## R Notebook



```
#install.packages(c("ggrepel", "dplyr", "gridExtra", "grid", "lattice", "ggpubr", "umap", "ggfortify", "magick",
"BiocManager", "readxl", "rmarkdown", "tibble", "rlist"))
#BiocManager::install(c('DESeq2', 'ComplexHeatmap', 'biomaRt', 'fgsea', 'msiqdbr'))
library(rmarkdown)
library(readxl)
library(DESeq2)
library(ComplexHeatmap)
library(circlize)
library(dendsort)
library(biomaRt)
library(fgsea)
library(data.table)
library(msigdbr)
library(ggrepel)
library(dplyr)
library(gridExtra)
library(grid)
library(lattice)
library(ggpubr)
library(umap)
library(ggfortify)
library(magick)
library(tibble)
library(rlist)
library(outliers)
```

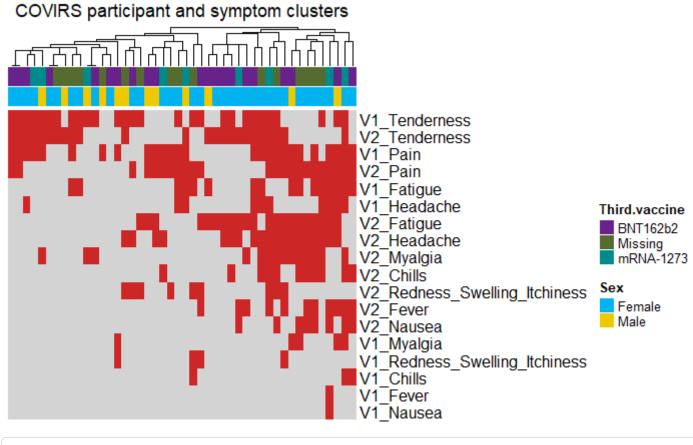
```
# ensembl=useMart("ensembl")
# ensembl = useDataset("hsapiens_gene_ensembl", mart=ensembl)
# attributes = listAttributes(ensembl)
# ensids=rownames(count_data_deseq)
# ens2gene = getBM(attributes=c('external_gene_name','ensembl_gene_id','gene_biotype','entrezgene_id', 'descripti on'),
# filters = 'ensembl_gene_id',
# values = ensids,
# mart = ensembl)
# ens2gene = ens2gene[!duplicated(ens2gene$ensembl_gene_id),]
# rownames(ens2gene) = ens2gene$ensembl_gene_id
# head(ens2gene)
# saveRDS(ens2gene, "ens2gene.RDS")
```

```
count_data = readRDS('data/Raw_RNASeq_Data.RDS')
ens2gene = readRDS("data/ens2gene.RDS")
meta_data = data.frame(read.csv('data/COVIRS_metadata.csv', header = TRUE))
```

### 1. Defining patient groups based on symptoms

```
meta_data = meta_data[meta_data$Timepoint == 'V2A' & meta_data$First.Second.vaccine == 'BNT162b2', ]
```

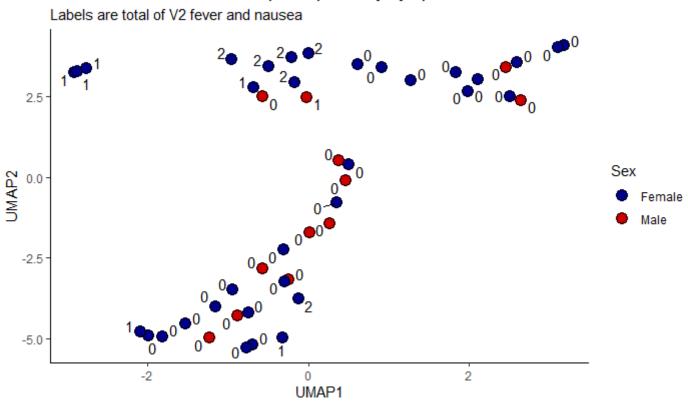
```
plot = data.frame(subset(meta_data, select = -c(id, Participant.ID, Run, SAGC_ID, First.Second.vaccine, Third.vaccine, Sex, Timepoint)))
col_dend = dendsort(hclust(dist(plot)))
ha_participants = HeatmapAnnotation(df = meta_data[ ,c('Third.vaccine', 'Sex')], col = list(Third.vaccine = c('mR
NA-1273' = 'darkcyan', 'BNT162b2' = 'darkorchid4', 'Missing' = 'darkolivegreen'), Sex = c('Male' = 'gold2', 'Fema
le' = 'deepskyblue2')), show_annotation_name = FALSE)
Heatmap(as.matrix(t(plot)), show_row_names = TRUE, show_column_names = FALSE, cluster_rows = TRUE, cluster_column
s = col_dend, show_row_dend = FALSE, top_annotation = ha_participants, show_heatmap_legend = FALSE, col = c('ligh
tgrey', 'firebrick3'), column_title = 'COVIRS participant and symptom clusters')
```



rm(plot); rm(ha\_participants); rm(col\_dend);

```
custom.config = umap.defaults
custom.config$random_state = 1
custom.config$n_neighbors = 3
plot = meta_data
umap = umap(as.matrix(apply(subset(plot, select = -c(id, Participant.ID, Run, SAGC_ID, Timepoint, First.Second.va
ccine, Third.vaccine, Sex)), 2, as.numeric)), config = custom.config)
umap = data.frame(umap$layout)
plot$UMAP1 = umap$X1
plot$UMAP2 = umap$X2
rm(umap)
ggplot(plot, aes(x = UMAP1, y = UMAP2, fill = Sex, label = apply(subset(plot, select = c(V2_Fever, V2_Nausea)) ==
1, 1, sum))) + geom_point(shape = 21, size = 4) +
    scale_fill_manual(values = c(Male = "red3", Female = "blue4")) + theme_classic() + geom_text_repel() +
    ggtitle('UMAP of Pfizer-vaccinated participants by symptoms', subtitle = 'Labels are total of V2 fever and naus
ea'); rm (plot)
```

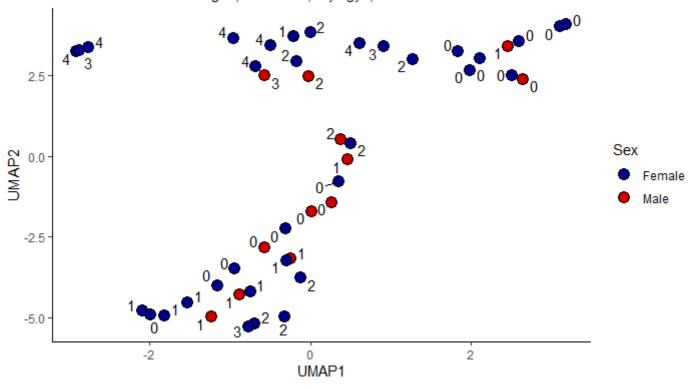
### UMAP of Pfizer-vaccinated participants by symptoms



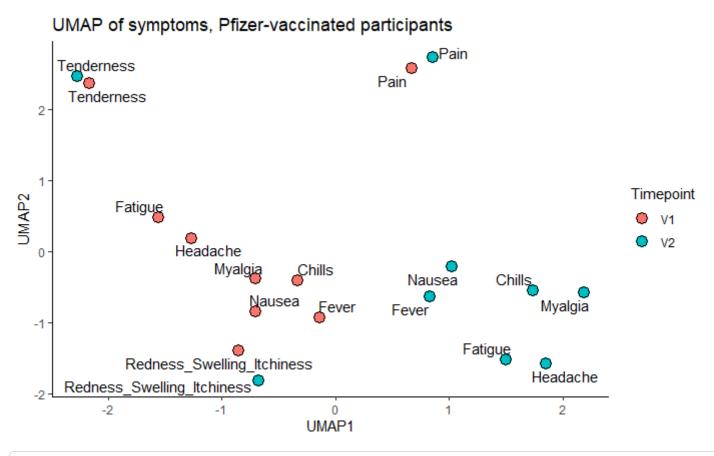
```
plot = meta_data
umap = umap(as.matrix(apply(subset(plot, select = -c(id, Participant.ID, Run, SAGC_ID, Timepoint, First.Second.va
ccine, Third.vaccine, Sex)), 2, as.numeric)), config = custom.config)
umap = data.frame(umap$layout)
plot$UMAP1 = umap$X1
plot$UMAP2 = umap$X2
rm(umap)
ggplot(plot, aes(x = UMAP1, y = UMAP2, fill = Sex, label = apply(subset(plot, select = c(V2_Fatigue, V2_Headache,
V2_Myalgia, V2_Chills)) == 1, 1, sum))) + geom_point(shape = 21, size = 4) +
    scale_fill_manual(values = c(Male = "red3", Female = "blue4")) + theme_classic() + geom_text_repel() +
    ggtitle('UMAP of Pfizer-vaccinated participants by symptoms', subtitle = 'Labels are total of V2 fatigue, heada
che, myalgya, and chills'); rm (plot)
```

#### UMAP of Pfizer-vaccinated participants by symptoms

Labels are total of V2 fatigue, headache, myalgya, and chills



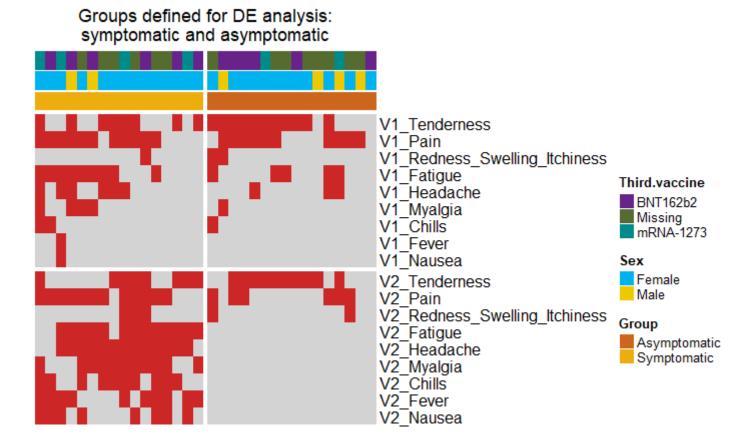
```
plot = data.frame(t(subset(meta_data, select = -c(id, Participant.ID, Run, SAGC_ID, Timepoint, First.Second.vacci
ne, Third.vaccine, Sex))))
umap = umap(as.matrix(apply(plot, 2, as.numeric)), config = custom.config); umap = data.frame(umap$layout)
plot$UMAP1 = umap$X1; plot$UMAP2 = umap$X2; rm(umap)
plot$Timepoint = ifelse(grepl('V1_', rownames(plot)), 'V1', 'V2')
plot$Symptom = gsub('V1_', '', rownames(plot)); plot$Symptom = gsub('V2_', '', plot$Symptom)
ggplot(plot, aes(x = UMAP1, y = UMAP2, label = Symptom, fill = Timepoint)) + geom_point(shape = 21, size = 4) + g
eom_text_repel() + theme_classic() + ggtitle('UMAP of symptoms, Pfizer-vaccinated participants')
```



```
rm(custom.config); rm(plot)
```

```
test = meta_data[apply(subset(meta_data, select = c(V2_Fatigue, V2_Headache, V2_Myalgia, V2_Chills, V2_Fever, V2_Nausea)) == 1, 1, sum) >= 3, ]; test$Group = "Symptomatic" control = subset(meta_data[apply(meta_data[, 18:ncol(meta_data)] == 1, 1, sum) <= 2, ], V2_Fatigue == 0 & V2_Myalgia == 0 & V2_Chills == 0 & V2_Headache == 0 & V2_Fever == 0 & V2_Nausea == 0); control$Group = "Asymptomatic"
```

```
meta_data = rbind(test, control); rm(test); rm(control)
count_data = count_data[, colnames(count_data) %in% meta_data$id]; meta_data = meta_data[order(meta_data$id), ]
count_data = count_data[, order(colnames(count_data))]; count_data = count_data[rownames(ens2gene), ]
count_data = data.frame(cbind(ensgene = rownames(count_data), count_data))
rownames(count_data) = NULL; rownames(meta_data) = NULL
```

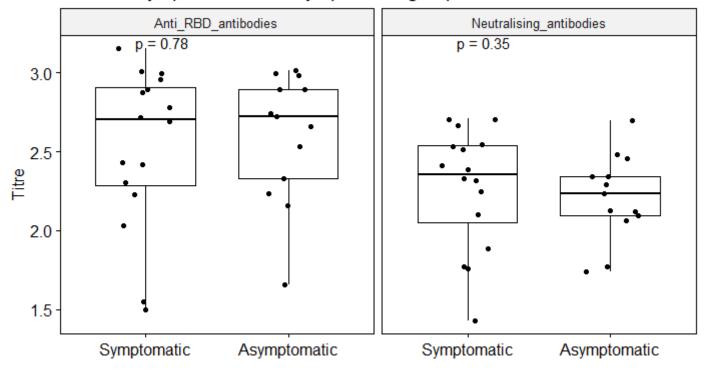


## 2. Antibodies and T cell responses

```
antibody_data = read.csv('data/antibodies.csv')
antibody_data = subset(antibody_data, Patient %in% meta_data$Participant.ID); meta_anti = meta_data[meta_data$Participant.ID %in% antibody_data$Patient, ]; antibody_data$Group = meta_anti$Group; rm(meta_anti)
```

 $ggboxplot(antibody\_data, x = 'Group', y = 'Titre', add = 'jitter', facet.by = 'Antibody.type') + stat\_compare\_mea \\ ns(label = "p.format") + xlab('') + ggtitle('COVID antibody titres are not significantly different\nbetween symptomatic and asymptomatic groups at V2B')$ 

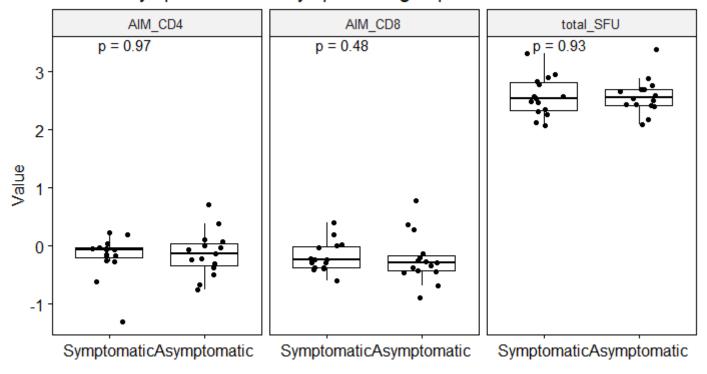
# COVID antibody titres are not significantly different between symptomatic and asymptomatic groups at V2B



```
tcell_data = read.csv('data/tcells.csv')
tcell_data = subset(tcell_data, Patient_ID %in% meta_data$Participant.ID); meta_tcell = meta_data[meta_data$Parti
cipant.ID %in% tcell_data$Patient_ID, ]; tcell_data$Group = meta_tcell$Group; rm(meta_tcell); tcell_data$Value =
log10(tcell_data$Value)
```

ggboxplot(tcell\_data, x = 'Group', y = 'Value', add = 'jitter', facet.by = 'Type') + stat\_compare\_means(label = "
p.format") + xlab('') + ggtitle('T cell responses are not significantly different\nbetween symptomatic and asympt
omatic groups at V2B')

# T cell responses are not significantly different between symptomatic and asymptomatic groups at V2B

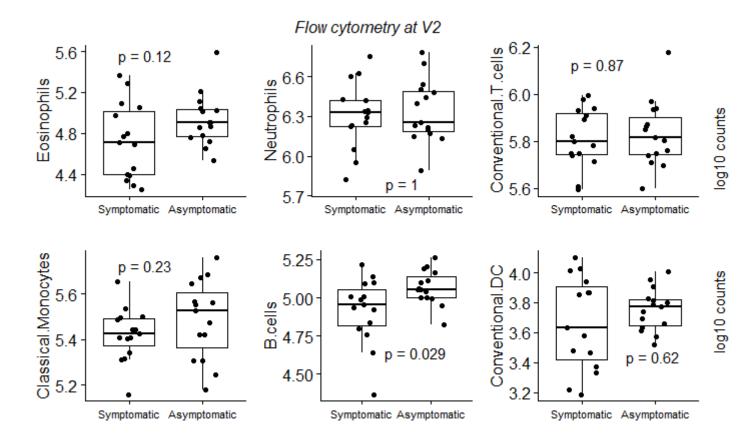


```
rm(antibody_data); rm(ha_participants); rm(tcell_data)
```

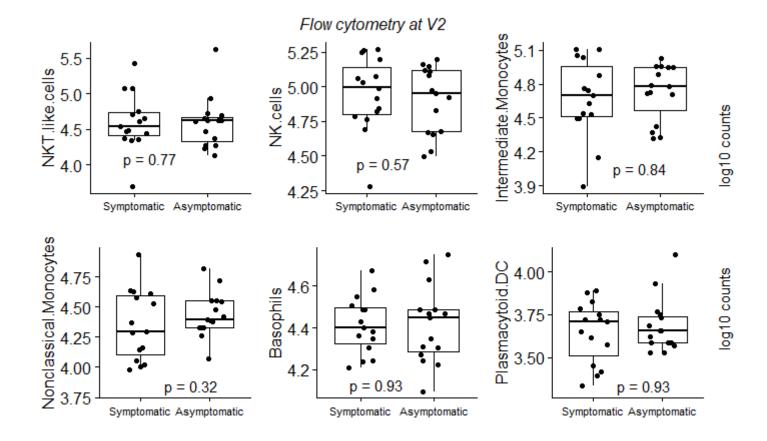
### 3. Flow cytometry

```
flow = read.csv('data/COVIRS_flow_data_20230202.csv'); flow = subset(flow, Patient %in% meta_data$Participant.ID
& Timepoint == 'V2A'); meta_flow = subset(meta_data, Participant.ID %in% flow$Patient); flow = flow[order(flow$Patient), ]; flow$Total = rowSums(flow[, 5:ncol(flow)-1]); flow$Group = meta_flow$Group;
```

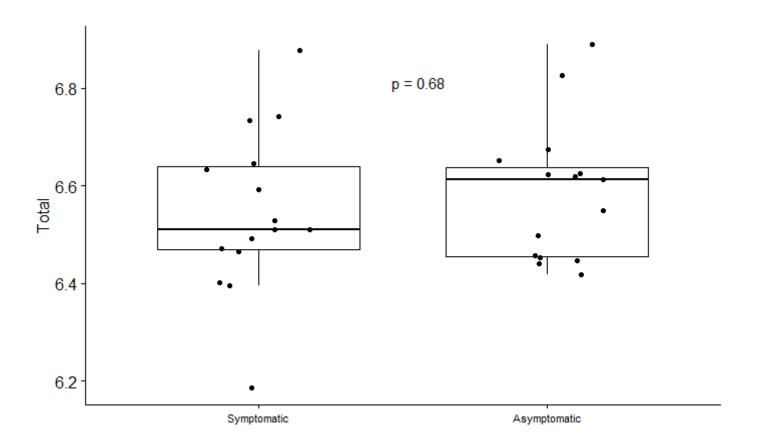
```
flow log = log10(flow[, 5:ncol(flow) - 1]); flow log$Group = meta flow$Group
plot1 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq) + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], ad
"p.format", label.x = 1, label.y = 5.5) + xlab('') + theme(axis.text.x = element text(size = 8))
plot2 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[2], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[2], add = 'jitter') + stat compare means(label = ggboxplot(flow log) + 
"p.format", label.x = 1.5, label.y = 5.75) + xlab('') + theme(axis.text.x = element text(size = 8))
plot3 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[3], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[3], add = 'jitter') + stat compare means(label = ggboxplot(flow log) = 
"p.format", label.x = 1, label.y = 6.1) + xlab('') + theme(axis.text.x = element text(size = 8))
plot6 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq) + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[6], ad
"p.format", label.x = 1, label.y = 5.7) + xlab('') + theme(axis.text.x = element text(size = 8))
plot9 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[9], add = 'jitter') + stat compare means(label = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[9], add = 'jitter') + stat compare means(label = qqboxplot(flow loq) = qqboxplot(flow loq) + stat compare means(label = qqboxplot(flow loq)) + stat compare means(label = qqbox
plot11 = qqboxplot(flow log, x = 'Group', y = colnames(flow log)[11], add = 'jitter') + stat compare means(label)
= "p.format", label.x = 1.75, label.y = 3.4) + xlab('') + theme(axis.text.x = element text(size = 8))
grid.arrange(plot1, plot2, plot3, plot6, plot9, plot11, ncol = 3, top = textGrob('Flow cytometry at V2', qp = qpa
r(fontface = 3)), right = textGrob('log10 counts
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                    log10 counts', rot = 90, qp = qpar(fontsize =
11)))
```



```
plot4 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[4], add = 'jitter') + stat compare means(label = loque)[4]
"p.format", label.x = 1, label.y = 4.0) + xlab('') + theme(axis.text.x = element text(size = 8))
plot5 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[5], add = 'jitter') + stat compare means(label = loque)[5]
"p.format", label.x = 1, label.y = 4.4) + xlab('') + theme(axis.text.x = element text(size = 8))
plot7 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[7], add = 'jitter') + stat compare means(label = 1)
"p.format", label.x = 1.5, label.y = 4.0) + xlab('') + theme(axis.text.x = element text(size = 8))
plot8 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add =
plot10 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[10], add = 'jitter') + stat compare means(label)
= "p.format", label.x = 1, label.y = 4.1) + xlab('') + theme(axis.text.x = element text(size = 8))
plot12 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[12], add = 'jitter') + stat compare means(label)
= "p.format", label.x = 1.5, label.y = 3.3) + xlab('') + theme(axis.text.x = element text(size = 8))
grid.arrange(plot4, plot5, plot7, plot8, plot10, plot12, ncol = 3, top = textGrob('Flow cytometry at V2', gp = gp
ar(fontface = 3)), right = textGrob('log10 counts
                                                                                                                                                           log10 counts', rot = 90, gp = gpar(fontsize
= 11)))
```

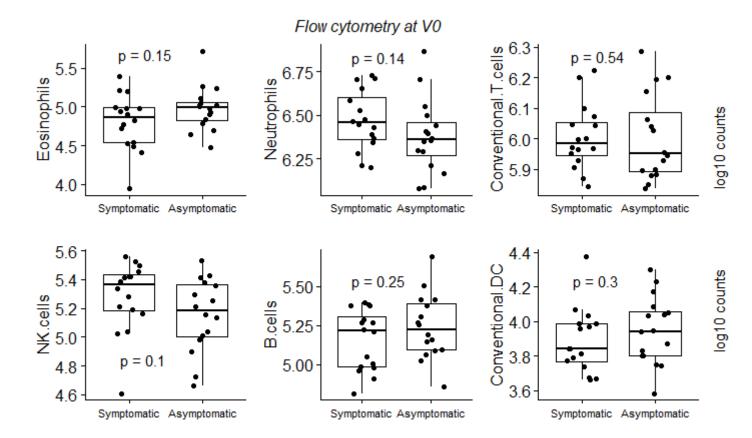


```
ggboxplot(flow_log, x = 'Group', y = 'Total', add = 'jitter') + stat_compare_means(label = "p.format", label.x = 1.5, label.y = 6.8) + xlab('') + theme(axis.text.x = element_text(size = 8))
```

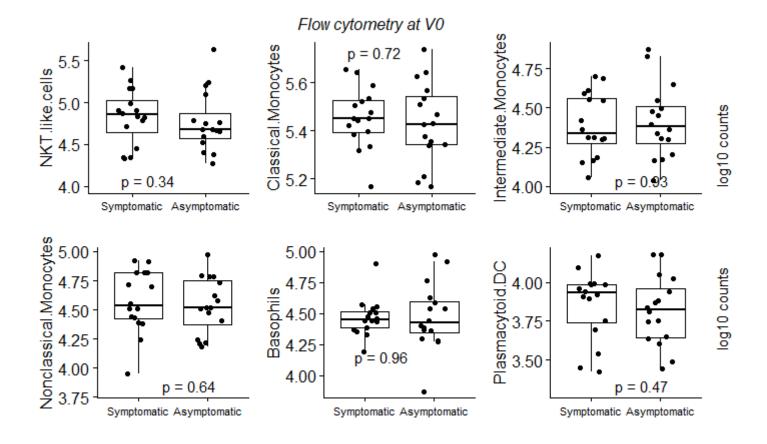


flow = read.csv('data/COVIRS\_flow\_data\_20230202.csv'); flow = subset(flow, Patient %in% meta\_data\$Participant.ID
& Timepoint == 'V0'); meta\_flow = subset(meta\_data, Participant.ID %in% flow\$Patient); flow = flow[order(flow\$Patient), ]; flow\$Total = rowSums(flow[, 5:ncol(flow)-1]); flow\$Group = meta\_flow\$Group

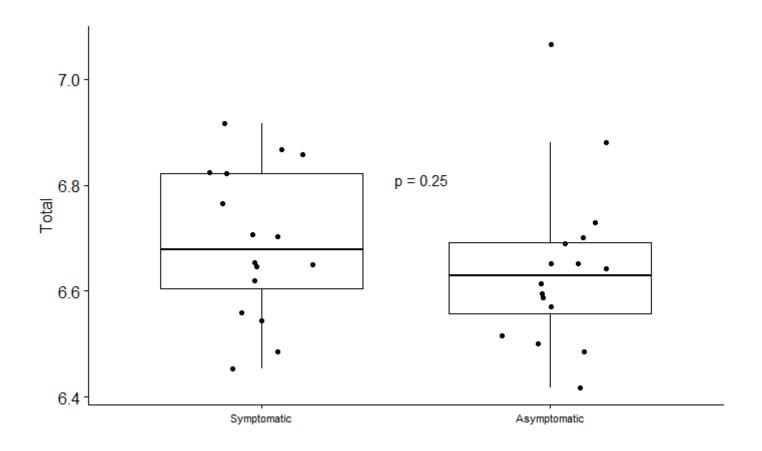
```
flow log = log10(flow[, 5:ncol(flow) - 1]); flow log$Group = meta flow$Group
plot1 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq) + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], add = 'jitter') + stat compare means(label = qqboxplot(flow loq))[1], ad
"p.format", label.x = 1, label.y = 5.6) + xlab('') + theme(axis.text.x = element text(size = 8))
plot2 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[2], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[2], add = 'jitter') + stat compare means(label = ggboxplot(flow log) + 
"p.format", label.x = 1, label.y = 6.8) + xlab('') + theme(axis.text.x = element text(size = 8))
plot3 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[3], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[3], add = 'jitter') + stat compare means(label = ggboxplot(flow log) = 
"p.format", label.x = 1, label.y = 6.25) + xlab('') + theme(axis.text.x = element text(size = 8))
plot5 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[5], add = 'jitter') + stat compare means(label = loque)[5]
"p.format", label.x = 1, label.y = 4.8) + xlab('') + theme(axis.text.x = element text(size = 8))
plot9 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[9], add = 'jitter') + stat compare means(label = loque)[9]
"p.format", label.x = 1, label.y = 5.5) + xlab('') + theme(axis.text.x = element text(size = 8))
plot11 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[11], add = 'jitter') + stat compare means(label)
= "p.format", label.x = 1, label.y = 4.2) + xlab('') + theme(axis.text.x = element text(size = 8))
grid.arrange(plot1, plot2, plot3, plot5, plot9, plot11, ncol = 3, top = textGrob('Flow cytometry at V0', qp = qpa
r(fontface = 3)), right = textGrob('log10 counts
                                                                                                                                                                                                                                                                                                                                  log10 counts', rot = 90, qp = qpar(fontsize =
11)))
```



```
plot4 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[4], add = 'jitter') + stat compare means(label = loque)[4]
"p.format", label.x = 1, label.y = 4.0) + xlab('') + theme(axis.text.x = element text(size = 8))
plot6 = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq, x = 'Group', y = colnames(flow loq)[6], add = 'jitter') + stat compare means(label = qqboxplot(flow loq) = 
"p.format", label.x = 1, label.y = 5.7) + xlab('') + theme(axis.text.x = element text(size = 8))
plot7 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[7], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[7], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[7], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = ggboxplot(flow log) = ggb
"p.format", label.x = 1.5, label.y = 4.0) + xlab('') + theme(axis.text.x = element text(size = 8))
plot8 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add = 'jitter') + stat compare means(label = ggboxplot(flow log)[8], add =
plot10 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[10], add = 'jitter') + stat compare means(label)
= "p.format", label.x = 1, label.y = 4.1) + xlab('') + theme(axis.text.x = element text(size = 8))
plot12 = ggboxplot(flow log, x = 'Group', y = colnames(flow log)[12], add = 'jitter') + stat compare means(label)
= "p.format", label.x = 1.5, label.y = 3.3) + xlab('') + theme(axis.text.x = element text(size = 8))
grid.arrange(plot4, plot6, plot7, plot8, plot10, plot12, ncol = 3, top = textGrob('Flow cytometry at V0', gp = gp
ar(fontface = 3)), right = textGrob('log10 counts
                                                                                                                                                                                                                                                                                                                                              log10 counts', rot = 90, gp = gpar(fontsize
= 11)))
```



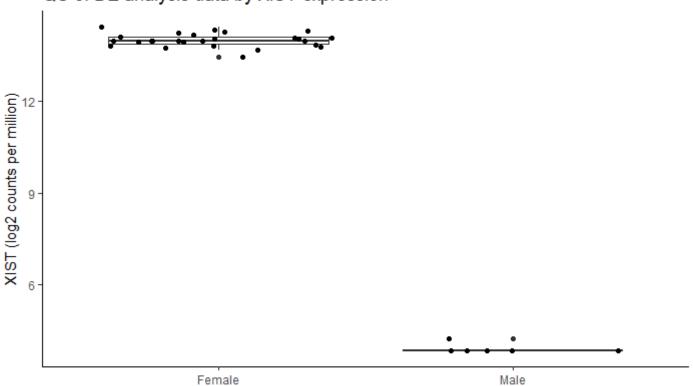
```
ggboxplot(flow_log, x = 'Group', y = 'Total', add = 'jitter') + stat_compare_means(label = "p.format", label.x = 1.5, label.y = 6.8) + xlab('') + theme(axis.text.x = element_text(size = 8))
```



```
rm(flow); rm(meta_flow); rm(flow_log); rm(plot4); rm(plot6); rm(plot7); rm(plot8); rm(plot10); rm(plot12); rm(plot12); rm(plot3); rm(plot5); rm(plot9); rm(plot11)
```

## 4. V2 gene set enrichment analysis

### QC of DE analysis data by XIST expression

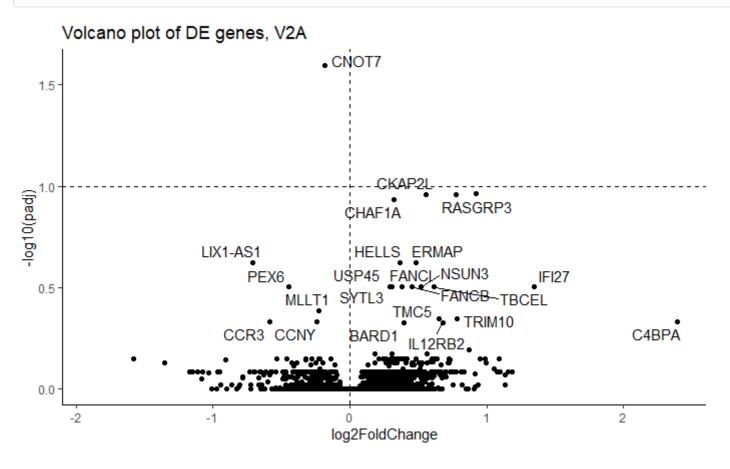


```
vsd = vst(dds, blind = FALSE); vsd.df = assay(vsd)
write.csv(vsd.df, file = 'results/counts norm v2.csv', row.names = TRUE)
keep = rowSums(counts(dds)>100) >= 4; dds = dds[keep, ]; dds = DESeq(dds)
estimating size factors
estimating dispersions
gene-wise dispersion estimates
mean-dispersion relationship
final dispersion estimates
fitting model and testing
-- replacing outliers and refitting for 20 genes
-- DESeq argument 'minReplicatesForReplace' = 7
-- original counts are preserved in counts(dds)
estimating dispersions
fitting model and testing
res = results(dds, contrast = c("Group", "Symptomatic", "Asymptomatic")); res = res[order(res$padj, decreasing =
FALSE), ]
res; summary(res)
```

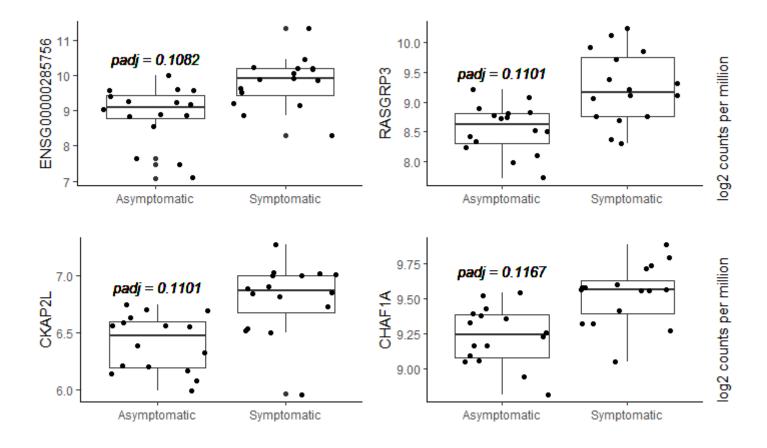
```
log2 fold change (MLE): Group Symptomatic vs Asymptomatic
Wald test p-value: Group Symptomatic vs Asymptomatic
DataFrame with 11403 rows and 6 columns
                baseMean log2FoldChange
                                            lfcSE
                                                                    pvalue
                                                          stat
                                                                                padj
                              <numeric> <numeric>
                                                     <numeric>
                                                                <numeric> <numeric>
               <numeric>
ENSG00000198791 1120.7692
                                                      -4.73151 2.22853e-06 0.025403
                              -0.183659 0.0388160
ENSG00000285756 750.6775
                               0.923212 0.2158807
                                                       4.27649 1.89861e-05 0.108211
ENSG00000152689 499.4901
                               0.774705 0.1856994
                                                       4.17182 3.02170e-05 0.110068
ENSG00000169607 73.5999
                               0.560044 0.1360796
                                                       4.11556 3.86239e-05 0.110068
ENSG00000167670 653.8076
                               0.322582 0.0796486
                                                       4.05006 5.12048e-05 0.116737
                     . . .
                                     . . .
                                                                       . . .
ENSG00000089048
                 319.702
                            3.87279e-06 0.0633473 6.11359e-05
                                                                  0.999951 0.999982
                1044.569
                           -1.88696e+00 0.6090192 -3.09836e+00
ENSG00000207005
                                                                                  NA
                                                                        NA
ENSG00000184500
                 222.258
                            -1.78817e-01 0.3468684 -5.15518e-01
                                                                        NA
                                                                                  NA
                  25.276
ENSG00000078114
                           -1.59377e+00 0.5449411 -2.92467e+00
                                                                        NA
                                                                                  NA
                 470.830
                            8.18822e-03 0.2855543 2.86748e-02
                                                                                  NA
ENSG00000107566
                                                                        NA
out of 11403 with nonzero total read count
adjusted p-value < 0.1
LFC > 0 (up)
                  : 0, 0%
LFC < 0 (down) : 1, 0.0088%
               : 4, 0.035%
outliers [1]
low counts [2]
               : 0, 0%
(mean count < 25)
[1] see 'cooksCutoff' argument of ?results
[2] see 'independentFiltering' argument of ?results
```

```
res.df = data.frame(res); res.df = data.frame(res.df, ens2gene[rownames(res.df), ])
write.csv(res.df, file = 'results/res_deseq_v2.csv', row.names = TRUE)
rm(keep)
```

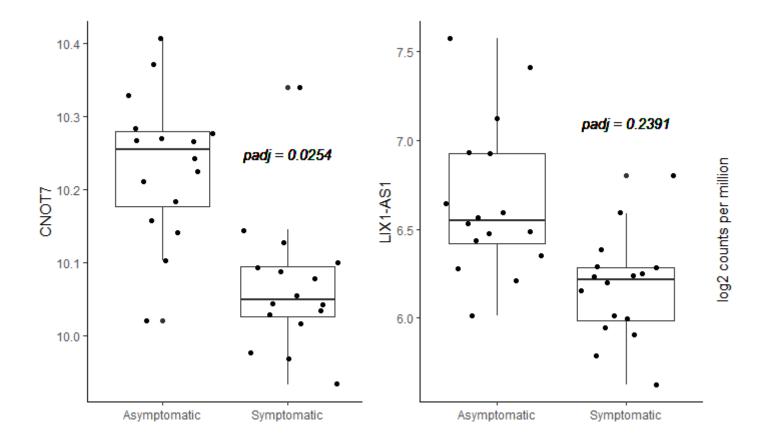
```
plot = subset(res.df, padj < 0.8)
ggplot(res.df, aes(x = log2FoldChange, y = -log10(padj))) + geom_point() +
   geom_hline(yintercept = -log10(0.1), linetype = 2) + geom_vline(xintercept = 0, linetype = 2) +
   theme_classic() + geom_text_repel(max.overlaps = 13, data = subset(plot, -log10(padj) > 0.2), aes(x = log2FoldChange, y = -log10(padj), label = external_gene_name)) + ggtitle("Volcano plot of DE genes, V2A"); rm(plot)
```



```
plot = rbind(filter(res.df, log2FoldChange > 0)[1:4, ], filter(res.df, log2FoldChange < 0)[1:2, ])</pre>
vsd.df = data.frame(t(vsd.df)); vsd.df$Group = meta data$Group; vsd.df$Sex = meta data$Sex
plot1 = qqplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[1]])) + qeom boxplot() + qeom jitter() + ylab(r
es.df[rownames(plot)[1], 'ensembl gene id']) + geom text(aes(x = 1, y = 10.5, fontface = 3), label = paste0('padj
= ', as.character(round(res.df[rownames(plot)[1], 'padj'], 4)))) + theme classic() + xlab('')
plot2 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[2]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[2], 'external gene name']) + geom text(aes(x = 1, y = 9.5, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[2], 'padj'], 4)))) + theme classic() + xlab('')
plot3 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[3]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[3], 'external gene name']) + geom text(aes(x = 1, y = 6.9, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[3], 'padj'], 4)))) + theme classic() + xlab('')
plot4 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[4]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[4], 'external gene name']) + geom text(aes(x = 1, y = 9.7, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[4], 'padj'], 4)))) + theme classic() + xlab('')
plot5 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[5]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[5], 'external gene name']) + geom text(aes(x = 2, y = 10.25, fontface = 3), label = paste0('
padj = ', as.character(round(res.df[rownames(plot)[5], 'padj'], 4)))) + theme classic() + xlab('')
plot6 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[6]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[6], 'external gene name']) + geom text(aes(x = 2, y = 7.1, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[6], 'padj'], 4)))) + theme classic() + xlab('')
grid.arrange(plot1, plot2, plot3, plot4, ncol = 2, right = textGrob('log2 counts per million
                                                                                                          log2 co
unts per million', rot = 90, qp = qpar(fontsize = 11)))
```



grid.arrange(plot5, plot6, ncol = 2, right = textGrob('log2 counts per million', rot = 90, gp = gpar(fontsize = 1
1)))



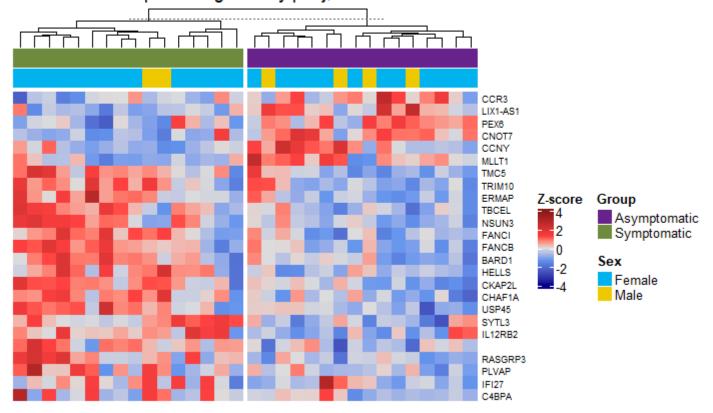
```
vsd.df$Sex = NULL; vsd.df$Group = NULL; vsd.df = data.frame(t(vsd.df))
rm(plot); rm(plot1); rm(plot2); rm(plot3); rm(plot4); rm(plot5); rm(plot6)
```

```
ha_top = HeatmapAnnotation(df = meta_data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show_annotation_na me = FALSE)

zzz = t(scale(t(vsd.df[rownames(res.df[1:25, ]), ]))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rown ames(zzz), 'external_gene_name']

Heatmap(zzz, show_row_names = TRUE, show_column_names = FALSE, cluster_rows = TRUE, cluster_columns = TRUE, show_row_dend = FALSE, top_annotation = ha_top, heatmap_legend_param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row_names_gp = gpar(fo ntsize = 7), column_title = 'Top 25 DE genes by padj, V2A', column_split = meta_data$Group)
```

#### Top 25 DE genes by padj, V2A



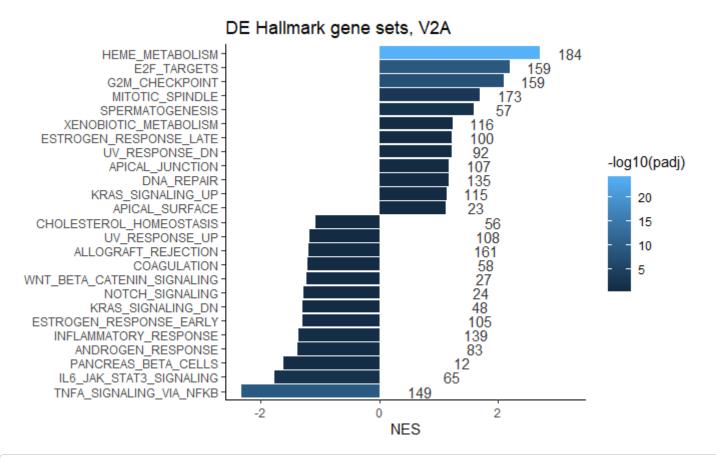
rm(ha\_top); rm(zzz)

```
ranks <- res.df %>%
  as_tibble() %>%
  dplyr::select(external_gene_name, log2FoldChange) %>%
  na.omit() %>%
  distinct() %>%
  group_by(external_gene_name) %>%
  summarize(log2FoldChange=mean(log2FoldChange))
ranks <- deframe(ranks)</pre>
```

```
msig.df = as.data.frame(msigdbr(species = "Homo sapiens", category = "H"))
msigDB = by(msig.df$gene_symbol, msig.df$gs_name, function(x) as.character(x))
names(msigDB) = gsub('HALLMARK_', '', names(msigDB))
fgseaRes = fgsea(pathways = msigDB, stats = ranks, nproc = 1); fgseaRes = fgseaRes[order(fgseaRes$padj, decreasin g = FALSE), ]
```

```
write.csv(subset(fgseaRes, select = -c(leadingEdge)), file = 'results/fgseaRes_H_v2.csv', row.names = FALSE)
```

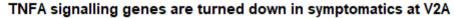
```
\begin{split} & \text{ggplot(fgseaRes[1:25, ], aes(y = reorder(pathway, NES), x = NES, fill = -log10(padj), label = size)) + \\ & \text{geom\_col()} + \text{geom\_text(color} = 'gray20', nudge\_x = ifelse(fgseaRes$NES > 0, 0.5, 3)) + \\ & \text{ylab('')} + \text{theme\_classic()} + \text{ggtitle('DE Hallmark gene sets, V2A')} \end{split}
```

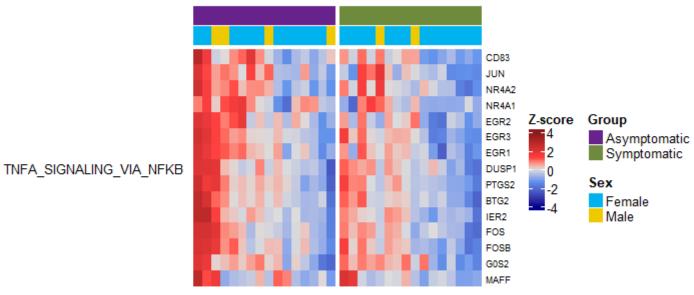


plotGseaTable(msigDB[fgseaRes[1:20, ][, pathway]], ranks, fgseaRes[1:20, ], gseaParam = 0.5, colwidths = c(3, 3, 0.8, 1.2, 1.2), pathwayLabelStyle = list(size = 10), headerLabelStyle = list(size = 10), valueStyle = list(size = 10))

Pathway HEME_METABOLISM	Gene ranks	NES 2.72	pval 7.2·10 <sup>-27</sup>	padj <sub>-25</sub> 3.6·10
TNFA SIGNALING VIA NFKB	But the contract of the contra	-2.32	2.3·10 <sup>-11</sup>	5.6·10 <sup>-10</sup>
E2F_TARGETS		2.20	5.6·10 <sup>-11</sup>	9.4·10 <sup>-10</sup>
G2M_CHECKPOINT		2.10	2.9·10 <sup>-9</sup>	3.6·10 <sup>-8</sup>
MITOTIC_SPINDLE		1.71	1.2·10 <sup>-4</sup>	1.2·10 <sup>-3</sup>
IL6_JAK_STAT3_SIGNALING		-1.77	2.0·10 <sup>-3</sup>	1.7·10 <sup>-2</sup>
SPERMATOGENESIS		1.60	5.6·10 <sup>-3</sup>	4.0·10 <sup>-2</sup>
INFLAMMATORY_RESPONSE		-1.37	1.8·10 <sup>-2</sup>	1.1·10 <sup>-1</sup>
ANDROGEN_RESPONSE		-1.38	4.1·10 <sup>-2</sup>	2.3·10 <sup>-1</sup>
ESTROGEN_RESPONSE_EARLY	1=1	-1.29	5.9·10 <sup>-2</sup>	2.8·10 <sup>-1</sup>
PANCREAS_BETA_CELLS	the second of the second	-1.61	6.2·10 <sup>-2</sup>	2.8·10 <sup>-1</sup>
ALLOGRAFT_REJECTION		-1.19	1.3·10 <sup>-1</sup>	4.0·10 <sup>-1</sup>
ESTROGEN_RESPONSE_LATE		1.23	1.2·10 <sup>-1</sup>	4.0·10 <sup>-1</sup>
KRAS_SIGNALING_DN		-1.29	1.1·10 <sup>-1</sup>	4.0·10 <sup>-1</sup>
UV_RESPONSE_DN	**************************************	1.22	1.2·10 <sup>-1</sup>	4.0·10 <sup>-1</sup>
XENOBIOTIC_METABOLISM	Home	1.24	1.1·10 <sup>-1</sup>	4.0·10 <sup>-1</sup>
COAGULATION	. In	-1.20	1.5·10 <sup>-1</sup>	4.2·10 <sup>-1</sup>
NOTCH_SIGNALING	the second secon	-1.28	1.5·10 <sup>-1</sup>	4.2·10 <sup>-1</sup>
APICAL_JUNCTION		1.18	1.8·10 <sup>-1</sup>	4.2·10 <sup>-1</sup>
DNA_REPAIR	100	1.18	1.7·10 <sup>-1</sup>	4.2·10 <sup>-1</sup>
	0 2500 5000 7500 10000			

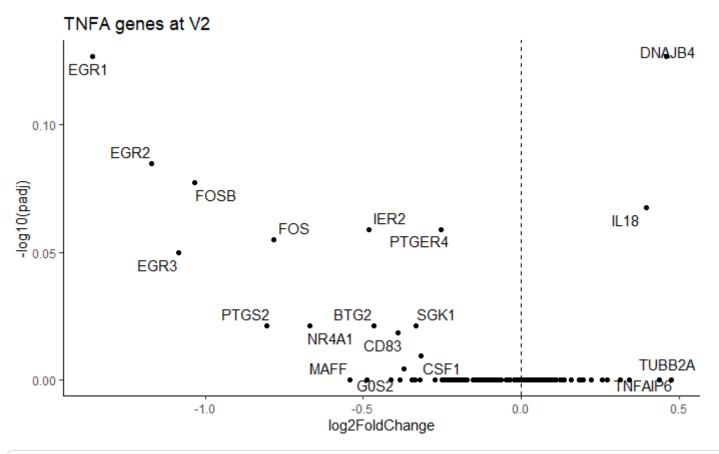
```
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
ha_top = HeatmapAnnotation(df = meta_data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show_annotation_na me = FALSE)
plot2 = vsd.df[rownames(subset(res.df, external_gene_name %in% msigDB[[fgseaRes[[2, 'pathway']]]] & log2FoldChang e < 0))[1:15], ]
zzz = t(scale(t(plot2))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external_gene_nam e']
plot2 = Heatmap(zzz, row_title = fgseaRes[[2, 'pathway']], show_row_names = TRUE, show_column_names = FALSE, clus ter_rows = TRUE, cluster_columns = TRUE, show_row_dend = FALSE, show_column_dend = FALSE, column_split = meta_dat a$Group, top_annotation = ha_top, heatmap_legend_param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row_names_gp = gpar(fontsize = 7), row_title_rot = 0, row_title_gp = gpar(fontsize = 10), column_title_gp = gpar(fontsize = 11, fontface = 2), column_title = 'TNFA signalling genes are turned down in symptomatics at V2A', heatmap_height = unit(8, "cm")); p lot2
```





```
rm(plot2); rm(ha_top); rm(zzz)

ggplot(subset(res.df, external_gene_name %in% msigDB[[fgseaRes[[2, 'pathway']]]]), aes(x = log2FoldChange, y = -l
og10(padj), label = external_gene_name)) + geom_point() + geom_text_repel() + theme_classic() + geom_vline(xinter
cept = 0, linetype = 2) + ggtitle('TNFA genes at V2')
```

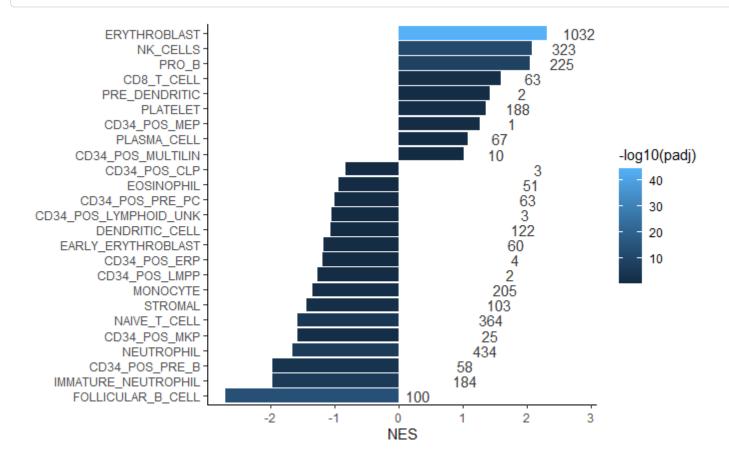


```
msig.df = as.data.frame(msigdbr(species = "Homo sapiens", category = "C8")); msig.df = msig.df[grepl('HAY_BONE_MA
RROW', msig.df$gs_name), ]
msigDB = by(msig.df$gene_symbol, msig.df$gs_name, function(x) as.character(x)); names(msigDB) = gsub('HAY_BONE_MA
RROW_', '', names(msigDB))
fgseaRes = fgsea(pathways = msigDB, stats = ranks, nproc = 1); fgseaRes = fgseaRes[order(fgseaRes$padj, decreasin
g = FALSE), ]
```

```
write.csv(subset(fgseaRes, select = -c(leadingEdge)), file = 'results/fgseaRes_C8_marrow_v2.csv', row.names = FAL
SE)

ggplot(fgseaRes[1:25, ], aes(y = reorder(pathway, NES), x = NES, fill = -log10(padj), label = size)) +
   geom_col() + geom_text(color = 'gray20', nudge_x = ifelse(fgseaRes$NES > 0, 0.5, 3)) +
   ylab('') + theme_classic()
```

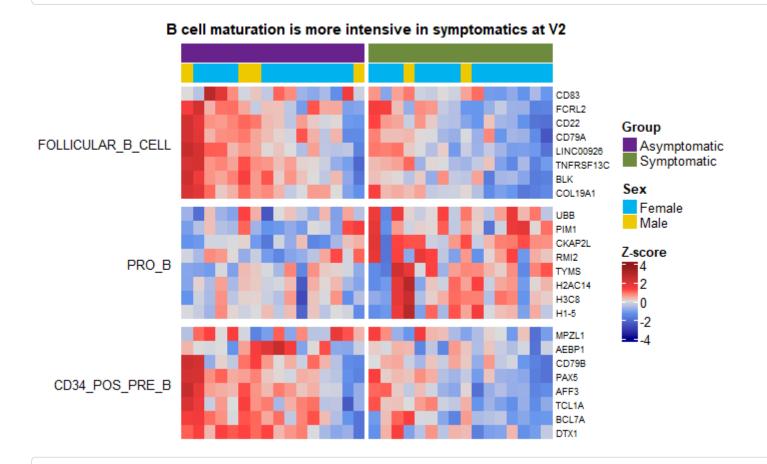
```
Warning in x + params$x :
  longer object length is not a multiple of shorter object length
```



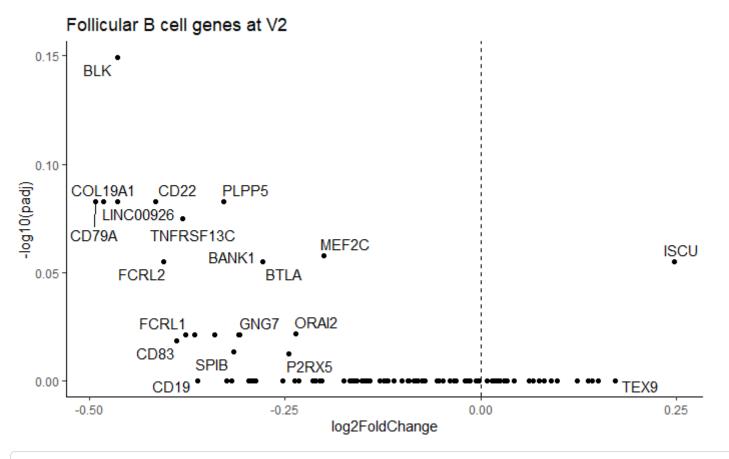
plotGseaTable(msigDB[fgseaRes[1:20, ][, pathway]], ranks, fgseaRes[1:20, ], gseaParam = 0.5, colwidths = c(3, 3, 0.8, 1.2, 1.2), pathwayLabelStyle = list(size = 10), headerLabelStyle = list(size = 10), valueStyle = list(size = 10))

Pathway ERYTHROBLAST	Gene ranks	NES 2.31	pval 10.0·10 <sup>-47</sup>	padj 2.9·10
FOLLICULAR_B_CELL		-2.71	1.9·10 <sup>-15</sup>	2.7·10 <sup>-14</sup>
NK_CELLS		2.07	1.1·10 <sup>-12</sup>	1.0·10 <sup>-11</sup>
PRO_B	<b></b>	2.05	2.6·10 <sup>-10</sup>	1.9·10 <sup>-9</sup>
IMMATURE_NEUTROPHIL	h an	-1.97	1.3·10 <sup>-7</sup>	7.3·10 <sup>-7</sup>
NEUTROPHIL	····	-1.66	1.7·10 <sup>-7</sup>	8.3·10 <sup>-7</sup>
NAIVE_T_CELL	IIII	-1.57	1.5·10 <sup>-5</sup>	6.4·10 <sup>-5</sup>
CD34_POS_PRE_B	transfer of the second	-1.97	1.0·10 <sup>-4</sup>	3.8·10 <sup>-4</sup>
CD8_T_CELL		1.59	8.7·10 <sup>-3</sup>	2.8·10 <sup>-2</sup>
MONOCYTE	1=:	-1.35	1.2·10 <sup>-2</sup>	3.4·10 <sup>-2</sup>
STROMAL		-1.44	1.3·10 <sup>-2</sup>	3.5·10 <sup>-2</sup>
CD34_POS_MKP	the management of the second of the second	-1.58	2.2·10 <sup>-2</sup>	5.4·10 <sup>-2</sup>
PLATELET		1.35	3.2·10 <sup>-2</sup>	7.2·10 <sup>-2</sup>
PRE_DENDRITIC		1.42	4.9·10 <sup>-2</sup>	1.0·10 <sup>-1</sup>
CD34_POS_MEP	•	1.26	1.1·10 <sup>-1</sup>	2.1·10 <sup>-1</sup>
CD34_POS_LMPP	1	-1.27	1.6·10 <sup>-1</sup>	2.9·10 <sup>-1</sup>
EARLY_ERYTHROBLAST		-1.17	1.9·10 <sup>-1</sup>	3.3·10 <sup>-1</sup>
DENDRITIC_CELL	h	-1.07	2.9·10 <sup>-1</sup>	4.7·10 <sup>-1</sup>
CD34_POS_ERP	the second second second second	-1.19	3.2·10 <sup>-1</sup>	4.8·10 <sup>-1</sup>
PLASMA_CELL	None of the product of the second of the second	1.08	3.5·10 <sup>-1</sup>	5.1·10 <sup>-1</sup>
	0 2500 5000 7500 10000			

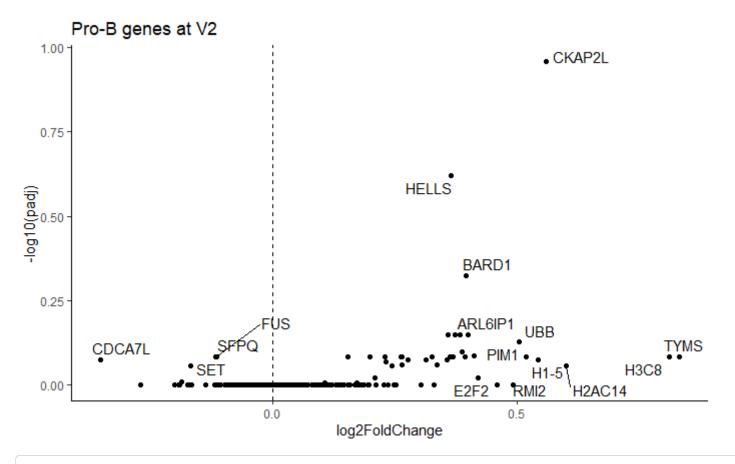
```
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
ha top = HeatmapAnnotation(df = meta data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr
een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show annotation na
me = FALSE)
plot1 = vsd.df[rownames(subset(res.df, external gene name %in% msigDB[[fgseaRes[[2, 'pathway']]]]] & log2FoldChang
e < 0))[1:8],
zzz = t(scale(t(plot1))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot1 = Heatmap(zzz, row title = fgseaRes[[2, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = FALSE, column split = meta dat
a$Group, top annotation = ha top, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25,
0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsize
= 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11), fontface = 2),
column title = 'B cell maturation is more intensive in symptomatics at V2')
res.df = res.df[order(res.df$log2FoldChange, decreasing = TRUE), ]
plot2 = vsd.df[rownames(subset(res.df, external gene name %in% msiqDB[[fgseaRes[[4, 'pathway']]]]] & log2FoldChang
e > 0))[1:8],
zzz = t(scale(t(plot2))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot2 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[4, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface =
2))
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
plot3 = vsd.df[rownames(subset(res.df, external gene name %in% msigDB[[fgseaRes[[8, 'pathway']]]] & log2FoldChang
e < 0))[1:8],
zzz = t(scale(t(plot3))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot3 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[8, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data\$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2))
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface =
2))
draw(plot1 %v% plot2 %v% plot3); rm(plot1); rm(plot2); rm(plot3); rm(zzz); rm(ha top)
```



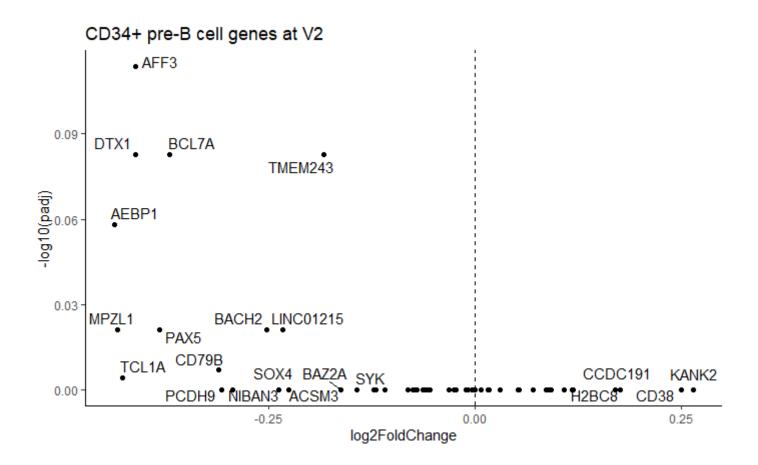
 $ggplot(subset(res.df, external\_gene\_name \%in\% \ msigDB[[fgseaRes[[2, 'pathway']]]]), \ aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + geom\_vline(xinter cept = 0, linetype = 2) + ggtitle('Follicular B cell genes at V2')$ 



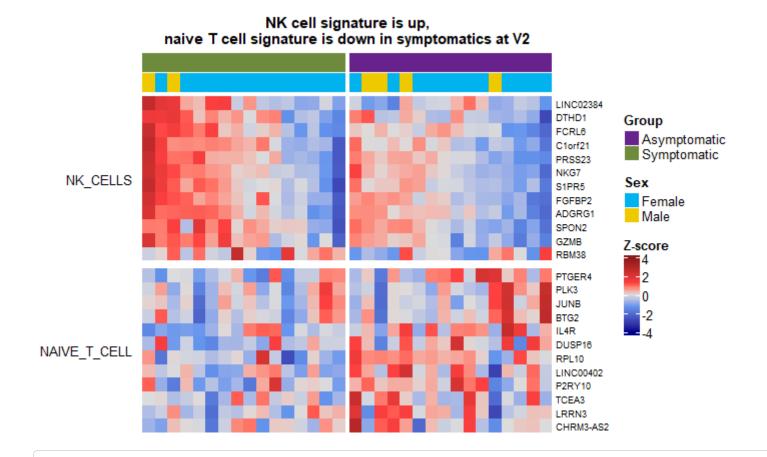
 $ggplot(subset(res.df, external\_gene\_name %in% msigDB[[fgseaRes[[4, 'pathway']]]]), aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom_text_repel() + theme_classic() + geom_vline(xinter cept = 0, linetype = 2) + ggtitle('Pro-B genes at V2')$ 



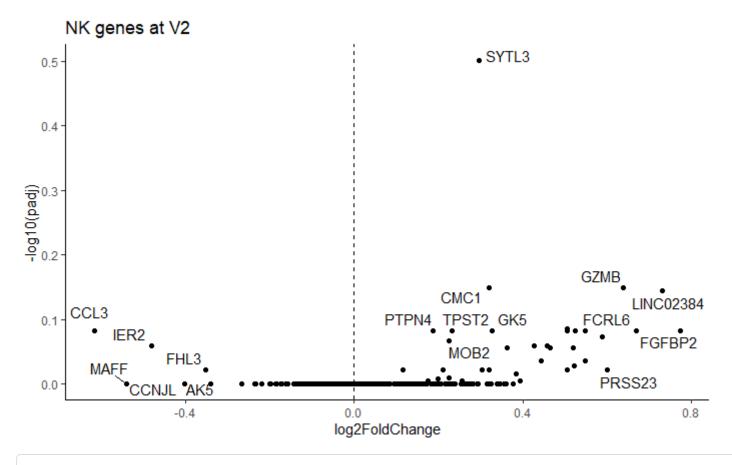
 $ggplot(subset(res.df, external\_gene\_name %in% msigDB[[fgseaRes[[8, 'pathway']]]]), aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + geom\_vline(xinter cept = 0, linetype = 2) + ggtitle('CD34+ pre-B cell genes at V2')$ 



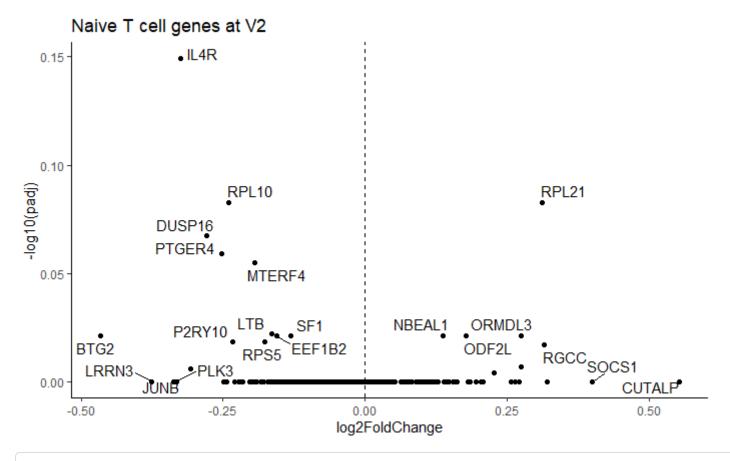
```
res.df = res.df[order(res.df$log2FoldChange, decreasing = TRUE), ]
ha top = HeatmapAnnotation(df = meta data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr
een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show annotation na
me = FALSE)
plot1 = vsd.df[rownames(subset(res.df, external gene name %in% msigDB[[fgseaRes[[3, 'pathway']]]] & log2FoldChang
e > 0))[1:12],
zzz = t(scale(t(plot1))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot1 = Heatmap(zzz, row title = fgseaRes[[3, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = FALSE, column split = meta dat
a$Group, top annotation = ha top, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25,
0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsize
= 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11), fontface = 2),
column title = 'NK cell signature is up,\nnaive T cell signature is down in symptomatics at V2')
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
plot2 = vsd.df[rownames(subset(res.df, external gene name %in% msiqDB[[fgseaRes[[7, 'pathway']]]] & log2FoldChang
e < 0))[1:12],
zzz = t(scale(t(plot2))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot2 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[7, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data\$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2))
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface =
2))
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
draw(plot1 %v% plot2); rm(plot1); rm(plot2); rm(zzz); rm(ha top)
```



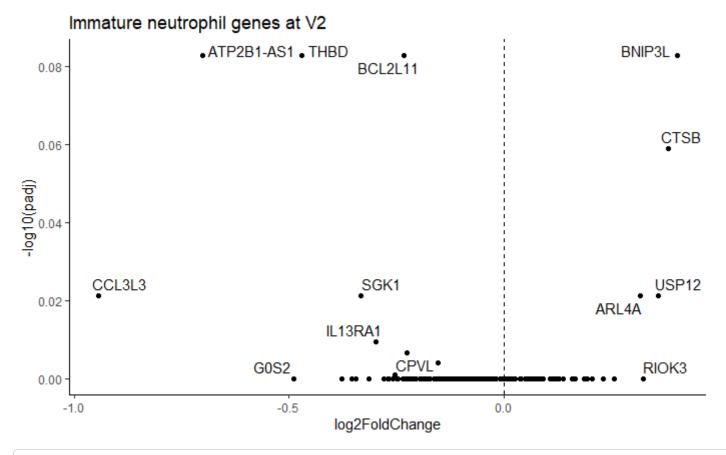
 $ggplot(subset(res.df, external\_gene\_name %in% msigDB[[fgseaRes[[3, 'pathway']]]]), aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom_text_repel() + theme_classic() + ggtitle('NK genes at V2') + geom_vline(xintercept = 0, linetype = 2)$ 



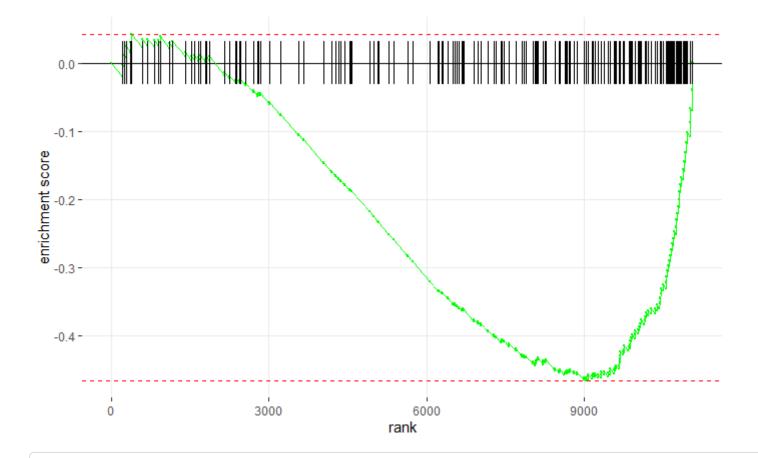
 $ggplot(subset(res.df, external\_gene\_name %in% msigDB[[fgseaRes[[7, 'pathway']]]]), aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom_text_repel() + theme_classic() + ggtitle('Naive T cell genes at V2') + geom_vline(xintercept = 0, linetype = 2)$ 



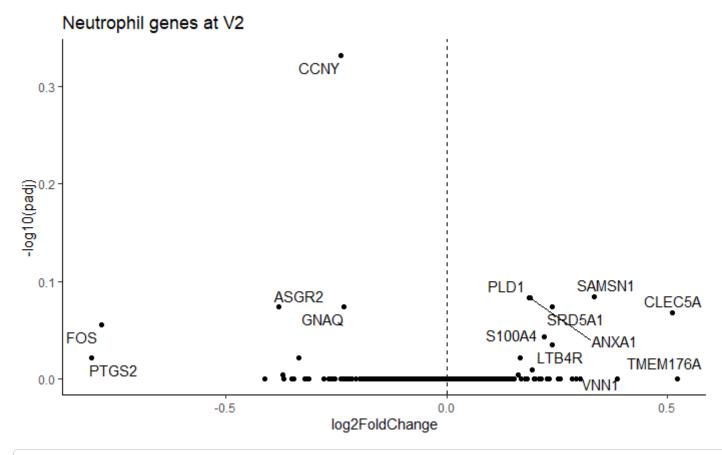
 $ggplot(subset(res.df, external\_gene\_name %in% \ msigDB[[fgseaRes[[5, 'pathway']]]]), \ aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Immature neutrophil genes at V2') + geom\_vline(xintercept = 0, linetype = 2)$ 



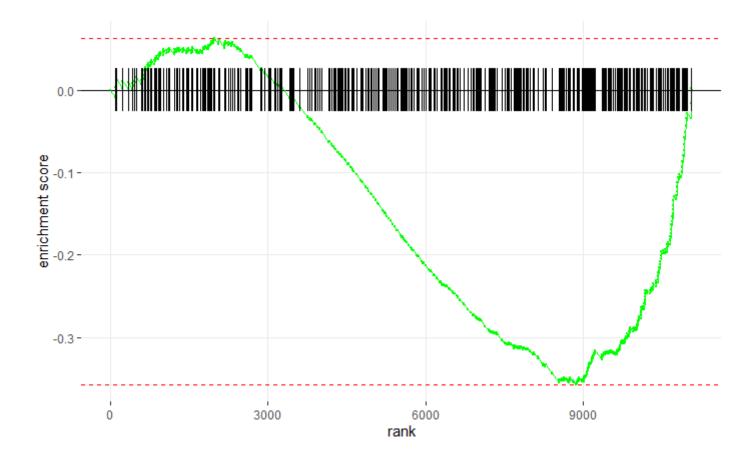
plotEnrichment(msigDB[[fgseaRes[[5, 'pathway']]]], ranks, gseaParam = 1, ticksSize = 0.2)



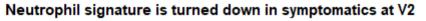
 $ggplot(subset(res.df, external\_gene\_name %in% msigDB[[fgseaRes[[6, 'pathway']]]]), aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Neutroph il genes at V2') + geom\_vline(xintercept = 0, linetype = 2)$ 

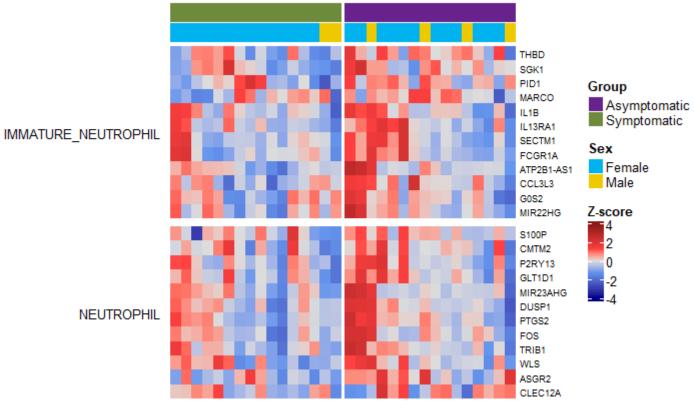


plotEnrichment(msigDB[[fgseaRes[[6, 'pathway']]]], ranks, gseaParam = 1, ticksSize = 0.2)



```
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
ha top = HeatmapAnnotation(df = meta data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr
een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show annotation na
me = FALSE)
plot1 = vsd.df[rownames(subset(res.df, external gene name %in% msigDB[[fgseaRes[[5, 'pathway']]]] & log2FoldChang
e < 0))[1:12],
zzz = t(scale(t(plot1))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot1 = Heatmap(zzz, row title = fgseaRes[[5, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = FALSE, column split = meta dat
a$Group, top annotation = ha top, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25,
0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsize
= 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11), fontface = 2),
column title = 'Neutrophil signature is turned down in symptomatics at V2')
plot2 = vsd.df[rownames(subset(res.df, external gene name %in% msiqDB[[fgseaRes[[6, 'pathway']]]]] & loq2FoldChang
e < 0))[1:12],
zzz = t(scale(t(plot2))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot2 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[6, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface =
2))
draw(plot1 %v% plot2); rm(plot1); rm(plot2); rm(zzz); rm(ha top)
```





btm = readRDS('data/Blood\_Transcript\_modules.RDS'); btm\_annot = read\_excel('data/btm\_annotation\_table.xls')

```
New names:

'Module size' -> 'Module size...4'

'Jaccard Index' -> 'Jaccard Index...10'

'Enrichment p value' -> 'Enrichment p value...11'

'Overlap size' -> 'Overlap size...13'

'Jaccard Index' -> 'Jaccard Index...15'

'Enrichment p value' -> 'Enrichment p value...16'

'Module size' -> 'Module size...17'

'Overlap size' -> 'Overlap size...18'

'Jaccard Index' -> 'Jaccard Index...20'

'Enrichment p value' -> 'Enrichment p value...21'

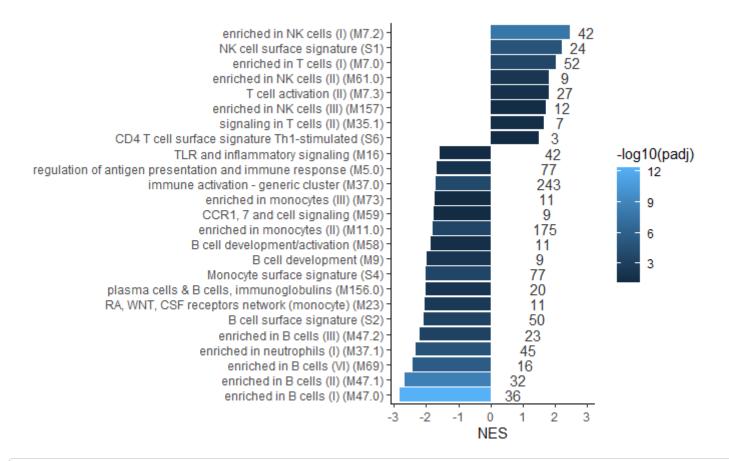
'Overlap size' -> 'Overlap size...23'
```

```
btm_annot = subset(btm_annot, `Module category` == 'immune'); btm = btm[c(btm_annot$ID)]; names(btm) = btm_anno
t$`Composite name`
fgseaRes = fgsea(pathways = btm, stats = ranks, nproc = 1); fgseaRes = fgseaRes[order(fgseaRes$padj, decreasing =
FALSE), ]
```

```
write.csv(subset(fgseaRes, select = -c(leadingEdge)), file = 'results/fgseaRes_btm_immune_v2.csv', row.names = FA
LSE)

fgseaRes = fgseaRes[pathway %in% subset(btm_annot, `Module category` == 'immune')$`Composite name`, ]
ggplot(fgseaRes[1:25, ], aes(y = reorder(pathway, NES), x = NES, fill = -log10(padj), label = size)) + geom_col()
+ geom_text(color = 'gray20', nudge_x = ifelse(fgseaRes$NES > 0, 0.5, 3.5)) + ylab('') + theme_classic()
```

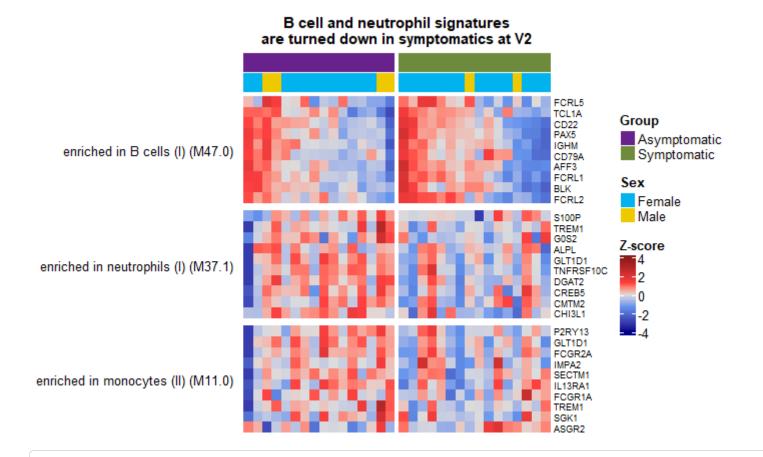
```
Warning in x + params$x : longer object length
```



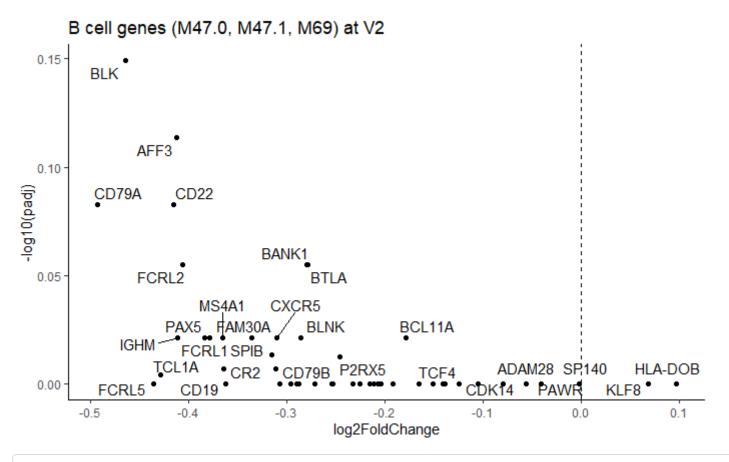
plotGseaTable(btm[fgseaRes[1:20, ][, pathway]], ranks, fgseaRes[1:20, ], gseaParam = 0.5, colwidths = c(6, 3, 0.
8, 1.2, 1.2), pathwayLabelStyle = list(size = 10), headerLabelStyle = list(size = 10), valueStyle = list(size = 10))

Pathway	Gene ranks NES	pval_ <sub>15</sub> 4.3·10 <sup>-15</sup>	padj_ <sub>13</sub> 4.5·10
enriched in B cells (I) (M47.0)	·	5.5·10 <sup>-11</sup>	2.8·10 <sup>-9</sup>
enriched in B cells (II) (M47.1)	-2.66		
enriched in NK cells (I) (M7.2)	· · · · · · · 2.45	5.8·10 <sup>-10</sup>	2.0.10-8
enriched in B cells (VI) (M69)	-2.42	1.1·10 <sup>-7</sup>	2.7·10 <sup>-8</sup>
NK cell surface signature (S1)	<b>■</b> 2.22	7.1·10 <sup>-7</sup>	1.5·10 <sup>-5</sup>
enriched in neutrophils (I) (M37.1)	-2.34	1.1·10 <sup>-8</sup>	1.8·10 <sup>-5</sup>
immune activation - generic cluster (M37.0)	<b></b> -1.70	1.1·10 <sup>-5</sup>	1.6·10 <sup>-4</sup>
enriched in T cells (I) (M7.0)	<b>*************************************</b>	1.8·10 <sup>-5</sup>	2.3-10 <sup>-4</sup>
enriched in monocytes (II) (M11.0)	-1.80	2.4·10 <sup>-5</sup>	2.8·10 <sup>-4</sup>
Monocyte surface signature (S4)	2.02	3.0·10 <sup>-5</sup>	3.1·10 <sup>-4</sup>
B cell surface signature (S2)	·····	6.8·10 <sup>-5</sup>	5.9·10 <sup>-4</sup>
enriched in B cells (III) (M47.2)	2.20	6.5·10 <sup>-5</sup>	5.9·10 <sup>-4</sup>
RA, WNT, CSF receptors network (monocyte) (M23)		4.7·10 <sup>-4</sup>	3.7·10 <sup>-3</sup>
enriched in NK cells (II) (M61.0)	1.80	8.7·10 <sup>-4</sup>	6.4·10 <sup>-3</sup>
B cell development (M9)	-1.97	1.1·10 <sup>-3</sup>	7.4·10 <sup>-3</sup>
plasma cells & B cells, immunoglobulins (M156.0)	-2.03	1.5·10 <sup>-3</sup>	9.8·10 <sup>-3</sup>
tion of antigen presentation and immune response (M5.0)	-1.68	2.7·10 <sup>-3</sup>	1.6·10 <sup>-2</sup>
T cell activation (II) (M7.3)	1.80	3.7·10 <sup>-3</sup>	2.1·10 <sup>-2</sup>
B cell development/activation (M58)	-1.86	5.6·10 <sup>-3</sup>	3.0·10 <sup>-2</sup>
signaling in T cells (II) (M35.1)	1.66	7.0·10 <sup>-3</sup>	3.6·10 <sup>-2</sup>
	0 2500 5000 7500 10000		

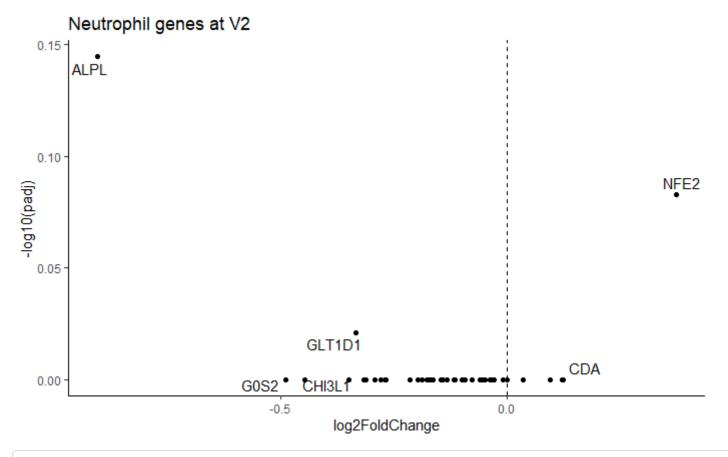
```
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
ha top = HeatmapAnnotation(df = meta data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr
een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show annotation na
me = FALSE)
plot1 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[1, 'pathway']]]] & log2FoldChange <</pre>
0))[1:10], ]
zzz = t(scale(t(plot1))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot1 = Heatmap(zzz, row title = fgseaRes[[1, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = FALSE, column split = meta dat
a$Group, top annotation = ha top, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25,
0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsize
= 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11), fontface = 2),
column title = 'B cell and neutrophil signatures\nare turned down in symptomatics at V2')
plot2 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[6, 'pathway']]]] & log2FoldChange <</pre>
0))[1:10], ]
zzz = t(scale(t(plot2))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot2 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[6, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data\$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2))
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11, fontface =
2))
plot3 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[9, 'pathway']]]] & log2FoldChange <</pre>
0))[1:10], ]
zzz = t(scale(t(plot3))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot3 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[9, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data\$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2))
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface =
2))
draw(plot1 %v% plot2 %v% plot3)
```



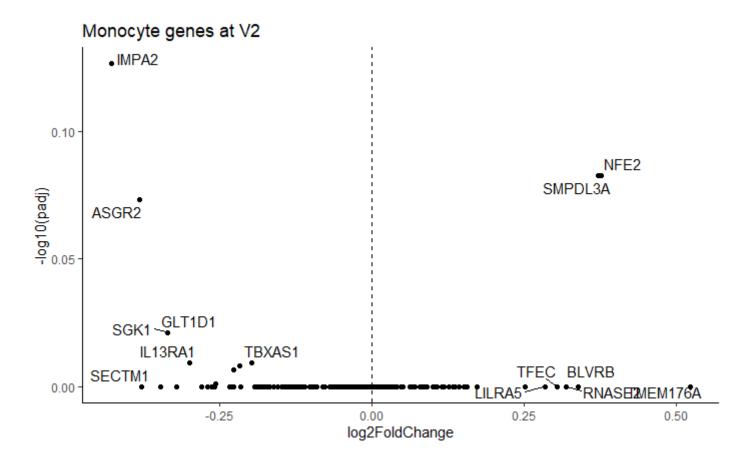
ggplot(subset(res.df, external\_gene\_name %in% btm[[fgseaRes[[1, 'pathway']]]] | external\_gene\_name %in% btm[[fgseaRes[[2, 'pathway']]]] | external\_gene\_name %in% btm[[fgseaRes[[4, 'pathway']]]]), aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('B cell g enes (M47.0, M47.1, M69) at V2') + geom\_vline(xintercept = 0, linetype = 2)



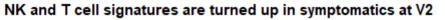
ggplot(subset(res.df, external\_gene\_name %in% btm[[fgseaRes[[6, 'pathway']]]]), aes(x = log2FoldChange, y = -log1
0(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Neutrophil
genes at V2') + geom\_vline(xintercept = 0, linetype = 2)

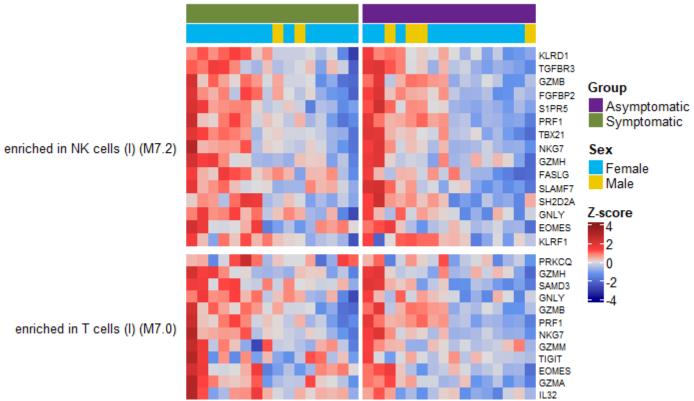


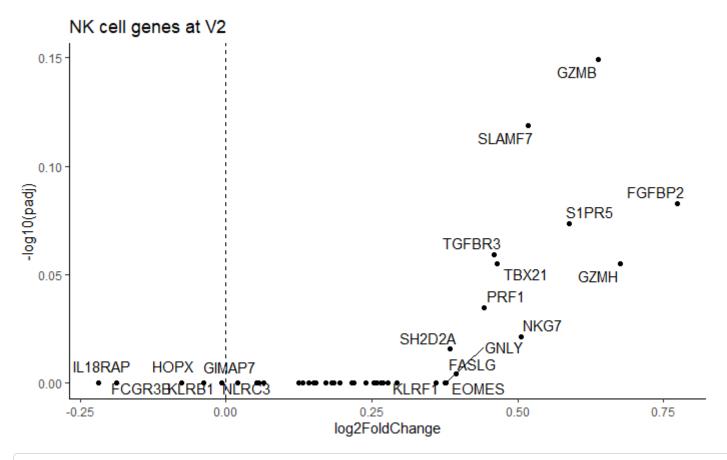
ggplot(subset(res.df, external\_gene\_name %in% btm[[fgseaRes[[9, 'pathway']]]]), aes(x = log2FoldChange, y = -log1
0(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Monocyte ge
nes at V2') + geom\_vline(xintercept = 0, linetype = 2)

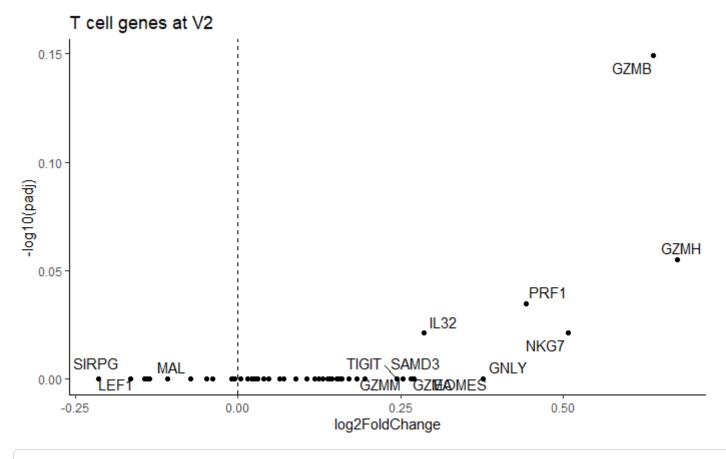


```
res.df = res.df[order(res.df$log2FoldChange, decreasing = TRUE), ]
plot4 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[3, 'pathway']]]] & log2FoldChange >
0))[1:15], ]
zzz = t(scale(t(plot4))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot4 = Heatmap(zzz, row title = fgseaRes[[3, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = FALSE, column split = meta dat
a$Group, top annotation = ha top, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25,
0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsize
= 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11), fontface = 2),
heatmap height = unit(7, 'cm'), column title = 'NK and T cell signatures are turned up in symptomatics at V2')
plot5 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[8, 'pathway']]]] & log2FoldChange >
0))[1:12], ]
zzz = t(scale(t(plot5))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e'1
plot5 = Heatmap(zzz, row title = fgseaRes[[8, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = TRUE, column split = meta dat
a$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25, 0, 1.25, 4), c("darkblue",
"cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names qp = qpar(fontsize = 7), row title rot = 0, ro
w title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface = 2), show heatmap legend = FALS
draw(plot4 %v% plot5)
```









```
rm(plot1); rm(plot2); rm(plot3); rm(plot4); rm(plot5); rm(zzz); rm(ha_top)
```

## 5. V0 GSEA

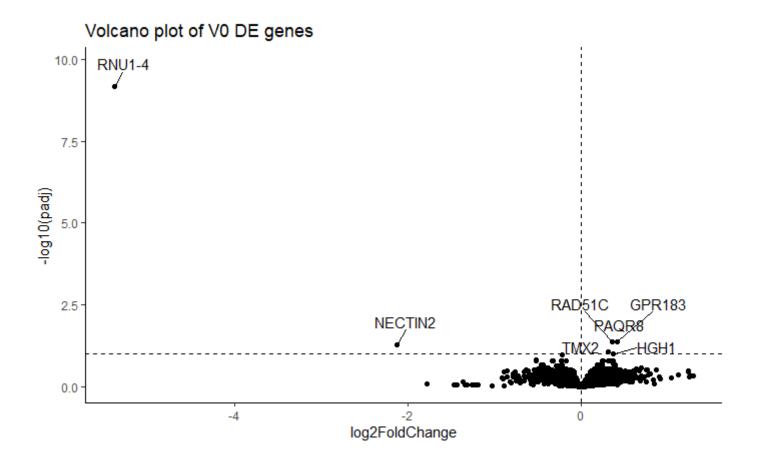
```
keep = data.frame(cbind(Participant.ID = meta_data$Participant.ID, Group = meta_data$Group))
```

```
count data = readRDS('data/Raw RNASeq Data.RDS')
ens2gene = readRDS("data/ens2gene.RDS")
meta data = data.frame(read.csv('data/COVIRS metadata.csv', header = TRUE))
meta data = meta data[meta data$Timepoint == 'V0' & meta data$First.Second.vaccine == 'BNT162b2', ]
meta data = meta data[meta data$Participant.ID %in% keep$Participant.ID, ]; meta data$Group = keep$Group
ensgene = rownames(count data); count data = count data[, colnames(count data) %in% meta data$id]; count data = d
ata.frame(cbind(ensgene = ensgene, count data)); rownames(count data) = NULL; rm(ensgene); rm(keep)
dds = DESeqDataSetFromMatrix(countData = count data,
                              colData = meta data,
                              design= ~ Group + Sex, tidy = TRUE)
vsd = vst(dds, blind = FALSE); vsd.df = assay(vsd); write.csv(vsd.df, file = 'results/counts norm v0.csv', row.na
mes = TRUE)
keep = rowSums(counts(dds) > 100) >= 4; dds = dds[keep, ]; dds = DESeg(dds)
estimating size factors
estimating dispersions
gene-wise dispersion estimates
mean-dispersion relationship
final dispersion estimates
fitting model and testing
-- replacing outliers and refitting for 19 genes
-- DESeq argument 'minReplicatesForReplace' = 7
-- original counts are preserved in counts(dds)
estimating dispersions
fitting model and testing
res = results(dds, contrast = c("Group", "Symptomatic", "Asymptomatic")); res = res[order(res$padj, decreasing =
FALSE), ]
res; summary(res); res.df = data.frame(res); res.df = data.frame(res.df, ens2gene[rownames(res.df), ]); write.csv
(res.df, file = 'results/res deseg v0.csv', row.names = TRUE)
```

```
log2 fold change (MLE): Group Symptomatic vs Asymptomatic
Wald test p-value: Group Symptomatic vs Asymptomatic
DataFrame with 11229 rows and 6 columns
                baseMean log2FoldChange
                                            lfcSE
                                                                  pvalue
                                                        stat
                                                                                padj
                              <numeric> <numeric> <numeric>
                                                               <numeric>
                <numeric>
                                                                           <numeric>
                 905.246
                               -5.386487 0.7176174
                                                    -7.50607 6.09286e-14 6.83863e-10
ENSG00000207389
                 313.277
ENSG00000170915
                               0.422574 0.0942592
                                                     4.48310 7.35649e-06 4.12846e-02
                 642.807
ENSG00000169508
                               0.421562 0.0966748
                                                     4.36061 1.29698e-05 4.24441e-02
ENSG00000108384
                 204.533
                               0.363181 0.0839366
                                                     4.32684 1.51262e-05 4.24441e-02
ENSG00000130202
                 125.110
                               -2.122512 0.5016004
                                                    -4.23148 2.32157e-05 5.21147e-02
                      . . .
                                     . . .
                                                         . . .
ENSG00000123838
                 41.6755
                              -3.2523776 0.624424 -5.2086061
                                                                      NA
                                                                                  NA
ENSG00000239899
                251.3011
                              1.1974009 0.575360 2.0811347
                                                                      NA
                                                                                  NA
ENSG00000233913
                206.2764
                              -0.0132661 0.346732 -0.0382605
                                                                      NA
                                                                                  NA
ENSG00000254612
                 75.9356
                              -0.1343210 0.199588 -0.6729914
                                                                      NA
                                                                                  NA
ENSG00000078114
                 23.3453
                              -0.7735165 0.495343 -1.5615775
                                                                                  NA
                                                                      NA
out of 11229 with nonzero total read count
adjusted p-value < 0.1
LFC > 0 (up)
                  : 5. 0.045%
LFC < 0 \text{ (down)} : 2, 0.018%
               : 5, 0.045%
outliers [1]
               : 0, 0%
low counts [2]
(mean count < 20)
[1] see 'cooksCutoff' argument of ?results
[2] see 'independentFiltering' argument of ?results
```

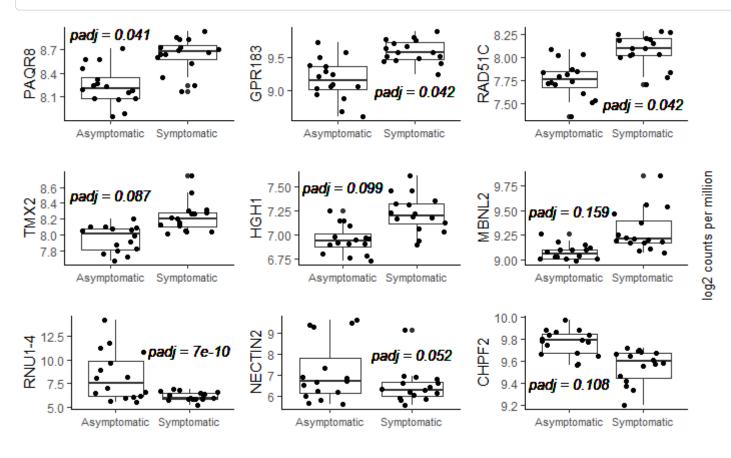
```
rm(keep)
```

```
ggplot(res.df, aes(x = log2FoldChange, y = -log10(padj))) + geom_point() +
  geom_hline(yintercept = -log10(0.1), linetype = 2) + geom_vline(xintercept = 0, linetype = 2) +
  theme_classic() + geom_text_repel(data = subset(res.df, -log10(padj) > 1), aes(x = log2FoldChange, y = -log10(padj), label = external_gene_name), max.overlaps = 9, position = position_nudge_repel(x = 0.1, y = 0.7)) + ggtitle
("Volcano plot of V0 DE genes")
```



```
plot = rbind(filter(res.df, log2FoldChange > 0)[1:6, ], filter(res.df, log2FoldChange < 0)[1:3, ])</pre>
vsd.df = data.frame(t(vsd.df)); vsd.df$Group = meta data$Group; vsd.df$Sex = meta data$Sex
plot1 = qqplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[1]])) + qeom boxplot() + qeom jitter() + ylab(r
es.df[rownames(plot)[1], 'external gene name']) + geom text(aes(x = 1, y = 8.9, fontface = 3), label = paste\theta('pa
dj = ', as.character(round(res.df[rownames(plot)[1], 'padj'], 3)))) + theme classic() + xlab('') + theme(axis.tex
t.x = element text(size = 8))
plot2 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[2]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[2], 'external gene name']) + geom text(aes(x = 2, y = 9.0, fontface = 3), label = paste\theta('pa
di = ', as.character(round(res.df[rownames(plot)[2], 'padj'], 3)))) + theme classic() + xlab('') + theme(axis.tex
t.x = element text(size = 8))
plot3 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[3]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[3], 'external gene name']) + geom text(aes(x = 2, y = 7.5, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[3], 'padj'], 3)))) + theme classic() + xlab('') + theme(axis.tex
t.x = element text(size = 8))
plot4 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[4]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[4], 'external gene name']) + geom text(aes(x = 1, y = 8.5, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[4], 'padj'], 3)))) + theme classic() + xlab('') + theme(axis.tex
t.x = element text(size = 8))
plot5 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[5]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[5], 'external gene name']) + geom text(aes(x = 1, y = 7.5, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[5], 'padj'], 3)))) + theme classic() + xlab('') + theme(axis.tex
t.x = element text(size = 8))
plot6 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[6]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[6], 'external gene name']) + geom text(aes(x = 1, y = 9.5, fontface = 3), label = paste0('pa
dj = ', as.character(round(res.df[rownames(plot)[6], 'padj'], 3)))) + theme classic() + xlab('') + theme(axis.tex
t.x = element text(size = 8))
plot7 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[7]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[7], 'external gene name']) + geom text(aes(x = 2, y = 11, fontface = 3), label = paste\theta('pad
j = ', as.character(round(res.df[rownames(plot)[7], 'padj'], 10)))) + theme classic() + xlab('') + theme(axis.tex
t.x = element text(size = 8))
plot8 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[8]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[8], 'external gene name']) + geom text(aes(x = 2, y = 8, fontface = 3), label = paste\theta('padj
= ', as.character(round(res.df[rownames(plot)[8], 'padj'], 3)))) + theme classic() + xlab('') + theme(axis.text.x
= element text(size = 8))
plot9 = ggplot(vsd.df, aes(x = Group, y = vsd.df[, rownames(plot)[9]])) + geom boxplot() + geom jitter() + ylab(r
es.df[rownames(plot)[9], 'external gene name']) + geom text(aes(x = 1, y = 9.4, fontface = 3), label = paste\theta('pa
```

```
dj = ', as.character(round(res.df[rownames(plot)[9], 'padj'], 3)))) + theme_classic() + xlab('') + theme(axis.tex
t.x = element_text(size = 8))
grid.arrange(plot1, plot2, plot3, plot4, plot5, plot6, plot7, plot8, plot9, ncol = 3, right = textGrob('log2 coun
ts per million', rot = 90, gp = gpar(fontsize = 10)))
```

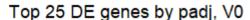


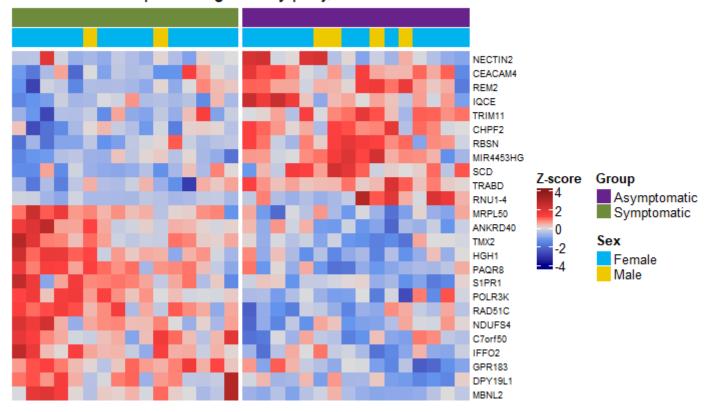
```
vsd.df$Sex = NULL; vsd.df$Group = NULL; vsd.df = data.frame(t(vsd.df))
rm(plot); rm(plot1); rm(plot2); rm(plot3); rm(plot4); rm(plot5); rm(plot6)
```

```
ha_top = HeatmapAnnotation(df = meta_data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr
een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show_annotation_na
me = FALSE)

zzz = t(scale(t(vsd.df[rownames(res.df[1:25, ]), ]))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rown
ames(zzz), 'external_gene_name']

Heatmap(zzz, show_row_names = TRUE, show_column_names = FALSE, cluster_rows = TRUE, cluster_columns = TRUE, show_
row_dend = FALSE, show_column_dend = FALSE, top_annotation = ha_top, heatmap_legend_param = list(title = "Z-scor
e"), col = colorRamp2(c(-4, -1.25, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick
4")), row_names_gp = gpar(fontsize = 7), column_title = 'Top 25 DE genes by padj, V0', column_split = meta_data$G
roup)
```





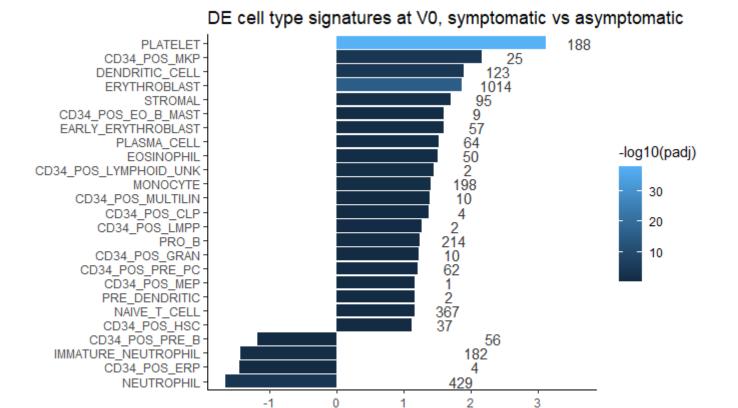
rm(ha\_top); rm(zzz)

```
ranks <- res.df %>%
  as_tibble() %>%
  dplyr::select(external_gene_name, log2FoldChange) %>%
  na.omit() %>%
  distinct() %>%
  group_by(external_gene_name) %>%
  summarize(log2FoldChange=mean(log2FoldChange))
ranks <- deframe(ranks)</pre>
```

```
msig.df = as.data.frame(msigdbr(species = "Homo sapiens", category = "C8")); msig.df = msig.df[grepl('HAY_BONE_MA
RROW', msig.df$gs_name), ]
msigDB = by(msig.df$gene_symbol, msig.df$gs_name, function(x) as.character(x)); names(msigDB) = gsub('HAY_BONE_MA
RROW_', '', names(msigDB))
fgseaRes = fgsea(pathways = msigDB, stats = ranks, nproc = 1); fgseaRes = fgseaRes[order(fgseaRes$padj, decreasin
g = FALSE), ]
```

```
write.csv(subset(fgseaRes, select = -c(leadingEdge)), file = 'results/fgseaRes_C8_marrow_v0.csv', row.names = FAL
SE)

ggplot(fgseaRes[1:25, ], aes(y = reorder(pathway, NES), x = NES, fill = -log10(padj), label = size)) +
    geom_col() + geom_text(color = 'gray20', nudge_x = ifelse(fgseaRes$NES > 0, 0.5, 3.5)) +
    ylab('') + theme_classic() + ggtitle('DE cell type signatures at V0, symptomatic vs asymptomatic')
```

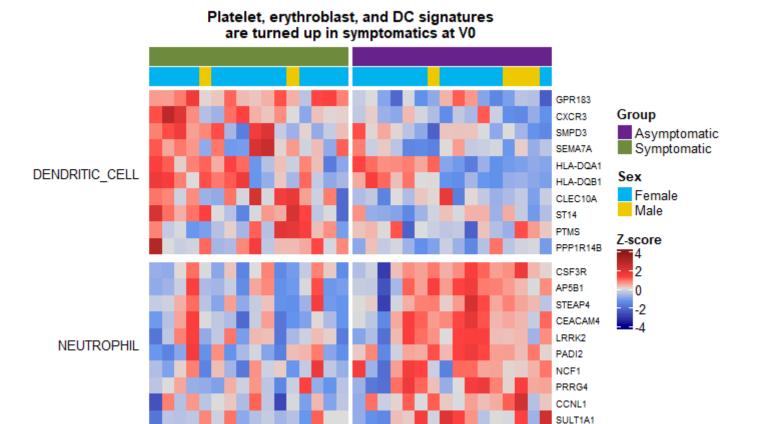


NES

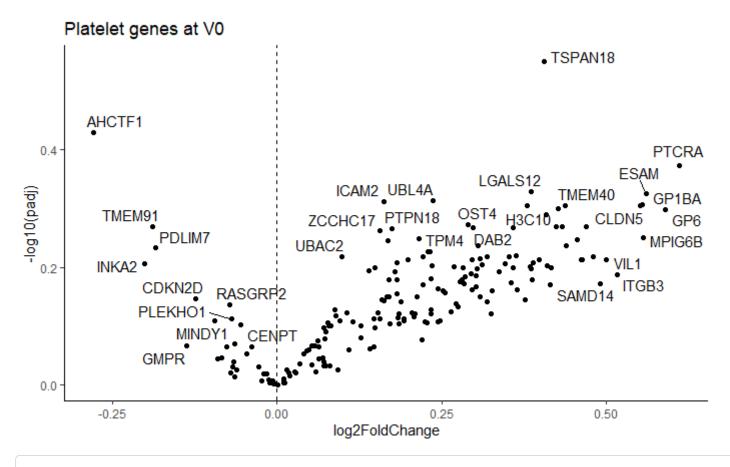
plotGseaTable(msigDB[fgseaRes[1:20, ][, pathway]], ranks, fgseaRes[1:20, ], gseaParam = 0.5, colwidths = c(3, 3, 0.8, 1.2, 1.2), pathwayLabelStyle = list(size = 10), headerLabelStyle = list(size = 10))

Pathway PLATELET	Gene ranks	NES 3.11	pval 3.8·10 <sup>-40</sup>	padj <sub>.38</sub> 1.1·10
ERYTHROBLAST		1.86	2.8·10 <sup>-17</sup>	4.0·10 <sup>-18</sup>
CD34_POS_MKP	Manager Control of the Control of th	2.16	1.3·10 <sup>-5</sup>	9.4·10 <sup>-5</sup>
DENDRITIC_CELL		1.89	1.2·10 <sup>-5</sup>	9.4·10 <sup>-5</sup>
NEUTROPHIL	<b></b>	-1.66	2.3·10 <sup>-5</sup>	1.3·10 <sup>-4</sup>
STROMAL		1.69	1.7·10 <sup>-3</sup>	8.4·10 <sup>-3</sup>
CD34_POS_LYMPHOID_UNK	1	1.45	1.3·10 <sup>-2</sup>	5.2·10 <sup>-2</sup>
EARLY_ERYTHROBLAST		1.59	1.7·10 <sup>-2</sup>	5.5·10 <sup>-2</sup>
IMMATURE_NEUTROPHIL	han	-1.43	1.8·10 <sup>-2</sup>	5.5·10 <sup>-2</sup>
MONOCYTE		1.40	1.9·10 <sup>-2</sup>	5.5·10 <sup>-2</sup>
CD34_POS_EO_B_MAST		1.59	3.3·10 <sup>-2</sup>	7.9·10 <sup>-2</sup>
PLASMA_CELL		1.51	3.1·10 <sup>-2</sup>	7.9·10 <sup>-2</sup>
EOSINOPHIL		1.50	4.0·10 <sup>-2</sup>	8.8·10 <sup>-2</sup>
PRO_B		1.24	8.2·10 <sup>-2</sup>	1.7·10 <sup>-1</sup>
CD34_POS_CLP	•	1.36	9.5·10 <sup>-2</sup>	1.8·10 <sup>-1</sup>
CD34_POS_ERP		-1.45	9.7·10 <sup>-2</sup>	1.8·10 <sup>-1</sup>
CD34_POS_MULTILIN		1.39	1.1·10 <sup>-1</sup>	1.9·10 <sup>-1</sup>
NAIVE_T_CELL		1.16	1.2·10 <sup>-1</sup>	2.0·10 <sup>-1</sup>
CD34_POS_LMPP	• •	1.26	1.6·10 <sup>-1</sup>	2.4·10 <sup>-1</sup>
CD34_POS_PRE_PC		1.21	1.8·10 <sup>-1</sup>	2.7·10 <sup>-1</sup>
	0 2500 5000 7500 10000			

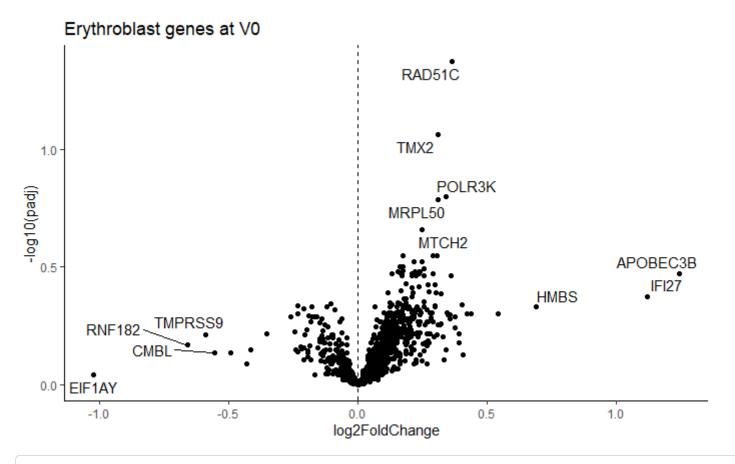
```
res.df = res.df[order(res.df$log2FoldChange, decreasing = TRUE), ]
ha top = HeatmapAnnotation(df = meta data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr
een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show annotation na
me = FALSE)
plot1 = vsd.df[rownames(subset(res.df, external gene name %in% msigDB[[fgseaRes[[4, 'pathway']]]] & log2FoldChang
e > 0))[1:10],
zzz = t(scale(t(plot1))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot1 = Heatmap(zzz, row title = fgseaRes[[4, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = FALSE, column split = meta dat
a$Group, top annotation = ha top, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25,
0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsize
= 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11), fontface = 2),
column title = 'Platelet, erythroblast, and DC signatures\nare turned up in symptomatics at V0')
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
plot2 = vsd.df[rownames(subset(res.df, external gene name %in% msiqDB[[fgseaRes[[5, 'pathway']]]] & log2FoldChang
e < 0))[1:10],
zzz = t(scale(t(plot2))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot2 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[5, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface =
2))
draw(plot1 %v% plot2); rm(plot1); rm(plot2); rm(zzz); rm(ha top)
```



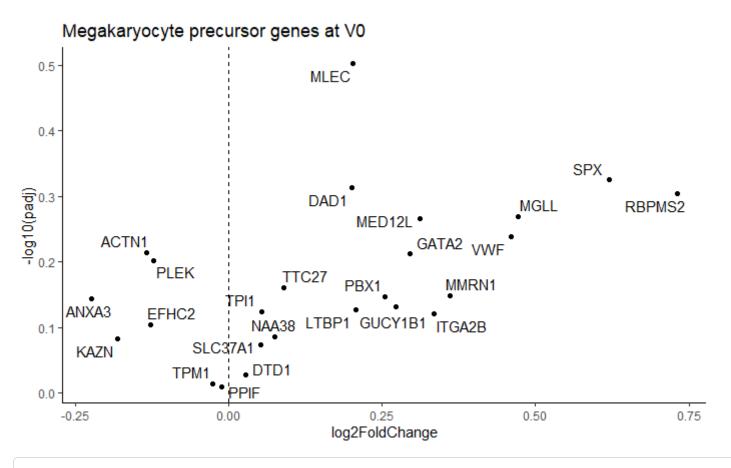
 $ggplot(subset(res.df, external\_gene\_name \%in\% \ msigDB[[fgseaRes[[1, 'pathway']]]]), \ aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Platelet genes at V0') + geom\_vline(xintercept = 0, linetype = 2)$ 



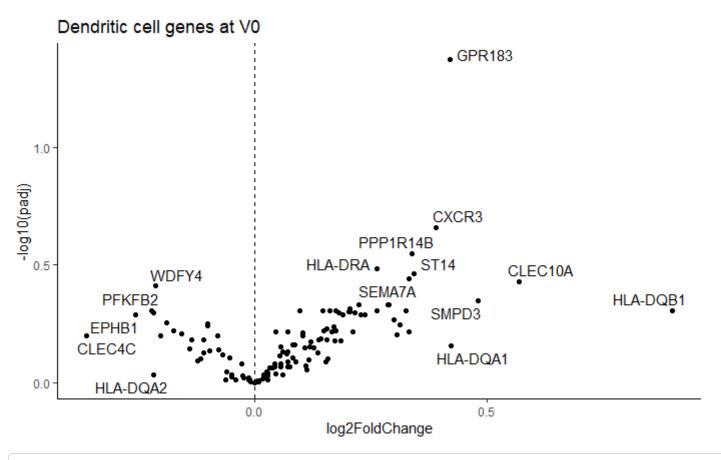
 $ggplot(subset(res.df, external\_gene\_name %in% \ msigDB[[fgseaRes[[2, 'pathway']]]]), \ aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Erythrob last genes at V0') + geom\_vline(xintercept = 0, linetype = 2)$ 



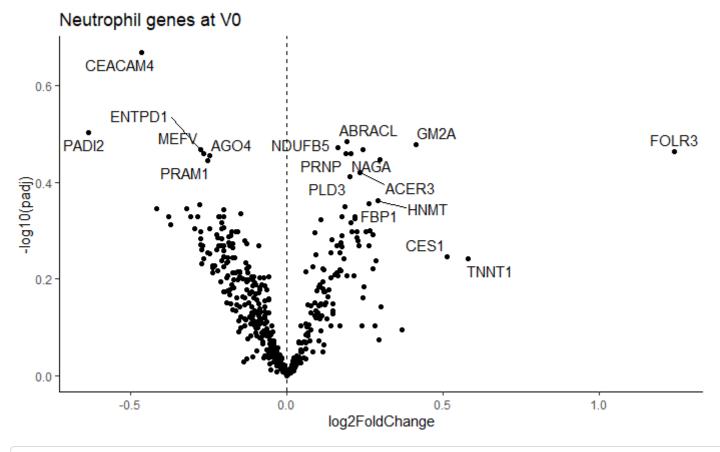
 $ggplot(subset(res.df, external\_gene\_name \%in\% \ msigDB[[fgseaRes[[3, 'pathway']]]]), \ aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Megakary ocyte precursor genes at V0') + geom\_vline(xintercept = 0, linetype = 2)$ 



 $ggplot(subset(res.df, external\_gene\_name %in% msigDB[[fgseaRes[[4, 'pathway']]]]), aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom_text_repel() + theme_classic() + ggtitle('Dendritic cell genes at V0') + geom_vline(xintercept = 0, linetype = 2)$ 



 $ggplot(subset(res.df, external\_gene\_name \%in\% \ msigDB[[fgseaRes[[5, 'pathway']]]]), \ aes(x = log2FoldChange, y = -log10(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Neutroph il genes at V0') + geom\_vline(xintercept = 0, linetype = 2)$ 



btm = readRDS('data/Blood\_Transcript\_modules.RDS'); btm\_annot = read\_excel('data/btm\_annotation\_table.xls')

```
New names:

'Module size' -> 'Module size...4'

'Jaccard Index' -> 'Jaccard Index...10'

'Enrichment p value' -> 'Enrichment p value...11'

'Overlap size' -> 'Overlap size...13'

'Jaccard Index' -> 'Jaccard Index...15'

'Enrichment p value' -> 'Enrichment p value...16'

'Module size' -> 'Module size...17'

'Overlap size' -> 'Overlap size...18'

'Jaccard Index' -> 'Jaccard Index...20'

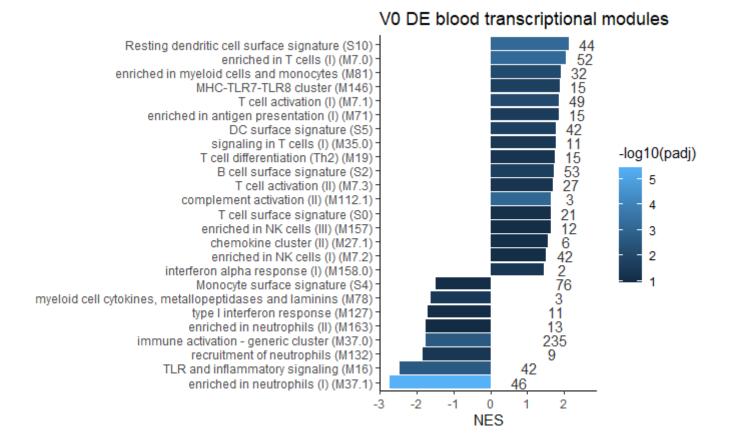
'Enrichment p value' -> 'Enrichment p value...21'

'Overlap size' -> 'Overlap size...23'
```

```
btm_annot = subset(btm_annot, `Module category` == 'immune'); btm = btm[c(btm_annot$ID)]; names(btm) = btm_anno
t$`Composite name`
fgseaRes = fgsea(pathways = btm, stats = ranks, nproc = 1); fgseaRes = fgseaRes[order(fgseaRes$padj, decreasing =
FALSE), ]
```

```
write.csv(subset(fgseaRes, select = -c(leadingEdge)), file = 'results/fgseaRes_btm_immune_v0.csv', row.names = FA
LSE)

fgseaRes = fgseaRes[pathway %in% subset(btm_annot, `Module category` == 'immune')$`Composite name`, ]
ggplot(fgseaRes[1:25, ], aes(y = reorder(pathway, NES), x = NES, fill = -log10(padj), label = size)) + geom_col()
+ geom_text(color = 'gray20', nudge_x = ifelse(fgseaRes$NES > 0, 0.5, 3.5)) + ylab('') + theme_classic() + ggtitl
e('V0 DE blood transcriptional modules')
```

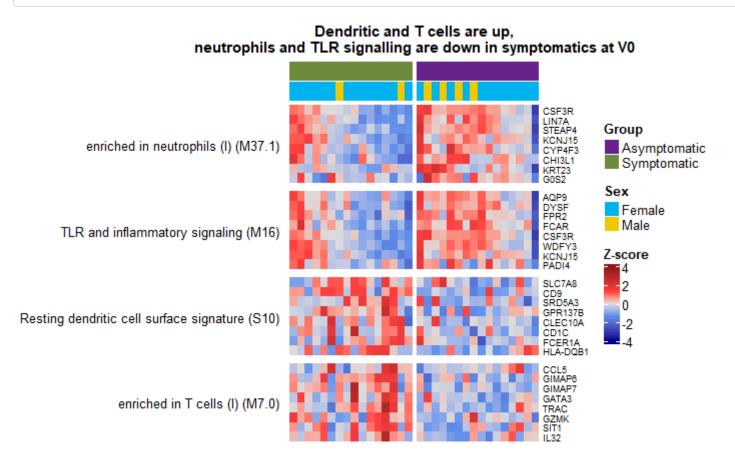


plotGseaTable(btm[fgseaRes[1:20, ][, pathway]], ranks, fgseaRes[1:20, ], gseaParam = 0.5, colwidths = c(6, 3, 0.
8, 1.2, 1.2), pathwayLabelStyle = list(size = 10), headerLabelStyle = list(size = 10), valueStyle = list(size = 10))

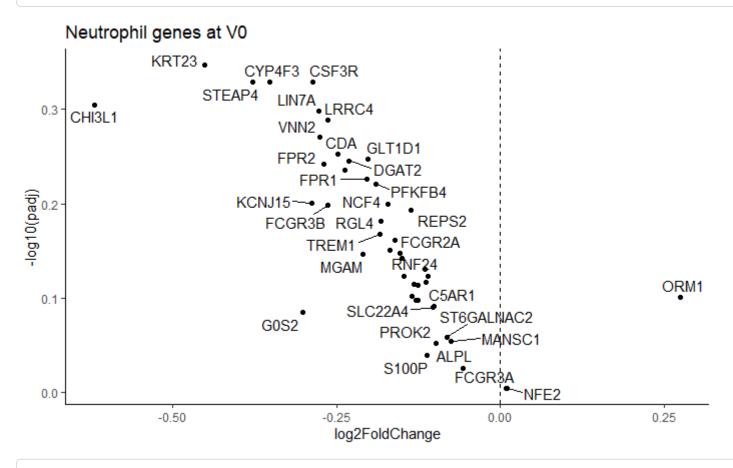
Pathway	Gene ranks NES		padj 3.5·10 <sup>-8</sup>
enriched in neutrophils (I) (M37.1)	· · · · · ·	_	
Resting dendritic cell surface signature (S10)	····· 2.1	3 1.5·10 <sup>-5</sup>	7.1·10 <sup>-4</sup>
enriched in T cells (I) (M7.0)	2.00	6 2.1·10 <sup>-5</sup>	7.1·10 <sup>-4</sup>
complement activation (II) (M112.1)	1.69	5 3.2·10 <sup>-5</sup>	8.1·10 <sup>-4</sup>
TLR and inflammatory signaling (M16)		6 1.2·10 <sup>-4</sup>	2.4·10 <sup>-3</sup>
immune activation - generic cluster (M37.0)	-1.7	6 1.4·10 <sup>-4</sup>	2.4·10 <sup>-3</sup>
T cell activation (I) (M7.1)	1.80	5.5·10 <sup>-4</sup>	8.0·10 <sup>-3</sup>
enriched in myeloid cells and monocytes (M81)	1.9	1 6.4·10 <sup>-4</sup>	8.3·10 <sup>-3</sup>
DC surface signature (S5)	1.78	3 1.6·10 <sup>-3</sup>	1.8·10 <sup>-2</sup>
MHC-TLR7-TLR8 cluster (M146)	····· 1.89	9 1.8·10 <sup>-3</sup>	1.9·10 <sup>-2</sup>
B cell surface signature (S2)	1.72	2.1·10 <sup>-3</sup>	2.0·10 <sup>-2</sup>
enriched in antigen presentation (I) (M71)	1.80	6 2.9·10 <sup>-3</sup>	2.5·10 <sup>-2</sup>
oid cell cytokines, metallopeptidases and laminins (M78)	1.6	3 4.2·10 <sup>-3</sup>	3.3·10 <sup>-2</sup>
signaling in T cells (I) (M35.0)	1.7	7 4.9·10 <sup>-3</sup>	3.6·10 <sup>-2</sup>
interferon alpha response (I) (M158.0)	1.49	5 6.5·10 <sup>-3</sup>	4.2·10 <sup>-2</sup>
recruitment of neutrophils (M132)	1.8	3 6.1·10 <sup>-3</sup>	4.2·10 <sup>-2</sup>
T cell differentiation (Th2) (M19)		5 8.2·10 <sup>-3</sup>	5.0·10 <sup>-2</sup>
T cell activation (II) (M7.3)	1.70	9.0·10 <sup>-3</sup>	5.1·10 <sup>-2</sup>
T cell surface signature (S0)	1.64	4 1.3·10 <sup>-2</sup>	7.0·10 <sup>-2</sup>
chemokine cluster (II) (M27.1)	1.50	6 1.6·10 <sup>-2</sup>	8.0·10 <sup>-2</sup>
	0 2500 5000 7500 10000		

```
res.df = res.df[order(res.df$log2FoldChange, decreasing = FALSE), ]
ha top = HeatmapAnnotation(df = meta data[, c('Group', 'Sex')], col = list(Group = c('Symptomatic' = 'darkolivegr
een4', 'Asymptomatic' = 'darkorchid4'), Sex = c('Male' = 'gold2', 'Female' = 'deepskyblue2')), show annotation na
me = FALSE)
plot1 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[1, 'pathway']]]] & log2FoldChange <</pre>
0))[1:8], ]
zzz = t(scale(t(plot1))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot1 = Heatmap(zzz, row title = fgseaRes[[1, 'pathway']], show row names = TRUE, show column names = FALSE, clus
ter rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = FALSE, column split = meta dat
a$Group, top annotation = ha top, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.25,
0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsize
= 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11), fontface = 2),
column title = 'Dendritic and T cells are up,\nneutrophils and TLR signalling are down in symptomatics at V0')
plot2 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[5, 'pathway']]]] & log2FoldChange <</pre>
0))[1:8], ]
zzz = t(scale(t(plot2))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot2 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[5, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title qp = qpar(fontsize = 10), column title qp = qpar(fontsize = 11, fontface =
2))
res.df = res.df[order(res.df$log2FoldChange, decreasing = TRUE), ]
plot3 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[2, 'pathway']]]] & log2FoldChange >
0))[1:8],
zzz = t(scale(t(plot3))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external gene nam
e']
plot3 = Heatmap(zzz, show heatmap legend = FALSE, row title = fgseaRes[[2, 'pathway']], show row names = TRUE, sh
ow column names = FALSE, cluster rows = TRUE, cluster columns = TRUE, show row dend = FALSE, show column dend = F
ALSE, column split = meta data$Group, heatmap legend param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row names gp = gpar(fontsi
ze = 7), row title rot = 0, row title gp = gpar(fontsize = 10), column title gp = gpar(fontsize = 11, fontface =
2))
plot4 = vsd.df[rownames(subset(res.df, external gene name %in% btm[[fgseaRes[[3, 'pathway']]]] & log2FoldChange >
0))[1:8], ]
```

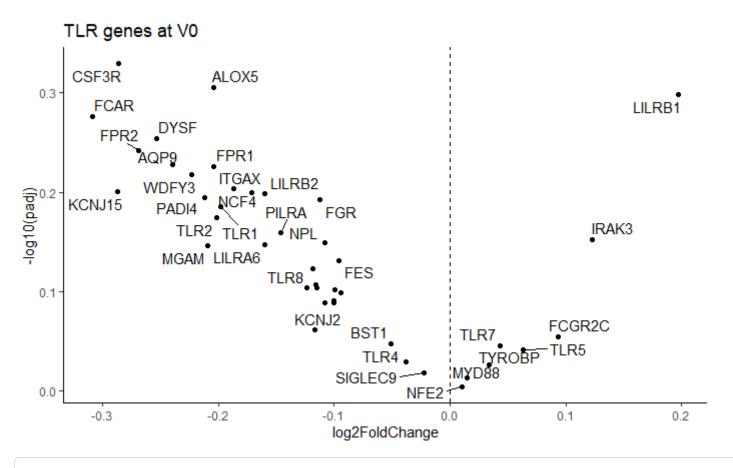
```
zzz = t(scale(t(plot4))); zzz = pmin(pmax(zzz, -3), 3); rownames(zzz) = res.df[rownames(zzz), 'external_gene_nam
e']
plot4 = Heatmap(zzz, show_heatmap_legend = FALSE, row_title = fgseaRes[[3, 'pathway']], show_row_names = TRUE, sh
ow_column_names = FALSE, cluster_rows = TRUE, cluster_columns = TRUE, show_row_dend = FALSE, show_column_dend = F
ALSE, column_split = meta_data$Group, heatmap_legend_param = list(title = "Z-score"), col = colorRamp2(c(-4, -1.2
5, 0, 1.25, 4), c("darkblue", "cornflowerblue", "gainsboro", "brown1", "firebrick4")), row_names_gp = gpar(fontsi
ze = 7), row_title_rot = 0, row_title_gp = gpar(fontsize = 10), column_title_gp = gpar(fontsize = 11, fontface =
2))
draw(plot1 %v% plot2 %v% plot3 %v% plot4); rm(plot1); rm(plot2); rm(plot3); rm(plot4); rm(ha_top); rm(zzz); rm(ra
nks)
```



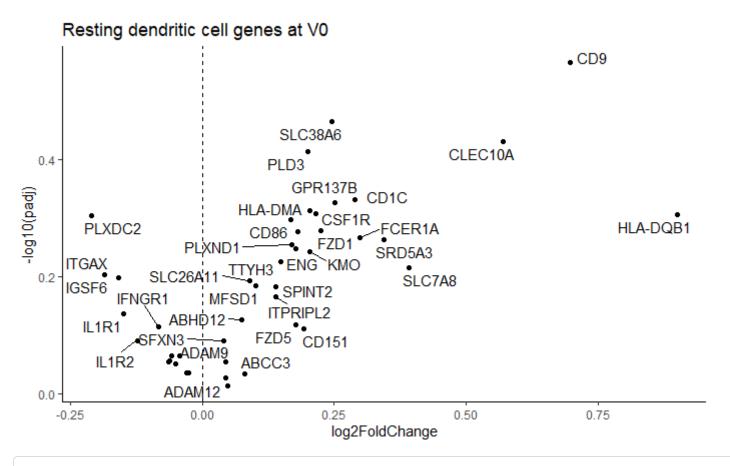
```
ggplot(subset(res.df, external_gene_name %in% btm[[fgseaRes[[1, 'pathway']]]]), aes(x = log2FoldChange, y = -log1
0(padj), label = external_gene_name)) + geom_point() + geom_text_repel() + theme_classic() + ggtitle('Neutrophil
genes at V0') + geom_vline(xintercept = 0, linetype = 2)
```



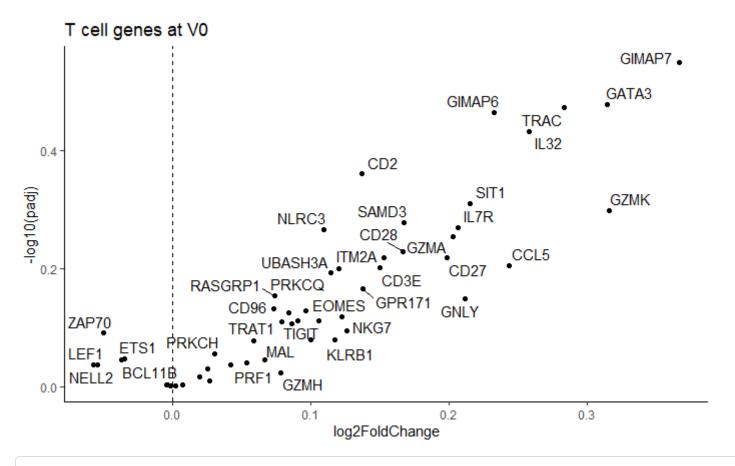
 $ggplot(subset(res.df, external\_gene\_name \%in\% btm[[fgseaRes[[5, 'pathway']]]]), aes(x = log2FoldChange, y = -log1 0(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('TLR genes a t V0') + geom\_vline(xintercept = 0, linetype = 2)$ 



ggplot(subset(res.df, external\_gene\_name %in% btm[[fgseaRes[[2, 'pathway']]]]), aes(x = log2FoldChange, y = -log1
0(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Resting den
dritic cell genes at V0') + geom\_vline(xintercept = 0, linetype = 2)



ggplot(subset(res.df, external\_gene\_name %in% btm[[fgseaRes[[3, 'pathway']]]]), aes(x = log2FoldChange, y = -log1
0(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('T cell gene
s at V0') + geom\_vline(xintercept = 0, linetype = 2)



 $ggplot(subset(res.df, external\_gene\_name %in% btm[[fgseaRes[[8, 'pathway']]]]), aes(x = log2FoldChange, y = -log1 0(padj), label = external\_gene\_name)) + geom\_point() + geom\_text\_repel() + theme\_classic() + ggtitle('Myeloid cel l and monocyte genes at V0') + geom\_vline(xintercept = 0, linetype = 2)$ 

