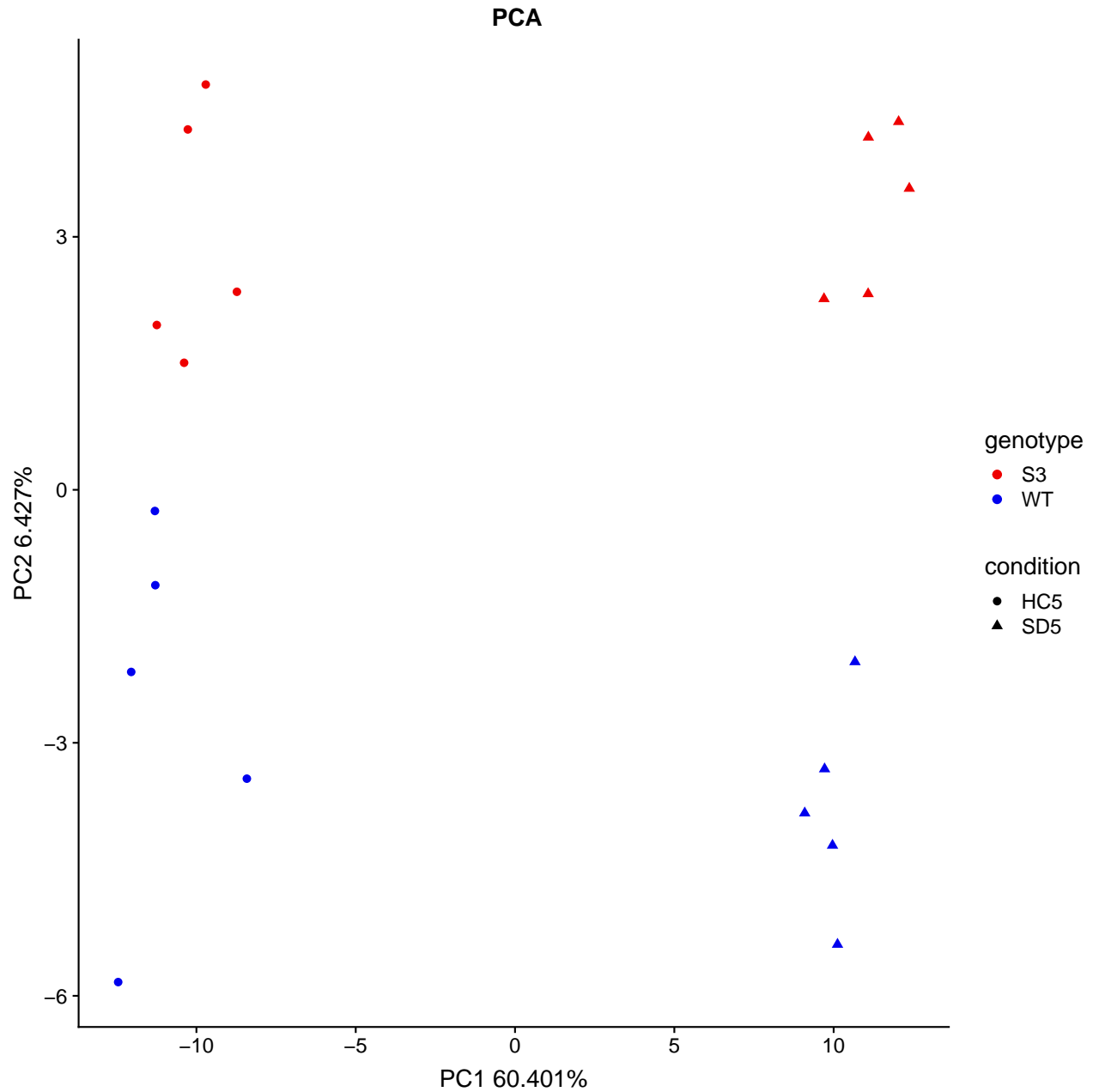
normExprData <- ruvedExprData$normalizedCounts

ggp <- PlotPCAPlotlyFunction(counts.data.frame=log1p(normExprData),
                             design.matrix=designMatrix, shapeColname="condition",
                             colorColname="genotype", xPCA="PC1", yPCA="PC2",
```

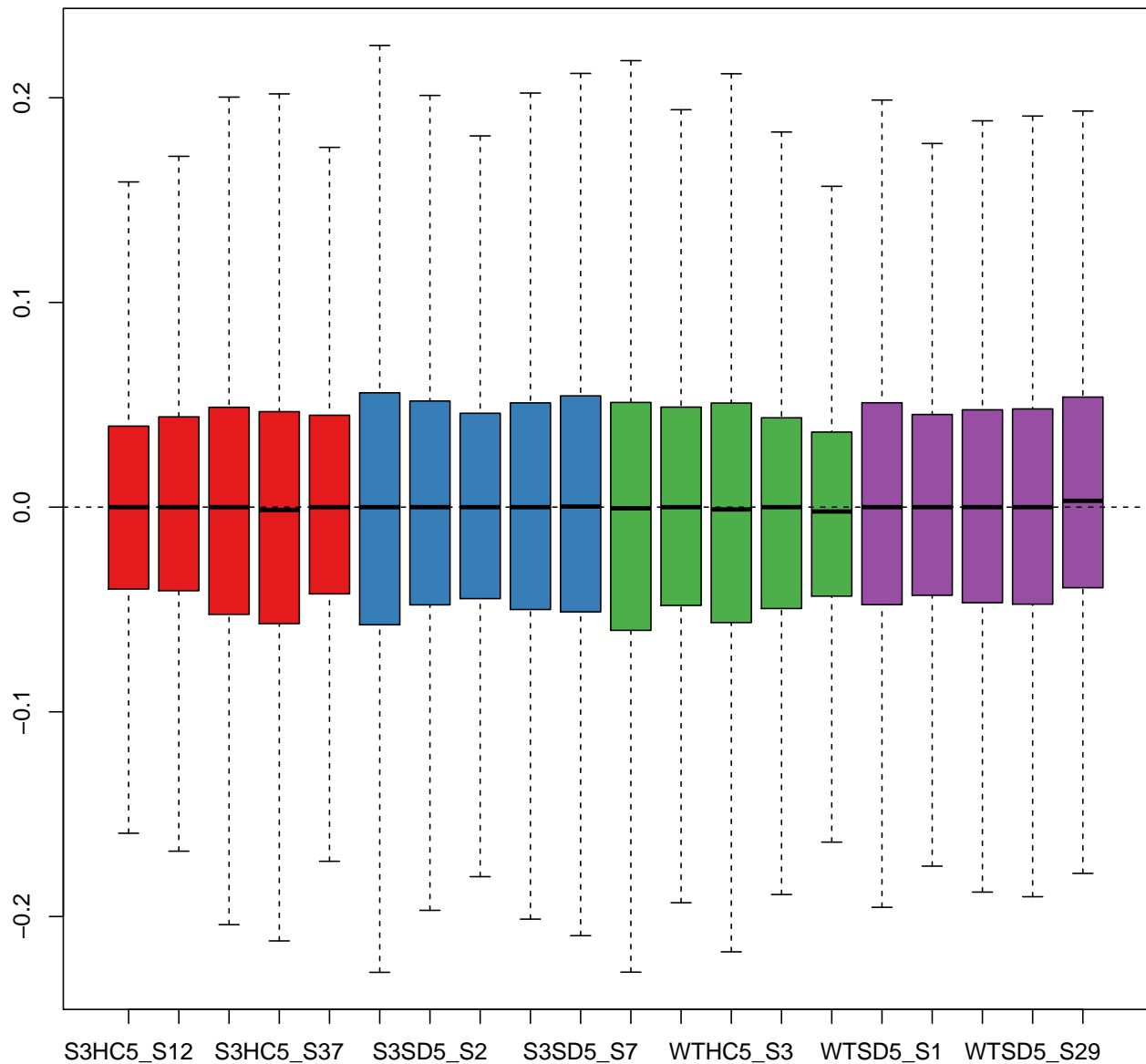
```
plotly.flag=FALSE, show.plot.flag=FALSE, save.plot=FALSE,  
prefix.plot=NULL)
```

```
## [1] FALSE
```

```
print(ggp)
```



```
dir.create("plots")  
save_plot(filename="plots/PCA.pdf", plot=ggp)  
  
pal <- RColorBrewer::brewer.pal(9, "Set1")  
plotRLE(normExprData, outline=FALSE, col=pal[designMatrix$gcondition])
```



## edgeR Differential Expression Analysis

Making differential expression analysis with *edgeR* package on four different contrasts.

Here is a brief legend:

- WTHC5: Wild Type Home Cage Control 5 days
- WTSD5: Wild Type Sleep Deprivation 5 days.
- KOHC5: Knock Out Home Cage Control 5 days.
- KOSD5: Knock Out Sleep Deprivation 5 days.

```
padj.thr <- 0.05
venn.padj.thr <- 0.1
desMat <- cbind(designMatrix, ruvedSEExprData$W)
colnames(desMat) <- c(colnames(designMatrix), colnames(ruvedSEExprData$W))
```



```

cc <- c("S3HC5 - WTHC5", "S3SD5 - WTSD5")

rescList1 <- applyEdgeR(counts=filteredCountsProp, design.matrix=desMat,
                        factors.column="gcondition",
                        weight.columns=c("W_1", "W_2", "W_3", "W_4", "W_5"),
                        contrasts=cc, useIntercept=FALSE, p.threshold=1,
                        is.normalized=FALSE, verbose=TRUE)

names <- names(rescList1)
rescList1 <- lapply(seq_along(rescList1), function(i)
{
  attachMeans(normalized.counts=normExprData, design.matrix=desMat,
              factor.column="gcondition", contrast.name=names(rescList1)[i],
              de.results=rescList1[[i]])
})
names(rescList1) <- names

```

## Shank3 Home Cage control VS Wild Type Home Cage Controls

### volcano plot

A volcano plot of differential expressed genes.

```

res.o.map1 <- convertGenesViaMouseDb(gene.list=rownames(rescList1[[1]]),
                                   fromType="ENTREZID")

res.o <- attachGeneColumnToDf(mainDf=rescList1[[1]],
                              genesMap=res.o.map1,
                              rowNamesIdentifier="ENTREZID",
                              mapFromIdentifier="ENTREZID",
                              mapToIdentifier="SYMBOL")

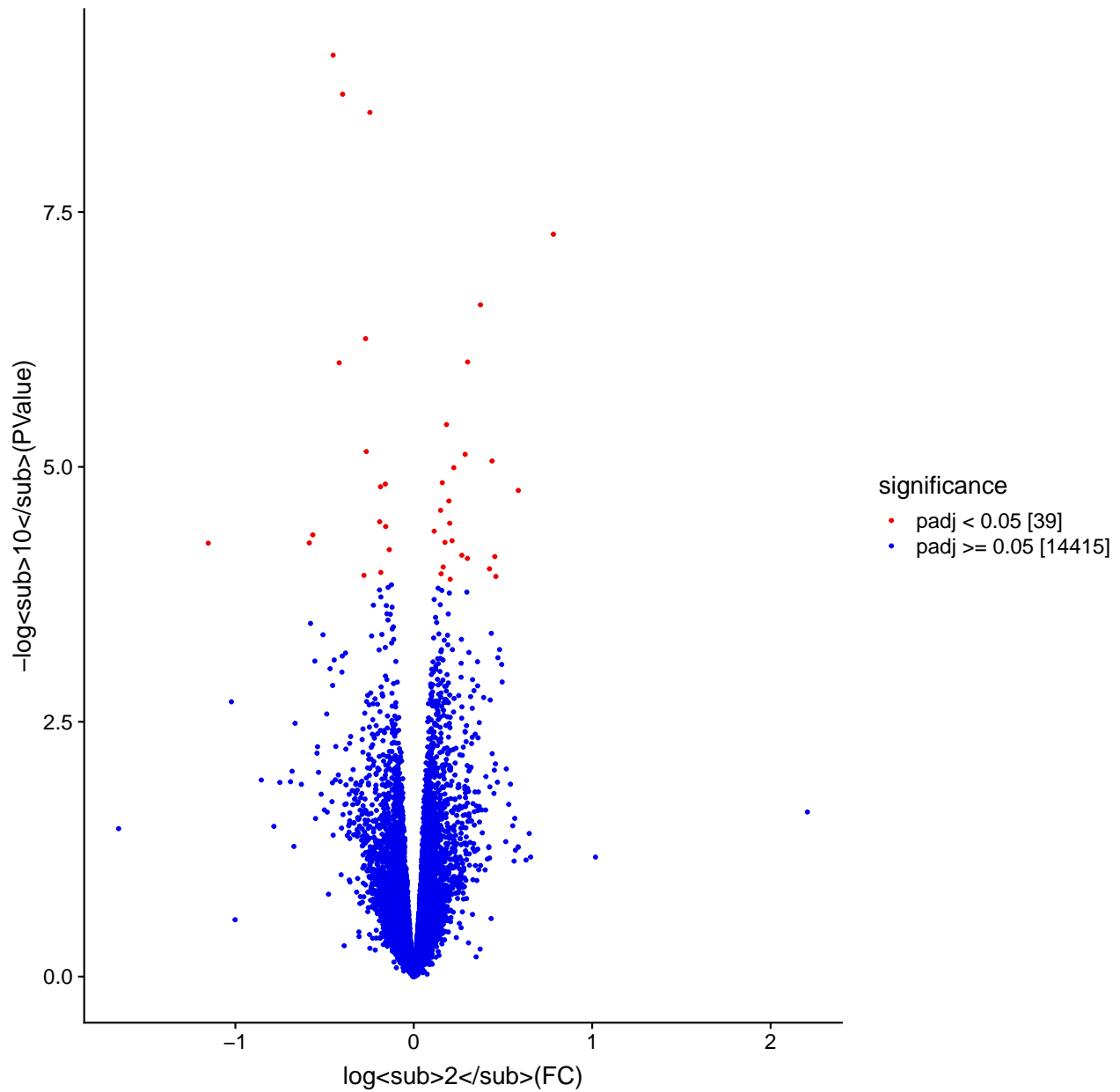
WriteDataFrameAsTsv(data.frame.to.save=res.o,
                    file.name.path=paste0(names(rescList1)[1], "_edgeR"))

vp <- luciaVolcanoPlot(res.o, prefix=names(rescList1)[1],
                      positive.controls.df=NULL,
                      threshold=padj.thr)

print(vp)

```

### S3HC5 – WTHC5 Volcano Plot



```
de <- sum(res.o$FDR < padj.thr)
nde <- sum(res.o$FDR >= padj.thr)
detable <- cbind(de,nde)
rownames(detable) <- names(rescList1)[1]
ddetable <- detable

tot.ctrls <- dim(sd.pos.ctrls)[1]
idx.pc <- which(tolower(res.o$gene) %in% tolower(sd.pos.ctrls[,1]))
tot.pc.de <- sum(res.o$FDR[idx.pc] < padj.thr)
tot.pc.nde <- length(idx.pc) - tot.pc.de

wt <- res.o[which(res.o$FDR < padj.thr),]
wt.sign.genes.entrez <- rownames(res.o)[which(res.o$FDR < venn.padj.thr)]
```

```
kowthc5 <- res.o[which(res.o$FDR < padj.thr),]
kowthc5.sign.genes.entrez <- rownames(res.o)[which(res.o$FDR < venn.padj.thr)]
```

## Shank3 Sled Deprivation VS Wild Type Sleep Deprivation

### volcano plot

A volcano plot of differential expressed genes.

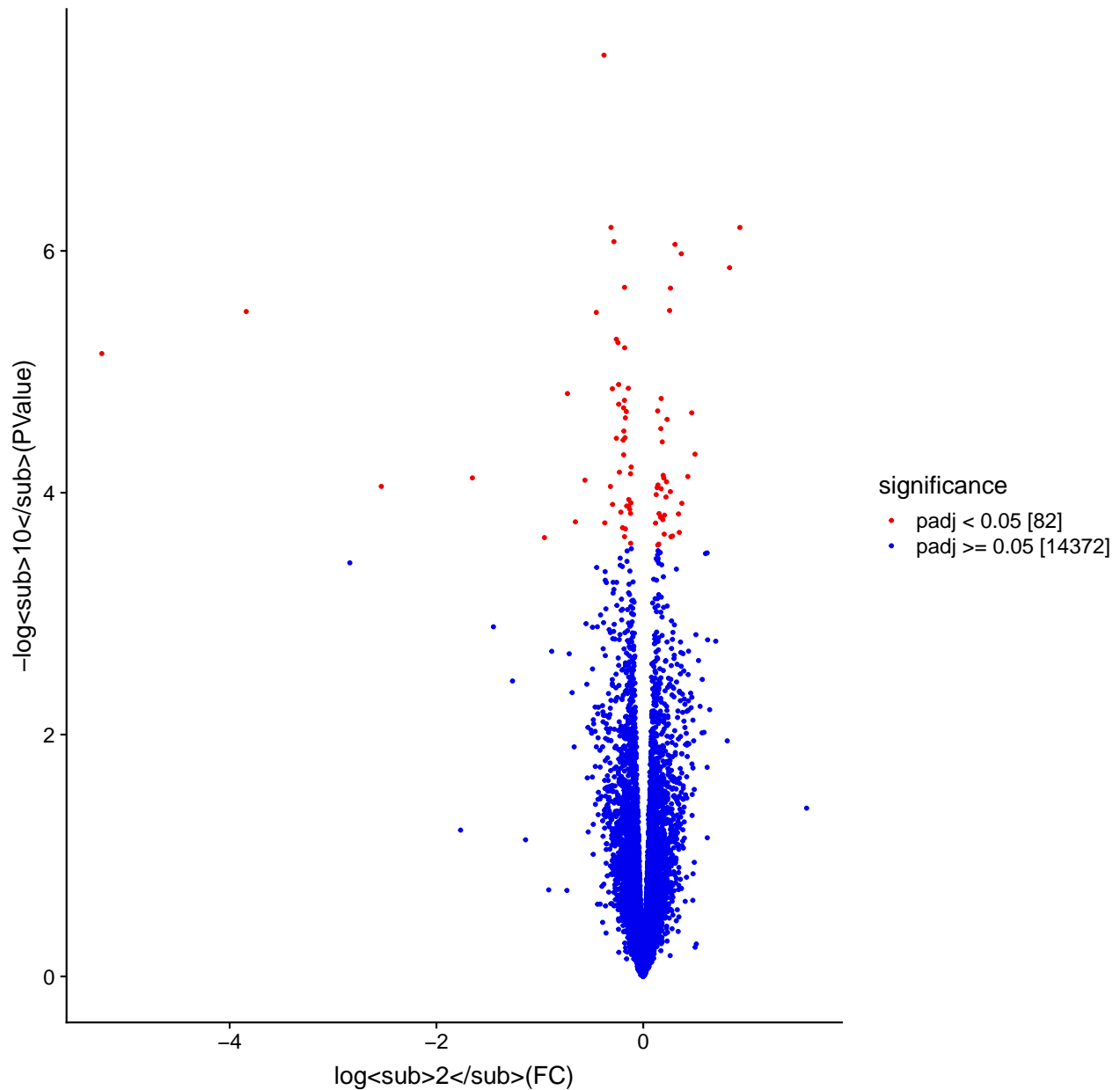
```
rs2.o.map <- convertGenesViaMouseDb(gene.list=rownames(rescList1[[2]]),
                                   fromType="ENTREZID")

res.rs2.o <- attachGeneColumnToDf(mainDf=rescList1[[2]],
                                   genesMap=rs2.o.map,
                                   rowNamesIdentifier="ENTREZID",
                                   mapFromIdentifier="ENTREZID",
                                   mapToIdentifier="SYMBOL")
WriteDataFrameAsTsv(data.frame.to.save=res.rs2.o,
                    file.name.path=paste0(names(rescList1)[2], "_edgeR"))

vp <- luciaVolcanoPlot(res.rs2.o, positive.controls.df=NULL,
                       prefix=names(rescList1)[2],
                       threshold=padj.thr)

print(vp)
```

### S3SD5 – WTSD5 Volcano Plot



```
de <- sum(res.rs2.o$FDR < padj.thr)
nde <- sum(res.rs2.o$FDR >= padj.thr)
detable <- cbind(de,nde)
rownames(detable) <- names(rescList1)[2]
ddetable <- rbind(ddetable, detable)
pos.df <- cbind(tot.ctrls, tot.pc.de, tot.pc.nde)
colnames(pos.df) <- c("total_p.ctrl", "p.ctrl_de_mapped",
                     "p.ctrl_notde_mapped")
rownames(pos.df) <- names(rescList1)[2]

kowtsd5 <- res.rs2.o[which(res.rs2.o$FDR < padj.thr),]
kowtsd5.sign.genes.entrez <- rownames(res.rs2.o)[which(res.rs2.o$FDR < venn.padj.thr)]
```

## DE TABLE + Positive Controls table

We present a summarization of the results. The first table is a summarization on how many genes are Differentially Expressed. The second table explains on the first column how many positive controls we have, on the second column how many positive controls have been identified over the differentially expressed genes, and, finally, on the third column how many positive controls have been identified on the NOT differentially expressed genes.

```
ddetable
```

```
##           de   nde
## S3HC5 - WTHC5 39 14415
## S3SD5 - WTSD5 82 14372
```

```
pos.df
```

```
##           total_p.ctrl p.ctrl_de_mapped p.ctrl_notde_mapped
## S3SD5 - WTSD5           579              3              553
```

## Venn Diagram

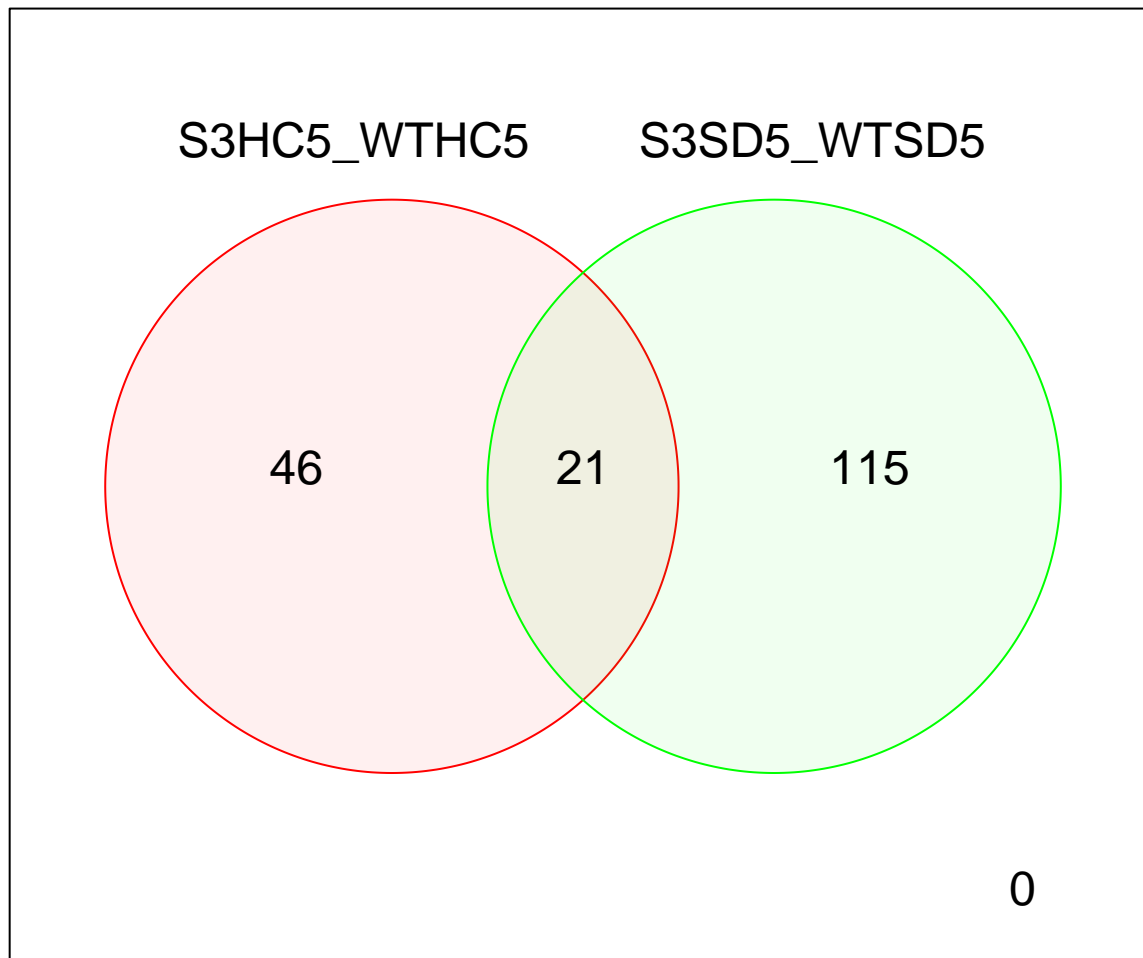
### KOHC5-WTHC5 vs KOSD5-WTSD5

We take the results of the two contrasts. *Knock Out Sleep Deprivation VS Wild Type Sleep Deprivation* and *Knock Out Home Cage control VS Wild Type Home Cage Controls*. And plot the results in a Venn Diagram

```
source("./R/venn2.R")
```

```
gene.map <- convertGenesViaMouseDb(gene.list=rownames(normExprData),
                                   fromType="ENTREZID", toType="SYMBOL")
venn <- Venn2de(x=kowthc5.sign.genes.entrez, y=kowtsd5.sign.genes.entrez,
               label1="S3HC5_WTHC5", label2="S3SD5_WTSD5",
               title="S3HC5_WTHC5 venn S3SD5_WTSD5", plot.dir=".",
               conversion.map=gene.map)
```

## S3HC5\_WTHC5 venn S3SD5\_WTSD5



## Heatmaps

Setting up the data structures for the heatmps.

```
source("./R/heatmapFunctions.R")
de.genes.entr <- union(rownames(venn$int), rownames(venn$XnoY))
de.genes.entr <- union(de.genes.entr, rownames(venn$YnoX))

gene.map <- convertGenesViaMouseDb(gene.list=de.genes.entr,
                                   fromType="ENTREZID")
```

```

de.genes.symb <- attachGeneColumnToDf(as.data.frame(de.genes.entr,
                                                    row.names=de.genes.entr),
                                     genesMap=gene.map,
                                     rowNamesIdentifier="ENTREZID",
                                     mapFromIdentifier="ENTREZID",
                                     mapToIdentifier="SYMBOL")

# de.genes.symb[which(is.na(de.genes.symb$gene)),]
de.genes.symb$gene[which(de.genes.symb$de.genes.entr=="100039826")] <- "Gm2444" ## not annotated in ncl
de.genes.symb$gene[which(de.genes.symb$de.genes.entr=="210541")] <- "Entrez:210541" ## not annotated in ncl

de.genes.counts <- normExprData[match(de.genes.symb$de.genes.entr, rownames(normExprData)),]
rownames(de.genes.counts) <- de.genes.symb$gene

de.gene.means <- computeGeneMeansOverGroups(counts=de.genes.counts,
                                             design=designMatrix, groupColumn="gcondition")

library(gplots)
library(clusterExperiment)
color.palette = clusterExperiment::seqPal3#c("black", "yellow")
pal <- colorRampPalette(color.palette)(n = 1000)

library(pheatmap)
filter2 <- rowMeans(de.gene.means)>0
filter <- apply(de.gene.means, 1, function(x) log(x[4]/x[3]) * log(x[2]/x[1]) < 0)
filter[is.na(filter)] <- FALSE

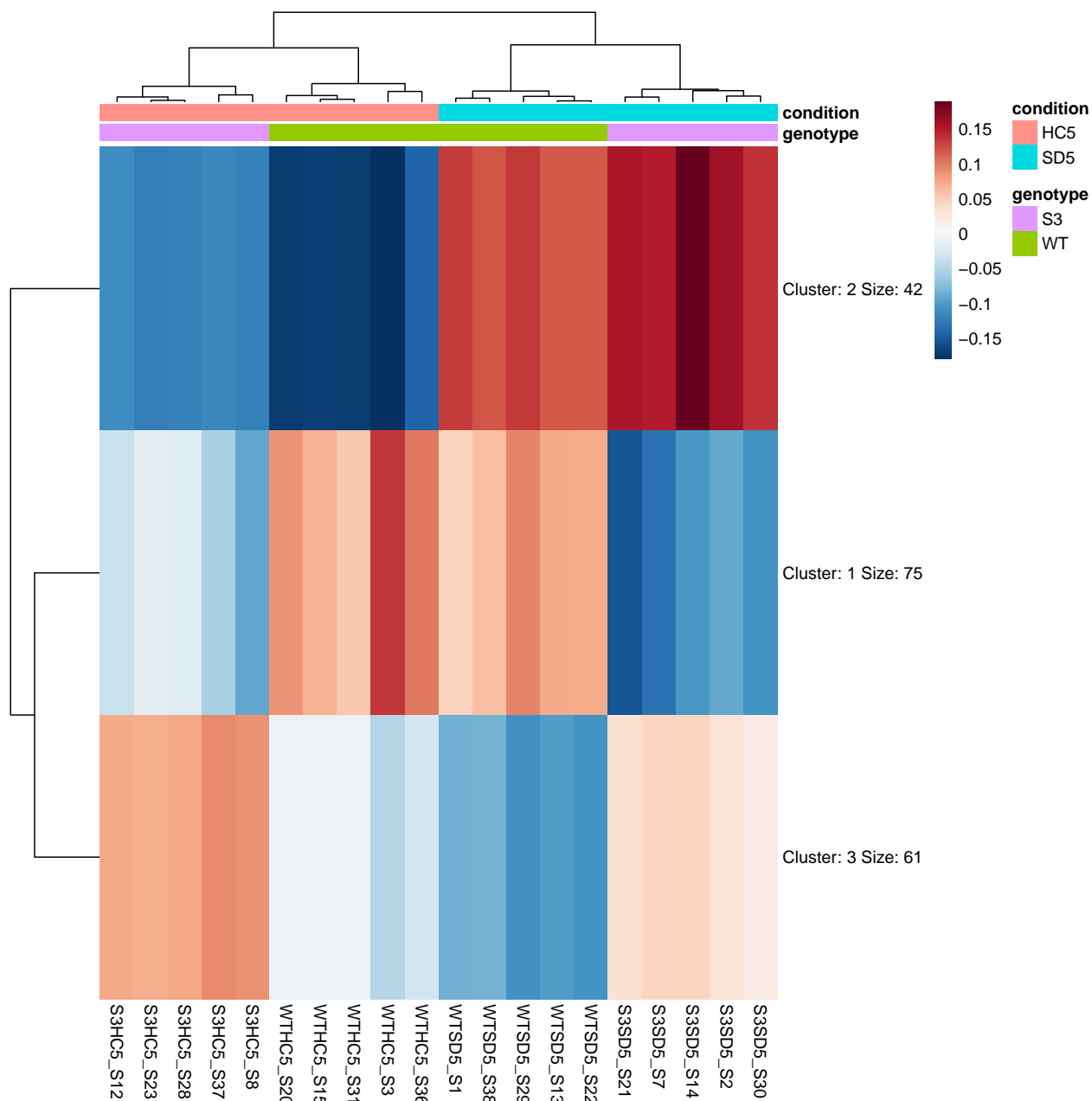
```

## Heatmap gene by bene

```

gene_names <- c("Nr1d1", "Fabp7", "Per3", "Jun", "Elk1", "Fosl2", "Mapk1",
               "Mapk3", "Mapk11", "Hmgcr", "Insig1", "Nfil3", "Stat4",
               "Kcnv1", "Kcnk1", "Kcnk2", "Dusp10", "Dusp3", "Ptprj",
               "Cntn1", "Ntrk2", "Reln", "Sema3a", "Tef", "Hlf", "Nr1d1",
               "Prkab2", "Bhlhe41", "Slc6a15", "Slc22a4", "Slc24a4")
idx <- which(!(rownames(de.genes.counts) %in% gene_names))
de.genes.counts1 <- de.genes.counts
rownames(de.genes.counts1)[idx] <- ""
ann.col <- desMat[, c(1:2)]
de.heatmap <- de.genes.counts[filter2,]
set.seed(0)
heatmap_data <- t(scale(t(log(de.heatmap+1)), center = TRUE, scale = FALSE))
phi <- pheatmap(heatmap_data, cluster_cols=TRUE, scale="none",
               color=pal, border_color=NA, fontsize_row=10, kmeans_k=3, annotation_col=ann.col)

```



```
clusterized.genes <- as.data.frame(ph1$kmeans$cluster)

gene.map <- convertGenesViaMouseDb(gene.list=rownames(clusterized.genes), fromType="SYMBOL")
converted.clusterized.gens <- attachGeneColumnToDf(mainDf=clusterized.genes, genesMap=gene.map,
  rowNamesIdentifier="SYMBOL", mapFromIdentifier="SYMBOL", mapToIdentifier="ENTREZID")

converted.clusterized.gens$gene[which(rownames(converted.clusterized.gens)=="Gm2444")] <- "100039826" ;
converted.clusterized.gens$gene[which(rownames(converted.clusterized.gens)=="Entrez:210541")] <- "210541" ;
converted.clusterized.gens <- converted.clusterized.gens[order(converted.clusterized.gens$ph1$kmeans$cluster)]

save_pheatmap_pdf(filename="plots/heatmap_kmeans_k3.pdf", plot=ph1, width=20, height=20)
```

```
## pdf
## 2
```



```

WriteDataFrameAsTsv(data.frame.to.save=converted.clusterized.gens, file.name.path="plots/clustered_genes")

ord.de.genes.counts <- de.heatmap[match(rownames(converted.clusterized.gens), rownames(de.heatmap)),]
idx <- which(!(rownames(ord.de.genes.counts) %in% gene_names))

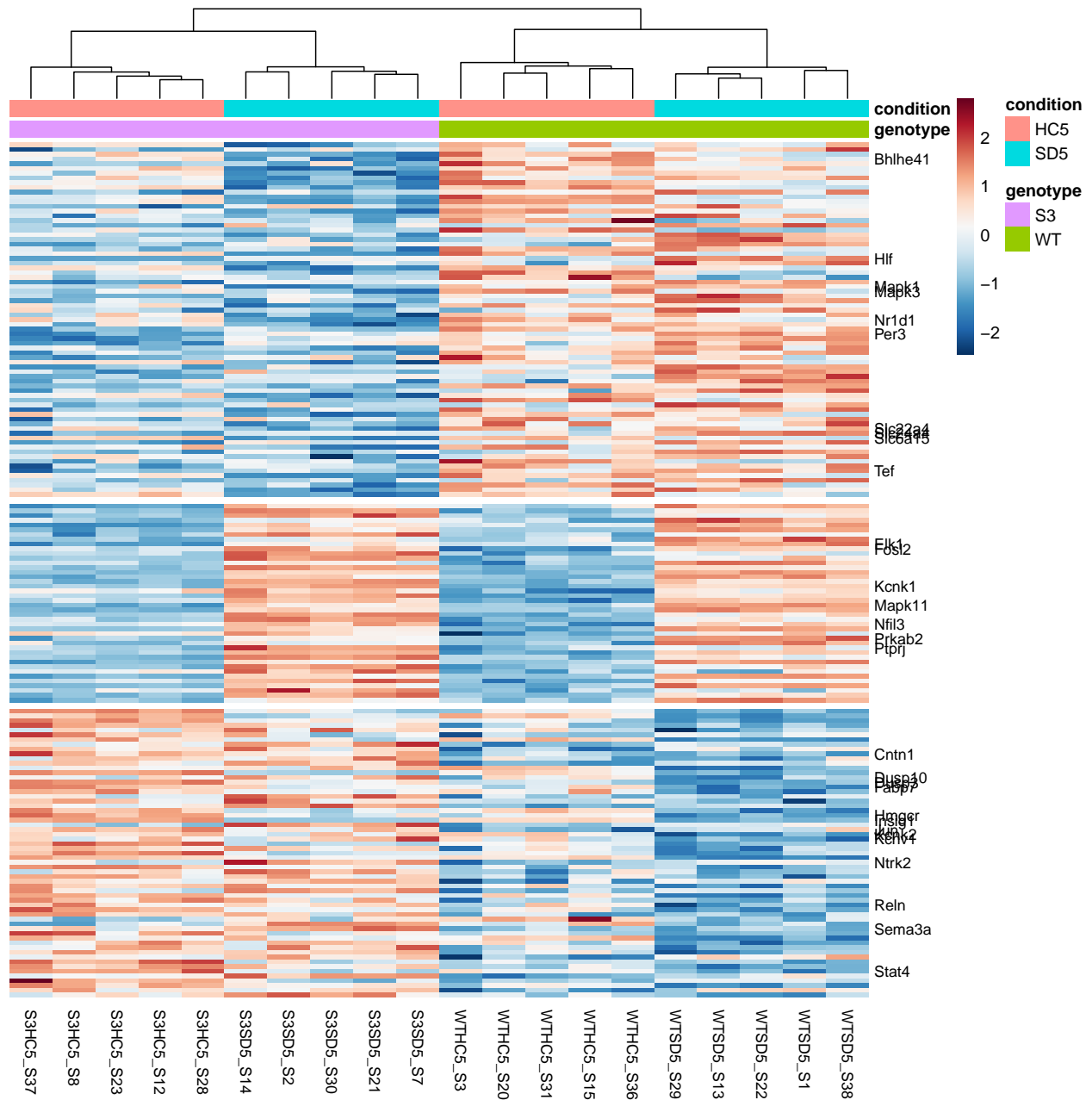
rownames(ord.de.genes.counts)[idx] <- ""
gaps.row <- c()
for(i in c(1:3))
{
  li <- length(which(converted.clusterized.gens$`ph1$kmeans$cluster`==i))
  l <- ifelse(i!=1, gaps.row[i-1]+li, li)
  gaps.row <- c(gaps.row, l)
}

heatmap_data_scaled <- t(scale(t(log(ord.de.genes.counts+1)), center = TRUE, scale = TRUE))

library(dendextend)
column_dend <- as.dendrogram(hclust(dist(t(heatmap_data_scaled))))
ord <- labels(column_dend)
ord[11:15] <- labels(column_dend)[16:20]
ord[16:20] <- labels(column_dend)[11:15]
column_dend <- rotate(column_dend, ord)

ph1 <- pheatmap(heatmap_data_scaled, cluster_cols=as.hclust(column_dend), scale="none",
  color=pal, border_color=NA, fontsize_row=9, fontsize_col=9, cluster_rows=FALSE,
  annotation_col=ann.col, gaps_row=gaps.row)

```



```
save_pheatmap_pdf(filename="plots/heatmap_gg_k3.pdf", plot=ph1, width=20, height=20)
```

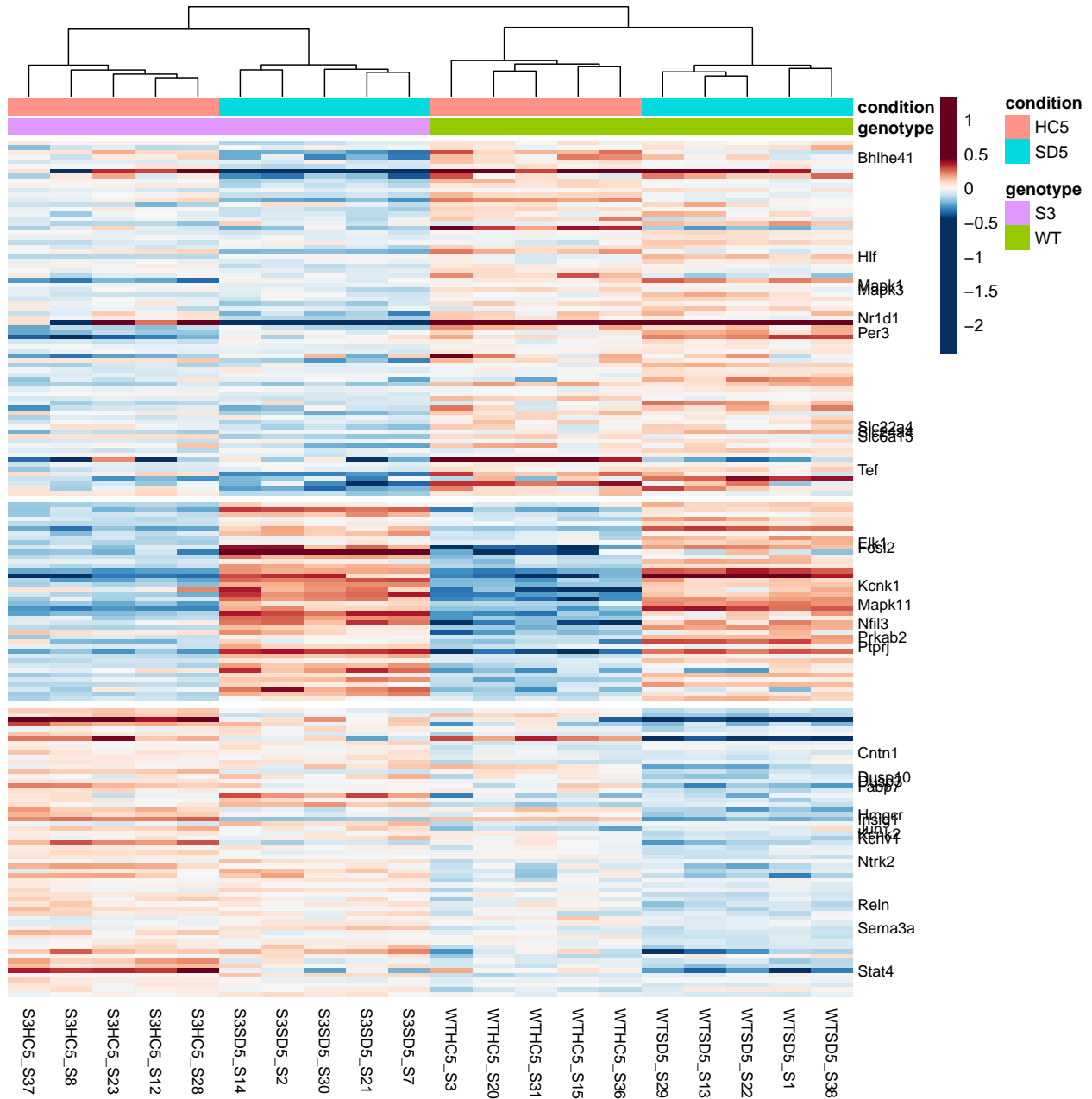
```
## pdf
## 2
```

## other heatmaps

```
heatmap_data <- t(scale(t(log(ord.de.genes.counts+1)), center = TRUE, scale = FALSE))

ph1 <- pheatmap(heatmap_data, cluster_cols=as.hclust(column_dend), scale="none",
  color=pal, border_color=NA, fontsize_row=9, fontsize_col=9, cluster_rows=FALSE,
  annotation_col=ann.col, gaps_row=gaps.row,
```

```
breaks = c(min(heatmap_data), seq(quantile(as.vector(heatmap_data), .01), quantile(as.vector(heatmap_data), .99), by = 0.5), max(heatmap_data))
```



```
save_pheatmap_pdf(filename="plots/heatmap_gg_k3_no_scale.pdf", plot=ph1, width=20, height=20)
```

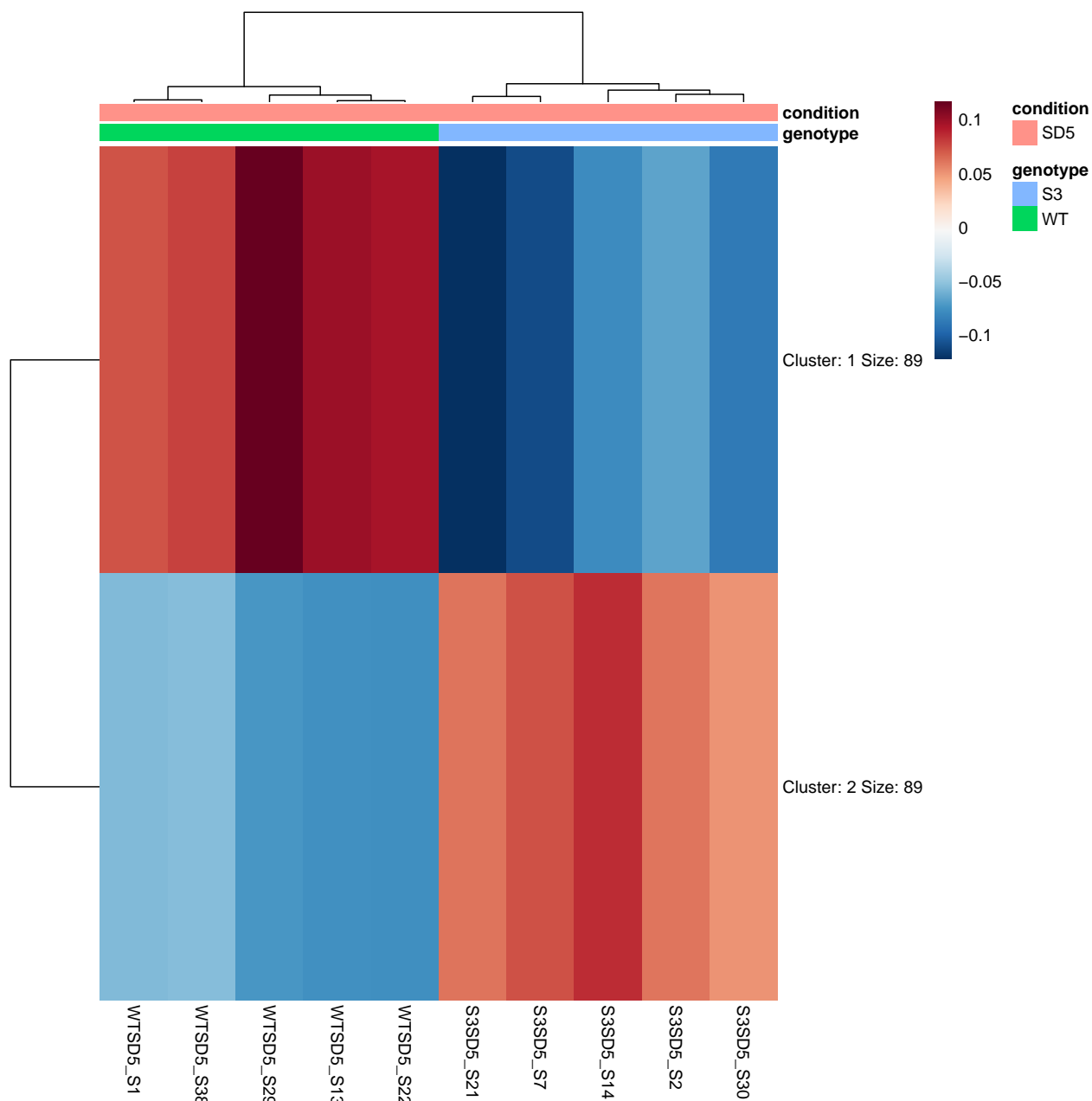
```
## pdf
```

```
## 2
```

```
## Only SD samples
```

```
heatmap_data <- t(scale(t(log(de.heatmap[, grep("^SD", desMat[,2]))+1)), center = TRUE, scale = FALSE))
```

```
ph1 <- pheatmap(heatmap_data, cluster_cols=TRUE, scale="none",  
color=pal, border_color=NA, fontsize_row=10, kmeans_k=2, annotation_col=ann.col)
```



```
clusterized.genes <- as.data.frame(ph1$kmeans$cluster)

gene.map <- convertGenesViaMouseDb(gene.list=rownames(clusterized.genes), fromType="SYMBOL")
converted.clusterized.gens <- attachGeneColumnToDf(mainDf=clusterized.genes, genesMap=gene.map,
  rowNamesIdentifier="SYMBOL", mapFromIdentifier="SYMBOL", mapToIdentifier="ENTREZID")

converted.clusterized.gens$gene[which(rownames(converted.clusterized.gens)=="Gm2444")] <- "100039826" ;
converted.clusterized.gens$gene[which(rownames(converted.clusterized.gens)=="Entrez:210541")] <- "210541" ;
converted.clusterized.gens <- converted.clusterized.gens[order(converted.clusterized.gens$ph1$kmeans$cluster),]

ord.de.genes.counts <- de.heatmap[match(rownames(converted.clusterized.gens), rownames(de.heatmap)),]
idx <- which(!(rownames(ord.de.genes.counts) %in% gene_names))
```

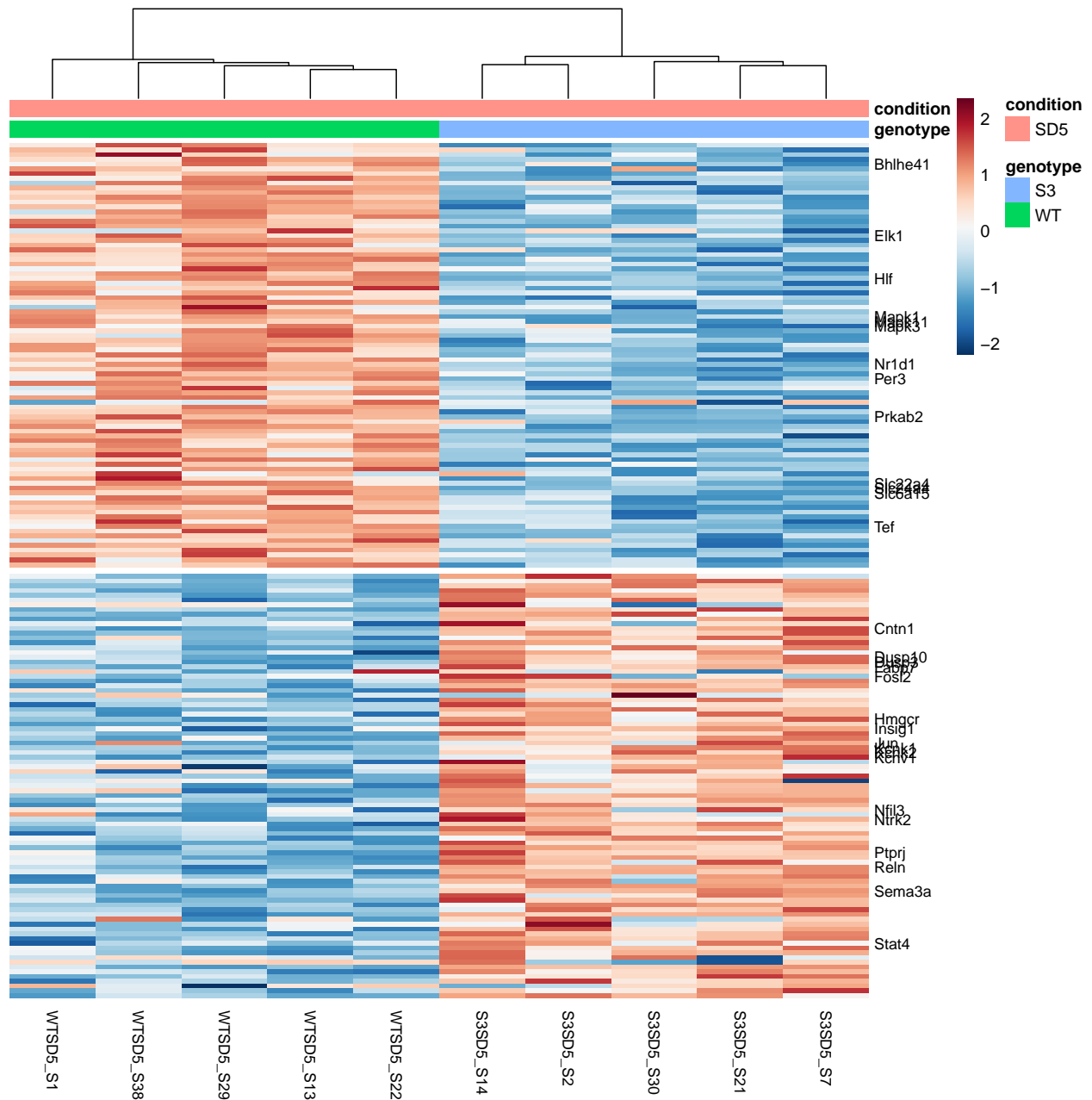
```

rownames(ord.de.genes.counts)[idx] <- ""
gaps.row <- c()
for(i in c(1:2))
{
  li <- length(which(converted.clusterized.genes$`ph1$kmeans$cluster`==i))
  l <- ifelse(i!=1, gaps.row[i-1]+li, li)
  gaps.row <- c(gaps.row, l)
}

heatmap_data <- t(scale(t(log(ord.de.genes.counts[, grep("^SD", desMat[,2]))+1)), center = TRUE, scale =

ph1 <- pheatmap(heatmap_data, cluster_cols=TRUE, scale="none",
  color=pal, border_color=NA, fontsize_row=9, fontsize_col=9, cluster_rows=FALSE,
  annotation_col=ann.col, gaps_row=gaps.row)

```



```
save_pheatmap_pdf(filename="plots/heatmap_gg_sd_only.pdf", plot=ph1, width=20, height=20)
```

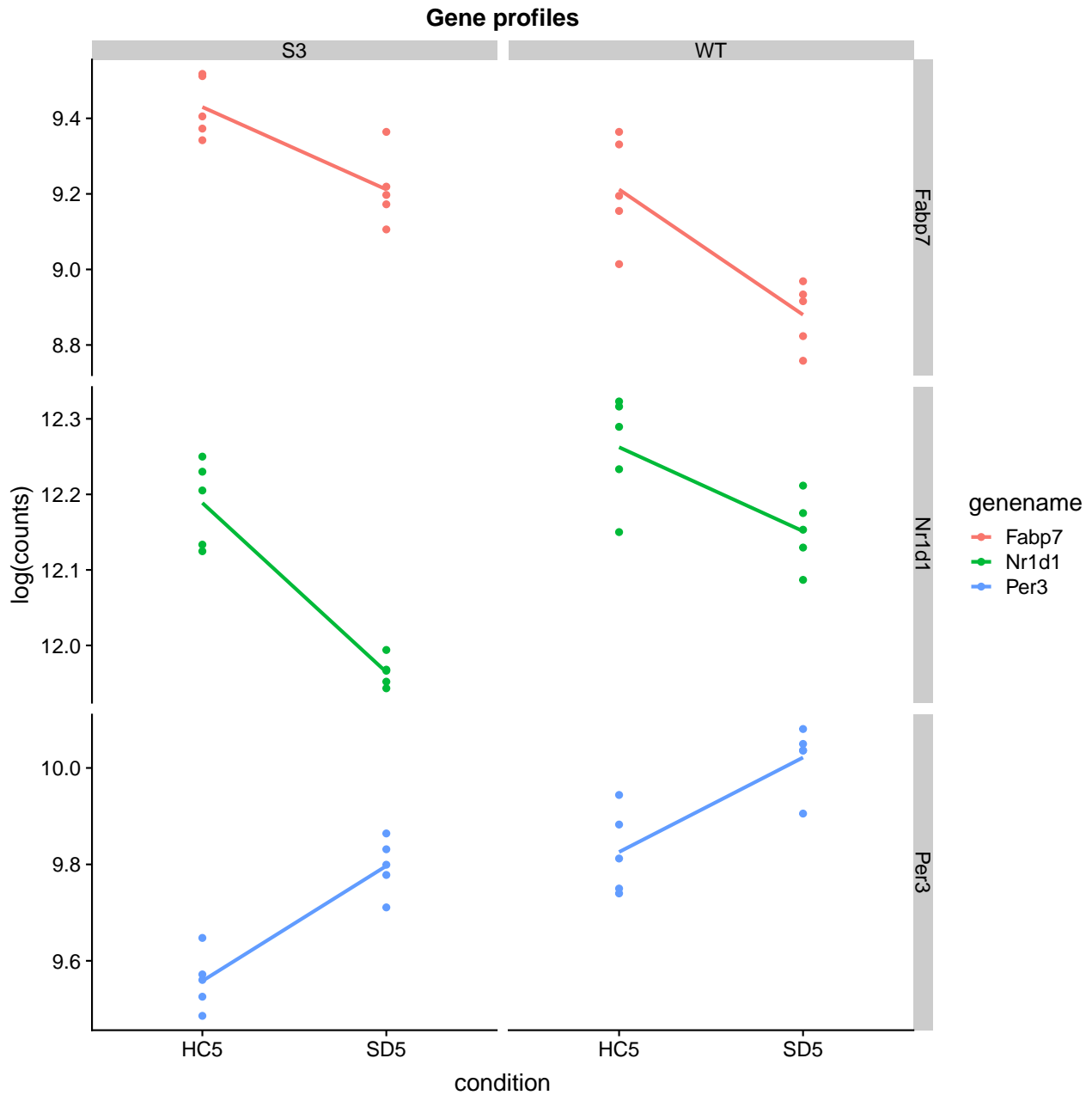
```
## pdf
## 2
```

## Group gene profiles

### Group gene profiles by genotype

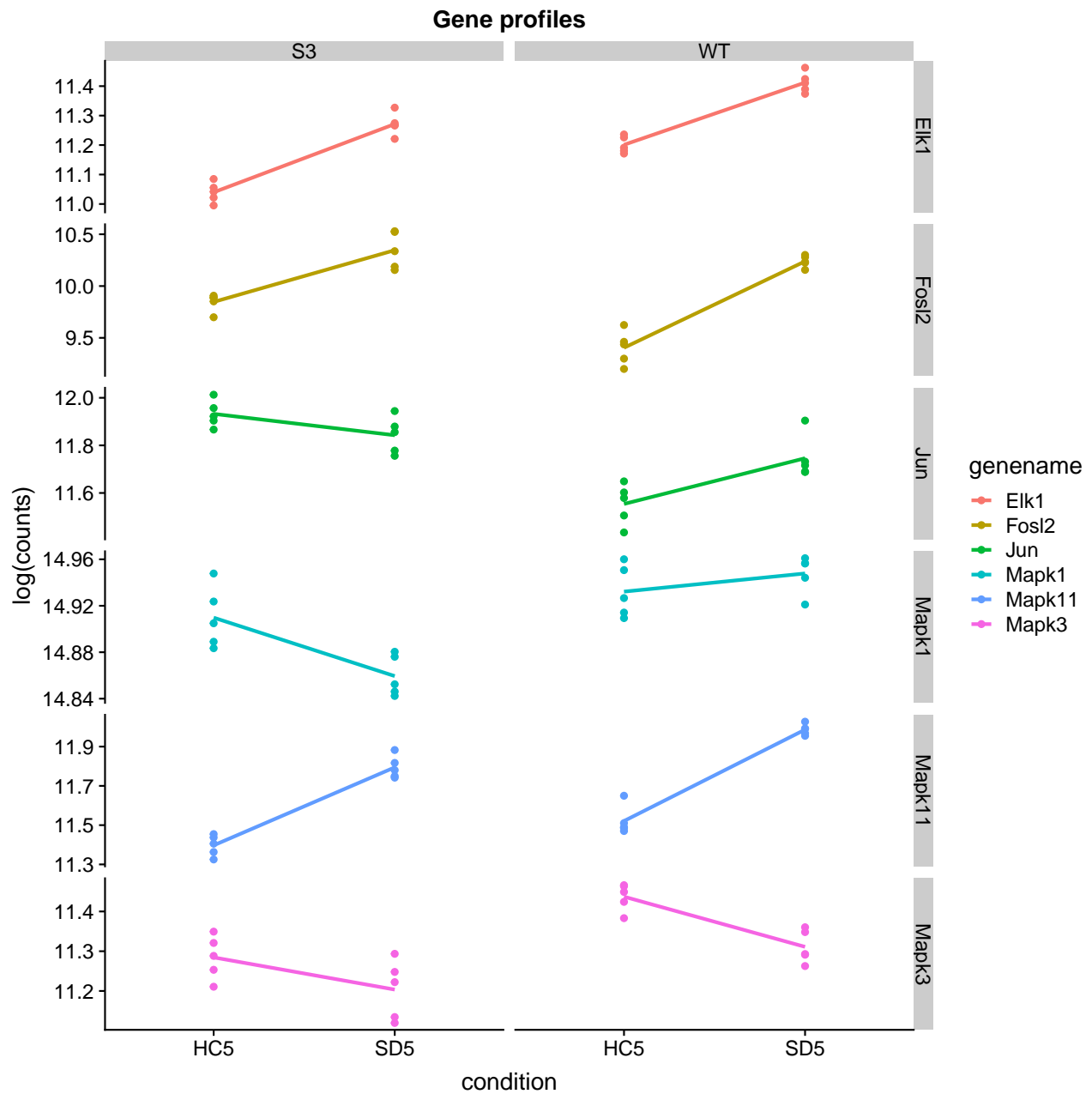
```
g <- geneGroupProfileRows(normalized.counts=normExprData, design.matrix=designMatrix,
  gene.names=c("Nr1d1", "Fabp7", "Per3"),
```

```
res.o=de.genes.symb, show.plot=TRUE, plotly.flag=FALSE, log.flag=TRUE)
print(g)
```



```
save_plot(filename=paste0("plots/", "Nr1d1_Fabp7_Per3", "_log_gene_profile_genotype.pdf"), plot=g,
           base_height=15, base_width=15)

g <- geneGroupProfileRows(normalized.counts=normExprData, design.matrix=designMatrix,
                           gene.names=c("Jun", "Elk1", "Fosl2", "Mapk1", "Mapk3", "Mapk11"),
                           res.o=de.genes.symb, show.plot=TRUE, plotly.flag=FALSE, log.flag=TRUE)
print(g)
```

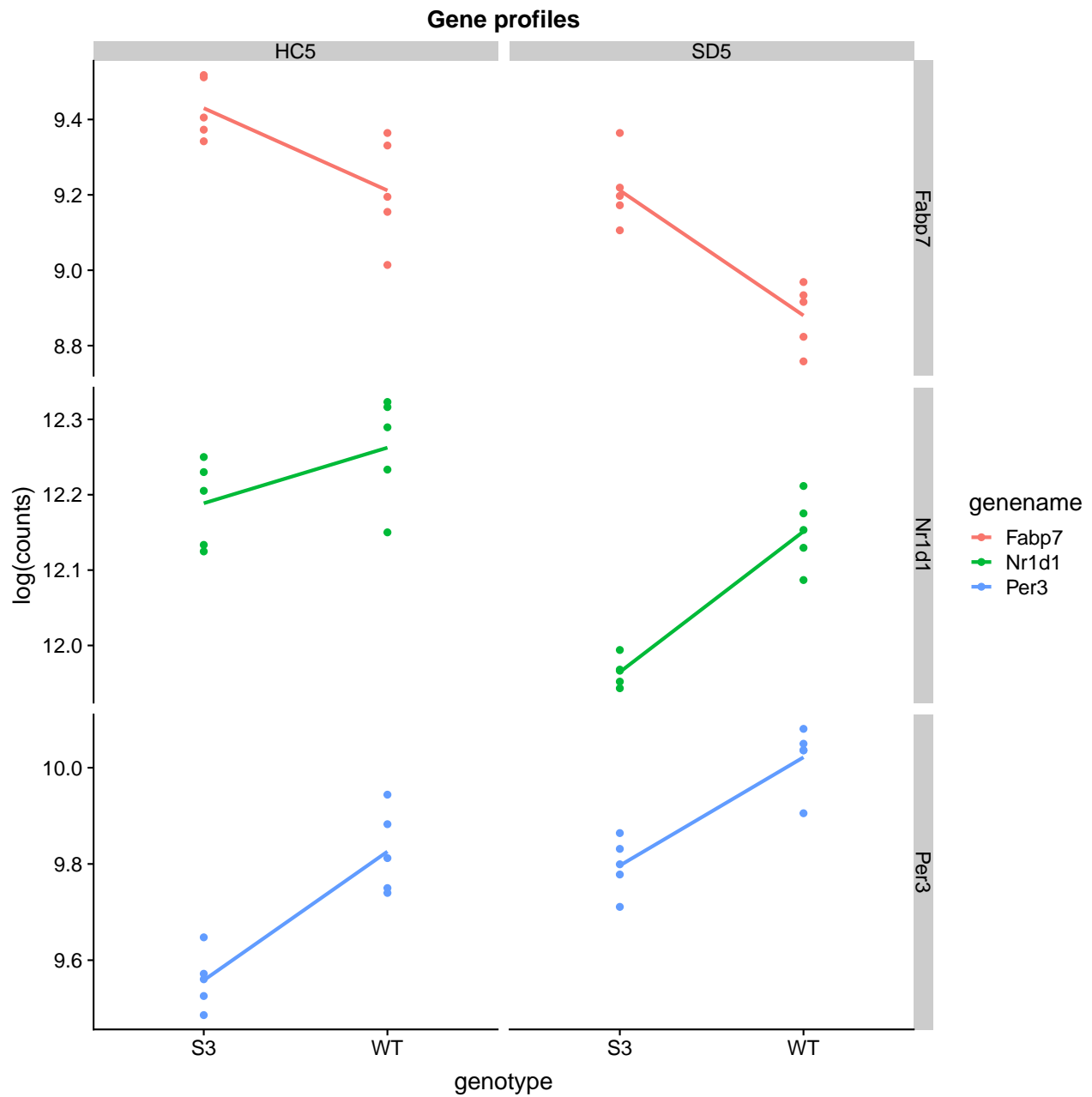


```
save_plot(filename=paste0("plots/", "Jun_Elk1_Fosl2_Mapk", "_log_gene_profile_genotype.pdf"), plot=g,
           base_height=15, base_width=15)
```

## Group gene profiles by condition

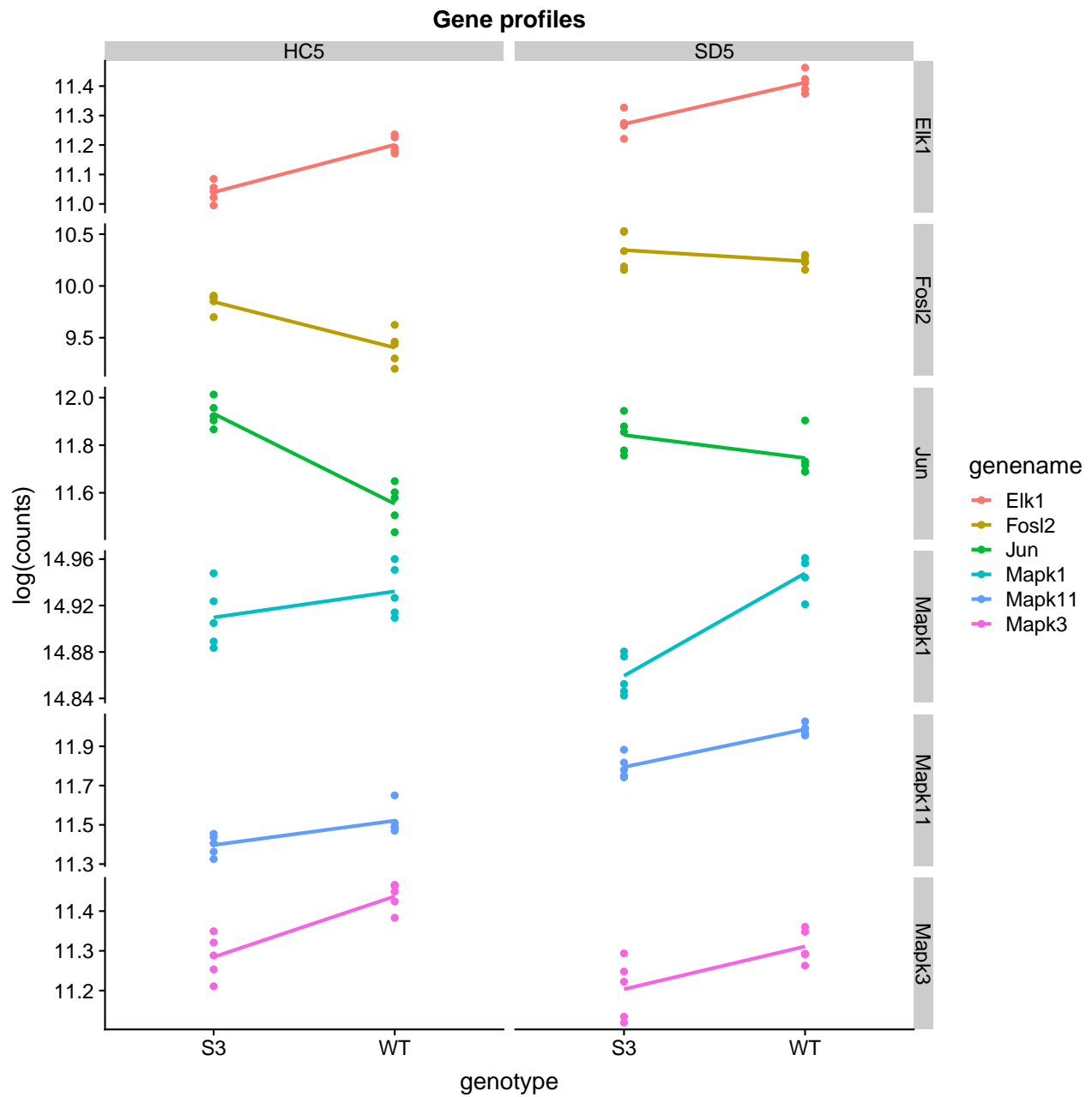
```
g <- geneGroupProfileRowsRev(normalized.counts=normExprData, design.matrix=designMatrix,
                             gene.names=c("Nr1d1", "Fabp7", "Per3"),
                             res.o=de.genes.symb, show.plot=TRUE, plotly.flag=FALSE, log.flag=TRUE)
print(g)
```





```
save_plot(filename=paste0("plots/", "Nr1d1_Fabp7_Per3", "_log_gene_profile_condition.pdf"), plot=g,
          base_height=15, base_width=15)

g <- geneGroupProfileRowsRev(normalized.counts=normExprData, design.matrix=designMatrix,
                             gene.names=c("Jun", "Elk1", "Fos12", "Mapk1", "Mapk3", "Mapk11"),
                             res.o=de.genes.symb, show.plot=TRUE, plotly.flag=FALSE, log.flag=TRUE)
print(g)
```



```
save_plot(filename=paste0("plots/", "Jun_Elk1_Fosl2_Mapk", "_log_gene_profile_condition.pdf"), plot=g,
          base_height=15, base_width=15)
```