

SIB - Enrichment Analysis

Exercise 1

```
library(clusterProfiler)
library(enrichplot)
library(pathview)
library(org.Hs.eg.db)
library(ggplot2)
library(ggrepel)
library(msigdb)

library(tidyverse) # for bonus code/dplyr/pipe

# set seed
set.seed(1234)
```

```
# Import DE table:
NK_vs_Th <- read.csv("data/NK_vs_Th_diff_gene_exercise_1.csv",
  header = T
)
# Look at the structure of the data.frame:
head(NK_vs_Th)
```

	ensembl_gene_id	symbol	logFC	t	P.Value	p.adj
1	ENSG000000000003	TSPAN6	-5.6436044	-4.672128	0.0000426000	7.358019e-04
2	ENSG000000000419	DPM1	-0.1818981	-1.101831	0.2780198240	5.176076e-01
3	ENSG000000000457	SCYL3	0.4969874	1.491035	0.1448690710	3.449889e-01
4	ENSG000000000460	C1orf112	1.1217991	1.445899	0.1570598770	3.630935e-01
5	ENSG000000000938	FGR	10.6706873	7.212342	0.0000000198	1.718657e-06
6	ENSG000000000971	CFH	-3.4129277	-2.788887	0.0084803000	4.610083e-02

```
# Search for a gene symbol in the data.frame, eg NCAM1 (CD56)
NK_vs_Th[which(NK_vs_Th$symbol == "NCAM1"), ]
```

	ensembl_gene_id	symbol	logFC	t	P.Value	p.adj
7624	ENSG00000149294	NCAM1	12.19755	6.992219	3.81e-08	2.845553e-06

Search for 2 genes in the data.frame, CPS1 and GZMB, and verify the effect of adjustment on their p-values

```
genes <- c("CPS1", "GZMB")
NK_vs_Th |>
  filter(symbol %in% genes) |>
  select(symbol, P.Value, p.adj)
```

	symbol	P.Value	p.adj
1	CPS1	0.044963086	1.565113e-01
2	GZMB	0.000000629	2.402609e-05

CPS1 is not significant, while GZMB is significant.

```
# Import the adaptive immune response gene set (gmt file)
adaptive <- clusterProfiler::read.gmt("data/GOBP_ADAPTIVE_IMMUNE_RESPONSE.v7.5.1.gmt")
nrow(adaptive) # 719
```

```
[1] 719
```

```
length(which(NK_vs_Th$symbol %in% adaptive$gene)) # 513
```

```
[1] 513
```

```
upregulated_th <- subset(
  NK_vs_Th,
  NK_vs_Th$p.adj <= 0.05 & NK_vs_Th$logFC < 0
)

not_significant_genes <- subset(
  NK_vs_Th,
```

```

NK_vs_Th$p.adj > 0.05
)

summary_upregulated <- summary(upregulated_th$symbol %in% adaptive$gene)
summary_not_significant <- summary(not_significant_genes$symbol %in% adaptive$gene)

contingency_table <- matrix(, nrow = 2, ncol = 2)
contingency_table[[1]] <- summary_upregulated[[3]] # up, in gene set
contingency_table[[2]] <- summary_upregulated[[2]] # up, not in gene set
contingency_table[[3]] <- summary_not_significant[[3]] # down, in gene set
contingency_table[[4]] <- summary_not_significant[[2]] # down, not in gene set

# Convert to numeric
contingency_table <- apply(contingency_table, 2, as.numeric)

# Add rows and columns
colnames(contingency_table) <- c("up", "down")
rownames(contingency_table) <- c("in_set", "not_in_set")

fisher.test(contingency_table)

```

Fisher's Exact Test for Count Data

```

data:  contingency_table
p-value < 2.2e-16
alternative hypothesis: true odds ratio is not equal to 1
95 percent confidence interval:
 3.697701 5.654348
sample estimates:
odds ratio
 4.580549

```

The odds ratio tells us how different the two proportions are.

If the confidence interval does not include 1, then p-value is small. We can reject null hypothesis that the odds ratio is equal to 1.

There are more genes that are upregulated in the gene set than the genes that are not upregulated in the gene set.

```
# Test 3 gene sets among the genes up-regulated in NK cells,
# with enricher()
# First, obtain the genes up-regulated in NK:

nk_up_genes <- subset(NK_vs_Th, NK_vs_Th$logFC > 0 & NK_vs_Th$p.adj <= 0.05)$symbol

# Import 2 other gene sets, 1 un-related to immune cells:
hair <- read.gmt("data/GOBP_HAIR_CELL_DIFFERENTIATION.v7.5.1.gmt")
dim(hair)
```

```
[1] 47  2
```

```
cell_active <- read.gmt("data/GOBP_CELL_ACTIVATION.v7.5.1.gmt")
dim(cell_active)
```

```
[1] 1095  2
```

```
# Combine the 3 gene sets into a single data.frame for the TERM2GENE argument:
genesets3 <- rbind(adaptive, hair, cell_active)

hyper_3genesets <- enricher(
  gene = nk_up_genes,
  universe = NK_vs_Th$symbol,
  TERM2GENE = genesets3,
  maxGSSize = 1000
)
hyper_3genesets@result
```

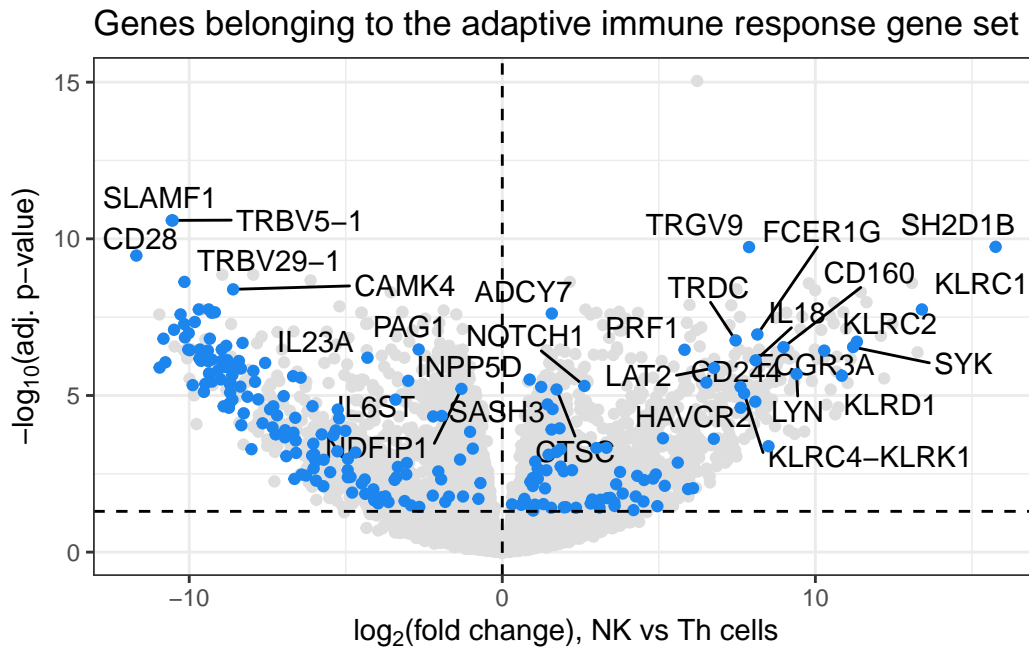
		ID
GOBP_CELL_ACTIVATION	GOBP_CELL_ACTIVATION	
GOBP_HAIR_CELL_DIFFERENTIATION	GOBP_HAIR_CELL_DIFFERENTIATION	
GOBP_ADAPTIVE_IMMUNE_RESPONSE	GOBP_ADAPTIVE_IMMUNE_RESPONSE	
	Description	GeneRatio
GOBP_CELL_ACTIVATION	GOBP_CELL_ACTIVATION	173/200
GOBP_HAIR_CELL_DIFFERENTIATION	GOBP_HAIR_CELL_DIFFERENTIATION	5/200
GOBP_ADAPTIVE_IMMUNE_RESPONSE	GOBP_ADAPTIVE_IMMUNE_RESPONSE	82/200
	BgRatio	pvalue
	p.adjust	qvalue
GOBP_CELL_ACTIVATION	896/1138	0.001505054
GOBP_HAIR_CELL_DIFFERENTIATION	34/1138	0.741306145
GOBP_ADAPTIVE_IMMUNE_RESPONSE	513/1138	0.912609682

GOBP_CELL_ACTIVATION	FGR/CD38/SKAP2/ITGAL/TYROBP/RUNX3/NR1H3/SLAMF7/IFNGR1/STAP1/H
GOBP_HAIR_CELL_DIFFERENTIATION	
GOBP_ADAPTIVE_IMMUNE_RESPONSE	
	Count
GOBP_CELL_ACTIVATION	173
GOBP_HAIR_CELL_DIFFERENTIATION	5
GOBP_ADAPTIVE_IMMUNE_RESPONSE	82

```
sig_genes <- subset(NK_vs_Th, NK_vs_Th$symbol %in% adaptive$gene &
  NK_vs_Th$p.adj <= 0.05)
sig_genes_label <- subset(sig_genes, sig_genes$p.adj <= 0.00001)

ggplot(NK_vs_Th, aes(
  x = logFC,
  y = -log10(p.adj)
)) +
  geom_point(color = "grey87") +
  ggtitle("Genes belonging to the adaptive immune response gene set") +
  theme_bw() +
  geom_text_repel(
    data = sig_genes_label,
    aes(
      x = logFC,
      y = -log10(p.adj), label = symbol
    ),
    max.overlaps = 20
  ) +
  geom_point(data = sig_genes, col = "dodgerblue2") +
  theme(legend.position = "none") +
  scale_x_continuous(name = expression("log"[2] * "(fold change), NK vs Th cells")) +
  scale_y_continuous(name = expression("-" * "log"[10] * "(adj. p-value)")) +
  geom_hline(yintercept = -log10(0.05), linetype = "dashed") +
  geom_vline(xintercept = 0, linetype = "dashed")
```

Warning: ggrepel: 51 unlabeled data points (too many overlaps). Consider increasing max.overlaps



Exercise 2 - Gene set enrichment analysis (GSEA)

```
gl <- NK_vs_Th$t
names(gl) <- make.names(NK_vs_Th$symbol, unique = T)
gl <- gl[order(gl, decreasing = T)]

GO_NK_Th <- gseGO(gl,
  ont = "BP",
  OrgDb = org.Hs.eg.db,
  keyType = "SYMBOL",
  minGSSize = 30,
  eps = 0,
  seed = T
)
```

preparing geneSet collections...

GSEA analysis...

Warning in preparePathwaysAndStats(pathways, stats, minSize, maxSize, gseaParam, : There are
The order of those tied genes will be arbitrary, which may produce unexpected results.

leading edge analysis...

done...

GO_NK_Th

```
#
# Gene Set Enrichment Analysis
#
#...@organism      Homo sapiens
#...@setType       BP
#...@keytype       SYMBOL
#...@geneList      Named num [1:20485] 19 13.1 12.1 12 10.7 ...
  - attr(*, "names")= chr [1:20485] "GHSR" "MLC1" "SH2D1B" "TRGV9" ...
#...nPerm
#...pvalues adjusted by 'BH' with cutoff <0.05
#...351 enriched terms found
'data.frame':   351 obs. of  11 variables:
 $ ID          : chr  "GO:0002181" "GO:0042254" "GO:0022613" "GO:0042273" ...
 $ Description  : chr  "cytoplasmic translation" "ribosome biogenesis" "ribonucleoprotein ..."
 $ setSize     : int   145 299 436 69 98 57 210 340 111 242 ...
 $ enrichmentScore: num  -0.808 -0.551 -0.491 -0.715 -0.648 ...
 $ NES          : num   -3.38 -2.52 -2.36 -2.67 -2.58 ...
 $ pvalue       : num   1.68e-48 6.30e-24 2.34e-22 1.48e-14 3.84e-14 ...
 $ p.adjust     : num   4.85e-45 9.08e-21 2.25e-19 1.07e-11 2.21e-11 ...
 $ qvalue       : num   3.77e-45 7.05e-21 1.75e-19 8.27e-12 1.72e-11 ...
 $ rank         : num   1385 3723 3746 2698 2371 ...
 $ leading_edge  : chr   "tags=58%, list=7%, signal=54%" "tags=38%, list=18%, signal=32%" "t..."
 $ core_enrichment: chr   "EIF2S2/EIF3M/RPL21/RPS28/EIF4A2/YBX1/FAU/PKM/RPS9/CNBP/RPL37/RPL37..."
#...Citation
T Wu, E Hu, S Xu, M Chen, P Guo, Z Dai, T Feng, L Zhou, W Tang, L Zhan, X Fu, S Liu, X Bo, ...
clusterProfiler 4.0: A universal enrichment tool for interpreting omics data.
The Innovation. 2021, 2(3):100141
```

```
# Class is gseaResult
class(GO_NK_Th)
```

```
[1] "gseaResult"
attr(,"package")
[1] "DOSE"
```

```
# Is the adaptive immune response gene set significant?
GO_NK_Th[GO_NK_Th@result$Description == "adaptive immune response", ] # yes
```

```

              ID              Description setSize enrichmentScore
GO:0002250 GO:0002250 adaptive immune response      423      -0.3652034
              NES      pvalue      p.adjust      qvalue rank
GO:0002250 -1.743904 1.001619e-08 1.804167e-06 1.400949e-06 1623
              leading_edge
GO:0002250 tags=23%, list=8%, signal=22%

GO:0002250 HFE/CD3E/CLU/PDCD1LG2/ADGRE1/JAK3/LEF1/IL18BP/ITK/CD80/ALCAM/TRAV34/AIRE/IGHM/BTLA
```

```
# How many gene sets are down- or up-regulated?
count_gene_sets <- function(gsea, p_value) {
  up <- summary(gsea@result$p.adjust < p_value & gsea@result$NES > 0)
  down <- summary(gsea@result$p.adjust < p_value & gsea@result$NES < 0)

  return(list(upregulated = up, downregulated = down))
}

# 290 upregulated, 61 downregulated
count_gene_sets(GO_NK_Th, 0.05)
```

```
$upregulated
  Mode  FALSE  TRUE
logical    61   290
```

```
$downregulated
  Mode  FALSE  TRUE
logical   290    61
```

```
GO_NK_Th_simplify <- clusterProfiler::simplify(GO_NK_Th)
GO_NK_Th_simplify@result[GO_NK_Th_simplify@result$Description == "adaptive immune response", ]
```

```

              ID              Description setSize enrichmentScore
GO:0002250 GO:0002250 adaptive immune response      423      -0.3652034
              NES      pvalue      p.adjust      qvalue rank
GO:0002250 -1.743904 1.001619e-08 1.804167e-06 1.400949e-06 1623
              leading_edge
GO:0002250 tags=23%, list=8%, signal=22%
```


GO:0002250 HFE/CD3E/CLU/PDCD1LG2/ADGRE1/JAK3/LEF1/IL18BP/ITK/CD80/ALCAM/TRAV34/AIRE/IGHM/BTLA

```
unlist(strsplit(
  GO_NK_Th@result[GO_NK_Th@result$Description == "adaptive immune response", 11],
  "\\\/"
))
```

[1]	"HFE"	"CD3E"	"CLU"	"PDCD1LG2"	"ADGRE1"	"JAK3"
[7]	"LEF1"	"IL18BP"	"ITK"	"CD80"	"ALCAM"	"TRAV34"
[13]	"AIRE"	"IGHM"	"BTLA"	"CR1"	"C1QBP"	"CD3G"
[19]	"CTSL"	"TRAJ42"	"TRBV16"	"TNF"	"CEACAM1"	"GPR183"
[25]	"CD27"	"CCR6"	"ICOSLG"	"TRDV1"	"CCR2"	"CD6"
[31]	"TRBD1"	"MCOLN2"	"TRAV14DV4"	"IL2"	"CR2"	"TRAV22"
[37]	"CD70"	"PDCD1"	"MALT1"	"EBAG9"	"TRAV30"	"CTLA4"
[43]	"TRAV23DV6"	"TNFRSF13C"	"KDM5D"	"TRAV40"	"TRAV18"	"IL6R"
[49]	"CD3D"	"TRAJ3"	"TRAV39"	"TRBC2"	"SAMSN1"	"IL7R"
[55]	"TRAV19"	"SUSD4"	"TRAV20"	"CD84"	"TRAV10"	"TRAV21"
[61]	"TRBV13"	"TRAV41"	"TRAV29DV5"	"NDFIP1"	"TRAV36DV7"	"THEMIS"
[67]	"TRBV18"	"TRAT1"	"SOCS3"	"IL6ST"	"TRBV9"	"TRAV24"
[73]	"TRAV3"	"TRAV27"	"TRAV4"	"TRAV6"	"TRAV2"	"TRAV5"
[79]	"JUNB"	"TRBV19"	"TRAV35"	"TRBV30"	"FOXP3"	"TRAV16"
[85]	"IL23A"	"TRBV2"	"TRBV14"	"PAG1"	"CD4"	"TRAV25"
[91]	"SIT1"	"TRAV17"	"CD40LG"	"CAMK4"	"TRAC"	"CD28"
[97]	"SLAMF1"					

GO_NK_Th@geneSets\$`GO:0002250`

[1]	"ADA"	"ADCY7"	"AGER"	"JAG1"	"AHR"
[6]	"ALCAM"	"ALOX15"	"ANXA1"	"AIRE"	"ARG1"
[11]	"ARG2"	"ASCL2"	"B2M"	"BCL3"	"BCL6"
[16]	"TNFRSF17"	"CEACAM1"	"PRDM1"	"BMX"	"BTK"
[21]	"C1QBP"	"SERPING1"	"C1QA"	"C1QB"	"C1QC"
[26]	"C1R"	"C1S"	"C2"	"C3"	"C4A"
[31]	"C4B"	"C4BPA"	"C4BPB"	"C5"	"C6"
[36]	"C7"	"C8A"	"C8B"	"C8G"	"C9"
[41]	"CAMK4"	"CD1A"	"CD1B"	"CD1C"	"CD1D"
[46]	"CD1E"	"CD3D"	"CD3E"	"CD3G"	"CD247"
[51]	"CD4"	"CD6"	"CD7"	"CD8A"	"CD8B"
[56]	"CD8B2"	"CD19"	"CD27"	"CD28"	"CD80"
[61]	"CD86"	"CD40"	"CD40LG"	"CD70"	"CD74"

[66]	"CD79A"	"CD79B"	"CD81"	"CTSC"	"CLC"
[71]	"CLU"	"CCR6"	"CR1"	"CR1L"	"CR2"
[76]	"CSF2RB"	"CSK"	"CTLA4"	"CTSH"	"CTSL"
[81]	"CTSS"	"CX3CR1"	"CD55"	"GPR183"	"EMP2"
[86]	"ADGRE1"	"EPHB2"	"ERCC1"	"PTK2B"	"FCER1A"
[91]	"FCER1G"	"FCER2"	"FCGR1A"	"FCGR1BP"	"FCGR2B"
[96]	"FCGR3A"	"FGA"	"FGB"	"FGL1"	"FOXJ1"
[101]	"MTOR"	"FUT7"	"FYN"	"GATA3"	"GNL1"
[106]	"MSH6"	"GZMM"	"NCKAP1L"	"HFE"	"HLA-A"
[111]	"HLA-B"	"HLA-C"	"HLA-DMA"	"HLA-DMB"	"HLA-DOA"
[116]	"HLA-DOB"	"HLA-DPA1"	"HLA-DPB1"	"HLA-DQA1"	"HLA-DQA2"
[121]	"HLA-DQB1"	"HLA-DQB2"	"HLA-DRA"	"HLA-DRB1"	"HLA-DRB3"
[126]	"HLA-DRB4"	"HLA-DRB5"	"HLA-E"	"HLA-F"	"HLA-G"
[131]	"HLA-H"	"MR1"	"HLX"	"HMGB1"	"HPRT1"
[136]	"HPX"	"HRAS"	"HSPD1"	"ICAM1"	"CFI"
[141]	"IFNA1"	"IFNA2"	"IFNA4"	"IFNA5"	"IFNA6"
[146]	"IFNA7"	"IFNA8"	"IFNA10"	"IFNA13"	"IFNA14"
[151]	"IFNA16"	"IFNA17"	"IFNA21"	"IFNB1"	"IFNG"
[156]	"IFNW1"	"IGHA1"	"IGHA2"	"IGHD"	"IGHE"
[161]	"IGHG1"	"IGHG2"	"IGHG3"	"IGHG4"	"IGHM"
[166]	"JCHAIN"	"IGKC"	"IGLC1"	"IGLC2"	"IGLC3"
[171]	"IGLC6"	"IGLL1"	"IL1B"	"IL1R1"	"IL2"
[176]	"IL2RB"	"IL4"	"IL4R"	"IL6"	"IL6R"
[181]	"IL6ST"	"IL7R"	"IL9"	"IL9R"	"IL10"
[186]	"IL12A"	"IL12B"	"IL12RB1"	"IL13RA2"	"IL17A"
[191]	"IL18"	"INPP5D"	"IRF1"	"IRF4"	"IRF7"
[196]	"ITK"	"JAK1"	"JAK2"	"JAK3"	"JUNB"
[201]	"KCNJ8"	"KLRC1"	"KLRC2"	"KLRD1"	"LAG3"
[206]	"LAIR1"	"LIG4"	"LTA"	"LY9"	"LYN"
[211]	"SH2D1A"	"SMAD7"	"MBL2"	"CD46"	"MEF2C"
[216]	"MICB"	"MLH1"	"MPL"	"MSH2"	"MYD88"
[221]	"NBN"	"NFKB2"	"NOTCH1"	"P2RX7"	"PDCD1"
[226]	"PHB1"	"PIK3CD"	"PIK3CG"	"PLA2G4A"	"PMS2"
[231]	"PPP3CB"	"PRF1"	"PRKCB"	"PRKCD"	"PKN1"
[236]	"PRKCQ"	"PRKCZ"	"PSG9"	"PTPN6"	"PTPRC"
[241]	"PVR"	"NECTIN2"	"RAB27A"	"RAG1"	"RAP1GAP"
[246]	"RELB"	"TRIM27"	"RORA"	"RORC"	"CCL19"
[251]	"XCL1"	"SIPA1"	"SLAMF1"	"SLC11A1"	"SPN"
[256]	"STAT3"	"STAT4"	"STAT6"	"SUPT6H"	"SYK"
[261]	"ADAM17"	"MAP3K7"	"TAP1"	"TAP2"	"TRA"
[266]	"TRAV6"	"TRB"	"TRGC1"	"TRGC2"	"TRGV1"
[271]	"TRGV2"	"TRGV3"	"TRGV4"	"TRGV5"	"TRGV8"
[276]	"TRGV9"	"TRGV10"	"TRGV11"	"TEC"	"TFE3"

[281]	"TFRC"	"TGFB1"	"TLR4"	"TNF"	"TNFAIP3"
[286]	"TNFRSF1B"	"TP53BP1"	"TRAF2"	"TRAF6"	"TSC1"
[291]	"TNFSF4"	"TXK"	"TYK2"	"UNG"	"WAS"
[296]	"LAT2"	"NSD2"	"ZAP70"	"ZP3"	"FZD5"
[301]	"TFEB"	"KDM5D"	"EOMES"	"STX7"	"SKAP1"
[306]	"TNFSF13"	"TNFRSF14"	"RIPK2"	"FADD"	"TNFRSF11A"
[311]	"IL18R1"	"CD84"	"BCL10"	"TNFSF18"	"SOCS3"
[316]	"RNF8"	"EXO1"	"EBAG9"	"IL1RL1"	"SLC22A13"
[321]	"IL27RA"	"SOCS5"	"THOC1"	"PARP3"	"IL18BP"
[326]	"EBI3"	"LILRB2"	"TCIRG1"	"CLEC4M"	"BTN3A3"
[331]	"RAPGEF3"	"MAD2L2"	"CLEC10A"	"BATF"	"CXCL13"
[336]	"CD226"	"TNFSF13B"	"MASP2"	"TRAF3IP2"	"LILRB1"
[341]	"ARID5A"	"MALT1"	"LILRB5"	"LILRB4"	"LILRA1"
[346]	"LILRB3"	"LILRA3"	"RIPK3"	"RAPGEF4"	"BTN3A2"
[351]	"BTN3A1"	"CD160"	"DUSP10"	"TREX1"	"ZBTB1"
[356]	"KLRK1"	"PAXIP1"	"SWAP70"	"RAP1GAP2"	"RFTN1"
[361]	"ICOSLG"	"SIRT1"	"TNFRSF13B"	"CLCF1"	"IL17RA"
[366]	"PRKD2"	"TMEM98"	"LAT"	"LAMP3"	"SIT1"
[371]	"TNFRSF21"	"IGKV1-5"	"IGHV8-51-1"	"IGHV7-81"	"IGHV6-1"
[376]	"IGHV5-10-1"	"IGHV5-51"	"IGHV4-38-2"	"IGHV4-61"	"IGHV4-59"
[381]	"IGHV4-39"	"IGHV4-34"	"IGHV4-31"	"IGHV4-30-4"	"IGHV4-28"
[386]	"IGHV4-4"	"IGHV3-38-3"	"IGHV3-74"	"IGHV3-73"	"IGHV3-72"
[391]	"IGHV3-66"	"IGHV3-64"	"IGHV3-53"	"IGHV3-49"	"IGHV3-48"
[396]	"IGHV3-43"	"IGHV3-38"	"IGHV3-35"	"IGHV3-33"	"IGHV3-30"
[401]	"IGHV3-23"	"IGHV3-21"	"IGHV3-20"	"IGHV3-16"	"IGHV3-15"
[406]	"IGHV3-13"	"IGHV3-11"	"IGHV3-9"	"IGHV3-7"	"IGHV2-70"
[411]	"IGHV2-26"	"IGHV2-5"	"IGHV1-69-2"	"IGHV1-38-4"	"IGHV1-69"
[416]	"IGHV1-58"	"IGHV1-45"	"IGHV1-24"	"IGHV1-18"	"IGHV1-8"
[421]	"IGHV1-3"	"IGHJ1"	"IGHD1-1"	"TRDV3"	"TRDV2"
[426]	"TRDV1"	"TRDJ1"	"TRDD1"	"TRDC"	"TRBV30"
[431]	"TRBV29-1"	"TRBV28"	"TRBV27"	"TRBV25-1"	"TRBV24-1"
[436]	"TRBV23-1"	"TRBV20-1"	"TRBV19"	"TRBV18"	"TRBV17"
[441]	"TRBV16"	"TRBV14"	"TRBV13"	"TRBV12-5"	"TRBV12-4"
[446]	"TRBV12-3"	"TRBV11-3"	"TRBV11-2"	"TRBV11-1"	"TRBV10-3"
[451]	"TRBV10-2"	"TRBV10-1"	"TRBV9"	"TRBV7-9"	"TRBV7-8"
[456]	"TRBV7-7"	"TRBV7-6"	"TRBV7-4"	"TRBV7-3"	"TRBV7-2"
[461]	"TRBV7-1"	"TRBV6-9"	"TRBV6-8"	"TRBV6-7"	"TRBV6-6"
[466]	"TRBV6-5"	"TRBV6-4"	"TRBV6-3"	"TRBV6-1"	"TRBV5-8"
[471]	"TRBV5-7"	"TRBV5-6"	"TRBV5-5"	"TRBV5-4"	"TRBV5-3"
[476]	"TRBV5-1"	"TRBV4-3"	"TRBV4-2"	"TRBV4-1"	"TRBV3-1"
[481]	"TRBV2"	"TRBJ2-7"	"TRBJ2-6"	"TRBJ2-5"	"TRBJ2-4"
[486]	"TRBJ2-3"	"TRBJ2-2"	"TRBJ2-1"	"TRBJ1-6"	"TRBJ1-5"
[491]	"TRBJ1-4"	"TRBJ1-3"	"TRBJ1-2"	"TRBJ1-1"	"TRBD1"

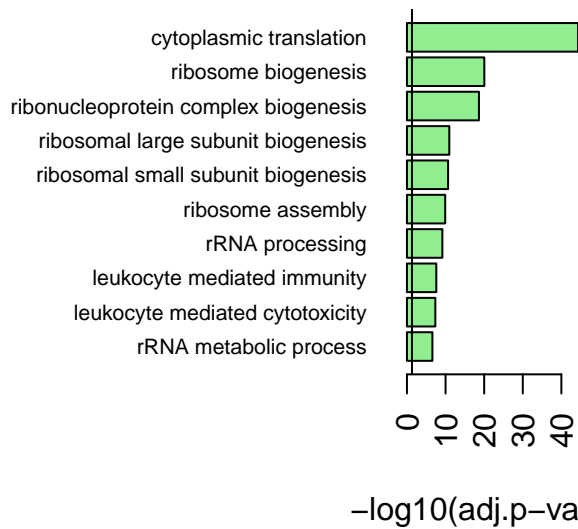
[496]	"TRBC2"	"TRBC1"	"TRAV41"	"TRAV40"	"TRAV39"
[501]	"TRAV38-2DV8"	"TRAV38-1"	"TRAV36DV7"	"TRAV35"	"TRAV34"
[506]	"TRAV30"	"TRAV29DV5"	"TRAV27"	"TRAV26-2"	"TRAV26-1"
[511]	"TRAV25"	"TRAV24"	"TRAV23DV6"	"TRAV22"	"TRAV21"
[516]	"TRAV20"	"TRAV19"	"TRAV18"	"TRAV17"	"TRAV16"
[521]	"TRAV14DV4"	"TRAV13-2"	"TRAV13-1"	"TRAV12-3"	"TRAV12-2"
[526]	"TRAV12-1"	"TRAV10"	"TRAV9-2"	"TRAV9-1"	"TRAV8-6"
[531]	"TRAV8-4"	"TRAV8-3"	"TRAV8-2"	"TRAV8-1"	"TRAV7"
[536]	"TRAV5"	"TRAV4"	"TRAV3"	"TRAV2"	"TRAV1-2"
[541]	"TRAV1-1"	"TRAJ42"	"TRAJ31"	"TRAJ3"	"TRAC"
[546]	"IGLV11-55"	"IGLV10-54"	"IGLV9-49"	"IGLV8-61"	"IGLV7-46"
[551]	"IGLV7-43"	"IGLV6-57"	"IGLV5-52"	"IGLV5-48"	"IGLV5-45"
[556]	"IGLV5-39"	"IGLV5-37"	"IGLV4-69"	"IGLV4-60"	"IGLV4-3"
[561]	"IGLV3-32"	"IGLV3-27"	"IGLV3-25"	"IGLV3-22"	"IGLV3-21"
[566]	"IGLV3-19"	"IGLV3-16"	"IGLV3-12"	"IGLV3-10"	"IGLV3-9"
[571]	"IGLV3-1"	"IGLV2-33"	"IGLV2-23"	"IGLV2-18"	"IGLV2-14"
[576]	"IGLV2-11"	"IGLV2-8"	"IGLV1-51"	"IGLV1-50"	"IGLV1-47"
[581]	"IGLV1-44"	"IGLV1-40"	"IGLV1-36"	"IGLJ1"	"IGLC7"
[586]	"IGKV6D-41"	"IGKV6D-21"	"IGKV3D-20"	"IGKV3D-15"	"IGKV3D-11"
[591]	"IGKV3D-7"	"IGKV2D-30"	"IGKV2D-29"	"IGKV2D-28"	"IGKV2D-26"
[596]	"IGKV2D-24"	"IGKV1D-43"	"IGKV1D-42"	"IGKV1D-39"	"IGKV1D-37"
[601]	"IGKV1D-33"	"IGKV1D-17"	"IGKV1D-13"	"IGKV1D-12"	"IGKV1D-8"
[606]	"IGKV6-21"	"IGKV5-2"	"IGKV4-1"	"IGKV3-20"	"IGKV3-15"
[611]	"IGKV3-7"	"IGKV2-40"	"IGKV2-30"	"IGKV2-29"	"IGKV2-28"
[616]	"IGKV2-24"	"IGKV1-39"	"IGKV1-37"	"IGKV1-27"	"IGKV1-17"
[621]	"IGKV1-16"	"IGKV1-13"	"IGKV1-12"	"IGKV1-9"	"IGKV1-8"
[626]	"IGKV1-6"	"IGKJ1"	"DBNL"	"PYCARD"	"CD274"
[631]	"TBX21"	"CD209"	"IL21R"	"TRAT1"	"CLEC4A"
[636]	"FOXP3"	"EXOSC3"	"ZBTB7B"	"KMT5B"	"LEF1"
[641]	"C1RL"	"TLR8"	"IL23A"	"CYRIB"	"CD244"
[646]	"ERAP1"	"IL20RB"	"TREM2"	"TREM1"	"SASH3"
[651]	"SHLD2"	"RC3H2"	"TRPM4"	"LAX1"	"LIME1"
[656]	"RNF125"	"SUSD4"	"AKIRIN2"	"RIF1"	"OTUB1"
[661]	"PAG1"	"CTNBL1"	"IFNK"	"DUSP22"	"HMCES"
[666]	"OTUD7B"	"ENTPD7"	"MCOLN1"	"IGHV7-4-1"	"IGHV3-30-3"
[671]	"AICDA"	"SLAMF7"	"HMBB1"	"BACH2"	"MYO1G"
[676]	"SAMS1"	"NOD2"	"ERAP2"	"CARD9"	"SEMA4A"
[681]	"NFKBIZ"	"DCLRE1C"	"CLEC7A"	"LILRA6"	"ULBP3"
[686]	"VTCN1"	"BTNL8"	"ATAD5"	"SVEP1"	"ZC3H12A"
[691]	"ULBP2"	"ULBP1"	"PDCD1LG2"	"PRR7"	"NDFIP1"
[696]	"FBX038"	"UNC93B1"	"FCRL4"	"JAM3"	"FCAMR"
[701]	"SLA2"	"SANBR"	"LOXL3"	"CRACR2A"	"KMT5C"
[706]	"NFKBID"	"HAVCR2"	"ORAI1"	"IGHV3-30-5"	"SIGLEC10"

[711]	"KLHL6"	"IL33"	"RSAD2"	"CLEC6A"	"IL17F"
[716]	"NLRP3"	"SLAMF6"	"TNFRSF13C"	"SH2D1B"	"EXOSC6"
[721]	"SLC15A4"	"RNF19B"	"RAET1E"	"RC3H1"	"IL23R"
[726]	"SHLD1"	"BTLA"	"RAET1L"	"DENND1B"	"RNF168"
[731]	"CLEC4C"	"APLF"	"UNC13D"	"GAPT"	"MARCHF8"
[736]	"IL27"	"MCOLN2"	"ZNF683"	"IL4I1"	"NLRP10"
[741]	"CLEC4D"	"IFNE"	"CLEC4G"	"RAET1G"	"NCR3LG1"
[746]	"THEMIS"	"MIR21"	"EIF2AK4"	"TARM1"	"SCART1"
[751]	"CCR2"	"C17orf99"	"IGLL5"	"MICA"	"KLRC4-KLRK1"
[756]	"IGHV2-70D"	"IGHV1-69D"	"IGHV3-64D"	"IGHV3-43D"	"SHLD3"

```
GO_enrich <- enrichGO(
  gene = nk_up_genes,
  OrgDb = org.Hs.eg.db,
  keyType = "SYMBOL",
  ont = "MF", # ont="MF" is the default
  minGSSize = 30, universe = NK_vs_Th$symbol
)
```

Exercise 3 - Visualization of enrichment results

```
par(mar = c(5, 20, 3, 3))
barplot(rev(-log10(GO_NK_Th@result$p.adjust[1:10])),
  horiz = T, names = rev(GO_NK_Th@result$Description[1:10]),
  las = 2, xlab = "-log10(adj.p-value)",
  cex.names = 0.7,
  col = "lightgreen"
)
abline(v = -log10(0.05))
```



```
sorted_GO_NK_Th <- GO_NK_Th@result[order(GO_NK_Th@result$NES, decreasing = F), ]
sorted_GO_NK_Th$colors <- ifelse(sorted_GO_NK_Th$NES > 0, "red", "blue")

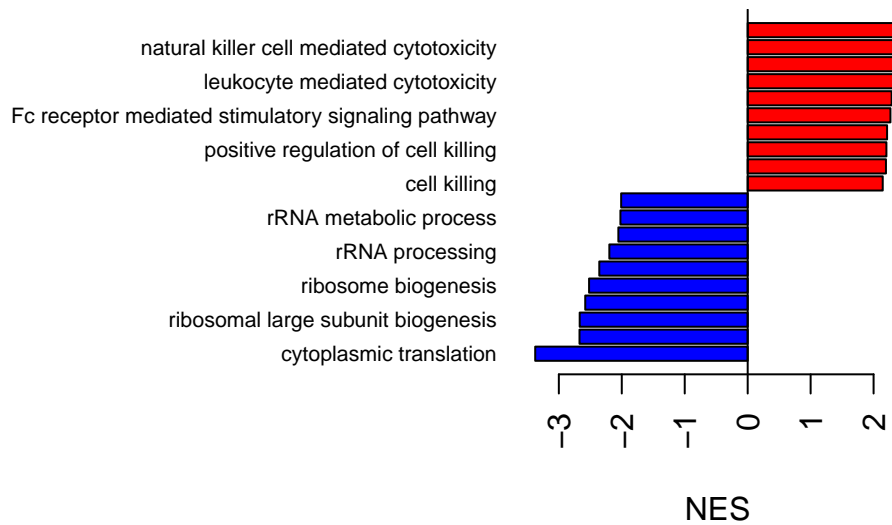
# Get the indices of the vector
bottom_values <- tail(seq_along(sorted_GO_NK_Th$NES), 10)

par(mar = c(5, 15, 3, 3)) # Make the figure canvas larger
barplot(sorted_GO_NK_Th$NES[c(1:10, bottom_values:nrow(sorted_GO_NK_Th))],
        horiz = T, names = sorted_GO_NK_Th$Description[c(1:10, bottom_values:nrow(sorted_GO_NK_Th))],
        las = 2, xlab = "NES",
        cex.names = 0.7,
        col = sorted_GO_NK_Th$color[c(1:10, (nrow(sorted_GO_NK_Th) - 9):nrow(sorted_GO_NK_Th))])
```

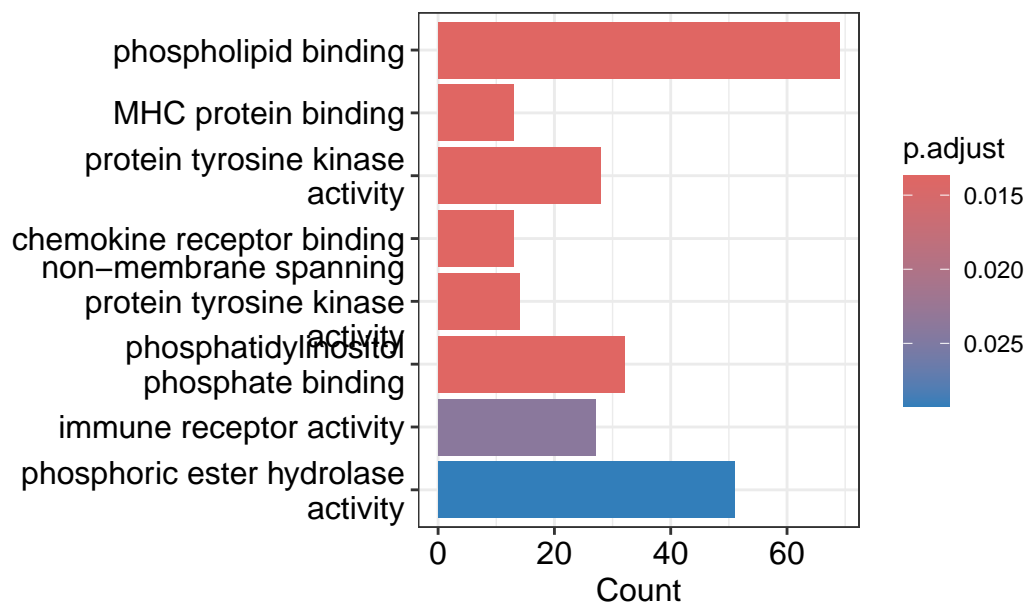
Warning in bottom_values:nrow(sorted_GO_NK_Th): numerical expression has 10 elements: only the first used

Warning in bottom_values:nrow(sorted_GO_NK_Th): numerical expression has 10 elements: only the first used

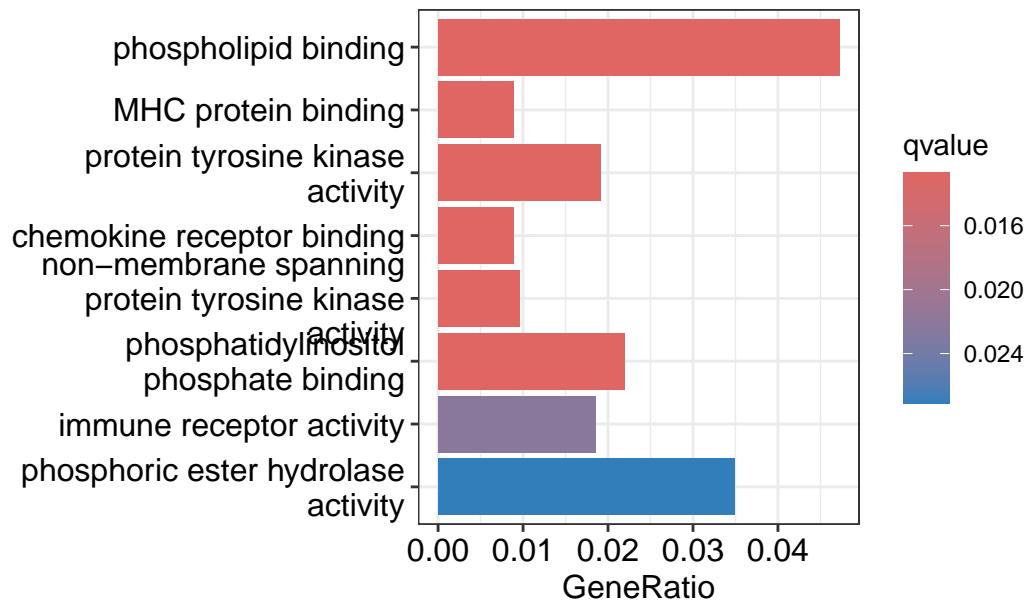
```
abline(v = 0)
```



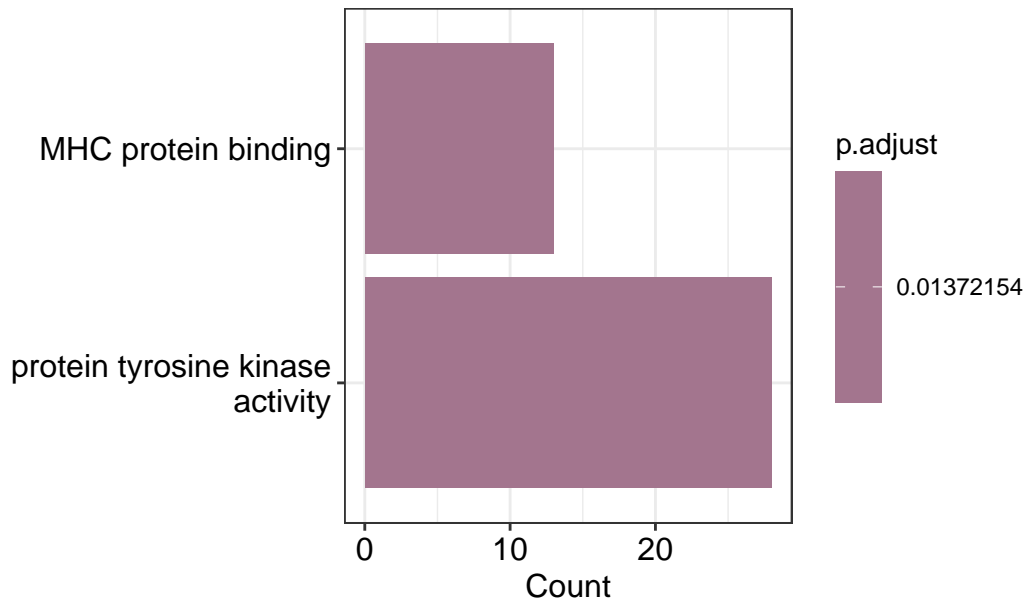
```
# Use the GO_enrich analysis performed above, of the over-representation analysis
# of genes up-regulated in NK cells:
# barplot() can be directly used on enrichResult objects: but not on gseaResult objects
graphics::barplot(GO_enrich)
```



```
graphics::barplot(GO_enrich, color = "qvalue", x = "GeneRatio")
```



```
# Select only 2 out of the significant gene sets:
ego_selection <- GO_enrich[GO_enrich$result$ID == "GO:0042287" | GO_enrich$result$ID == "GO:0005488"]
barplot(ego_selection)
```



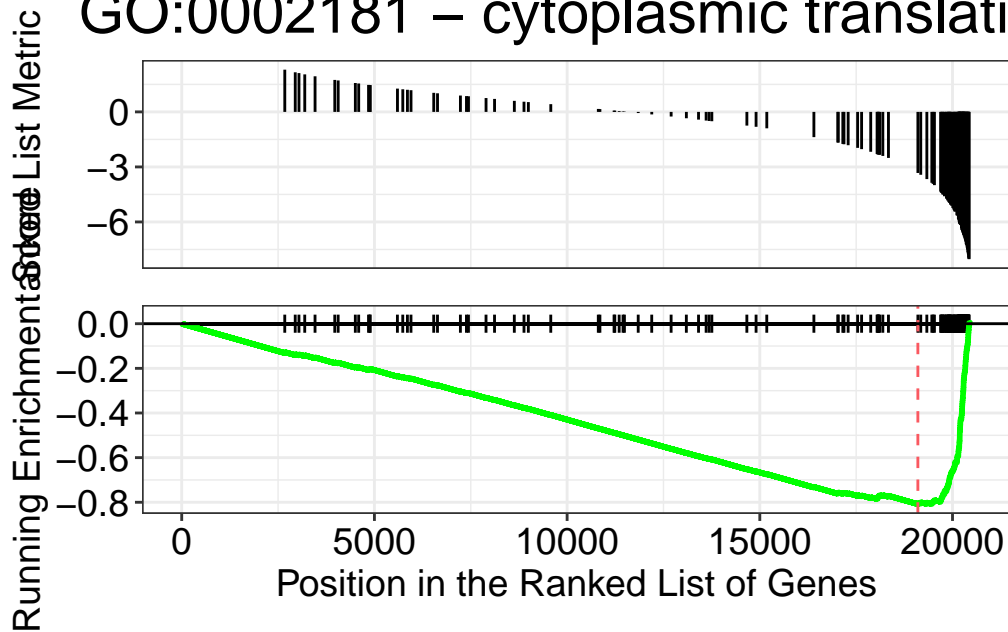
```
# Barcode plot
# You need the ID of the GO gene set to plot:
GO_NK_Th$result[1:10, 1:6]
```


ID	Description	setSize
G0:0002181 G0:0002181	cytoplasmic translation	145
G0:0042254 G0:0042254	ribosome biogenesis	299
G0:0022613 G0:0022613	ribonucleoprotein complex biogenesis	436
G0:0042273 G0:0042273	ribosomal large subunit biogenesis	69
G0:0042274 G0:0042274	ribosomal small subunit biogenesis	98
G0:0042255 G0:0042255	ribosome assembly	57
G0:0006364 G0:0006364	rRNA processing	210
G0:0002443 G0:0002443	leukocyte mediated immunity	340
G0:0001909 G0:0001909	leukocyte mediated cytotoxicity	111
G0:0016072 G0:0016072	rRNA metabolic process	242

	enrichmentScore	NES	pvalue
G0:0002181	-0.8083663	-3.375135	1.684522e-48
G0:0042254	-0.5505406	-2.519660	6.298394e-24
G0:0022613	-0.4906673	-2.357458	2.339798e-22
G0:0042273	-0.7146074	-2.668647	1.478487e-14
G0:0042274	-0.6484741	-2.580517	3.835801e-14
G0:0042255	-0.7361084	-2.671760	2.555521e-13
G0:0006364	-0.5004485	-2.197559	1.614149e-12
G0:0002443	0.4169045	1.955354	6.943557e-11
G0:0001909	0.5720092	2.333365	1.381646e-10
G0:0016072	-0.4491031	-2.021848	8.811745e-10

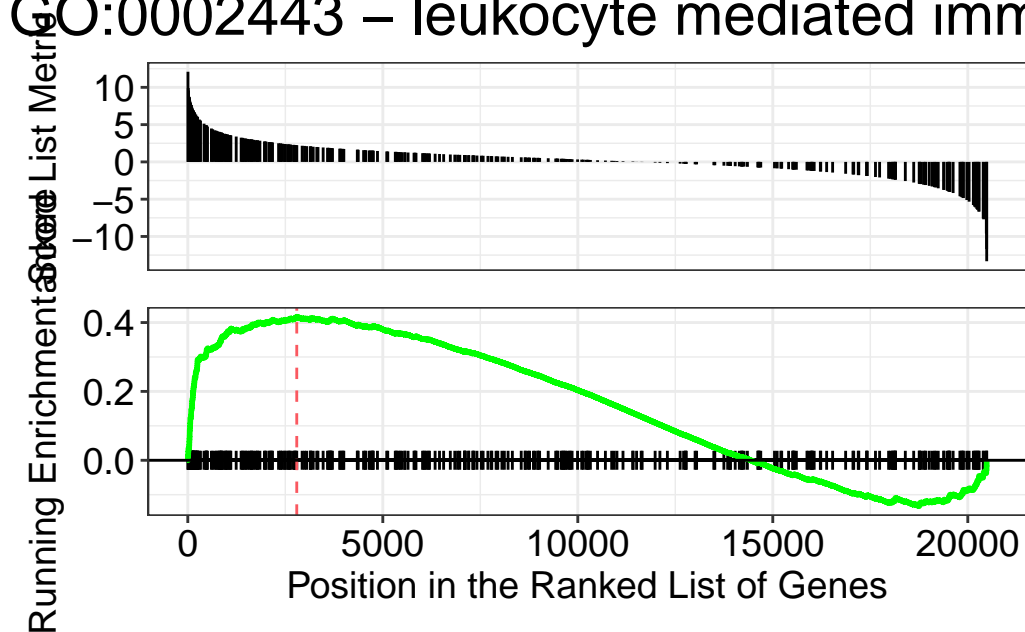
```
# For a gene set that is down-regulated in NK cells:
gseaplot(GO_NK_Th,
  geneSetID = "G0:0002181",
  title = "G0:0002181 - cytoplasmic translation"
)
```

GO:0002181 – cytoplasmic translation

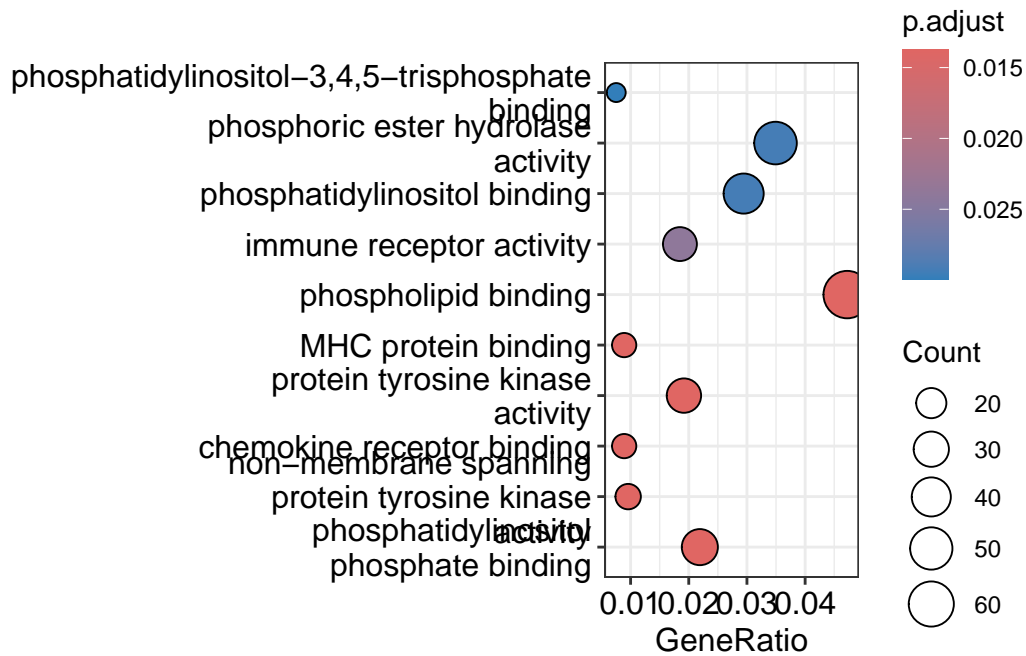


```
# And one that is up-regulated in NK cells
gseaplot(GO_NK_Th,
  geneSetID = "GO:0002443",
  title = "GO:0002443 – leukocyte mediated immunity"
)
```

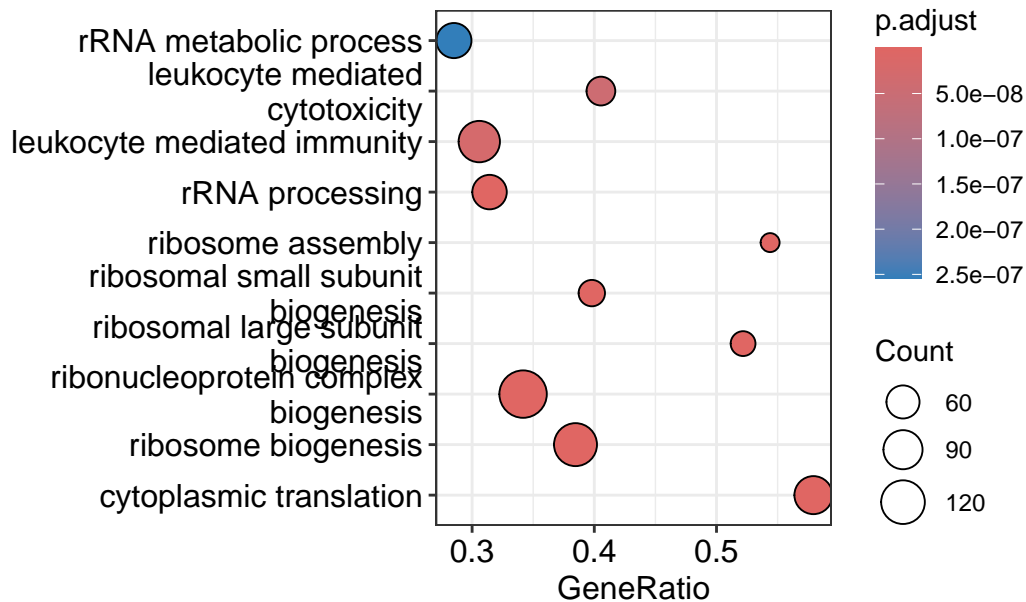
GO:0002443 – leukocyte mediated immunity



```
enrichplot::dotplot(GO_enrich, orderBy = "p.adjust")
```

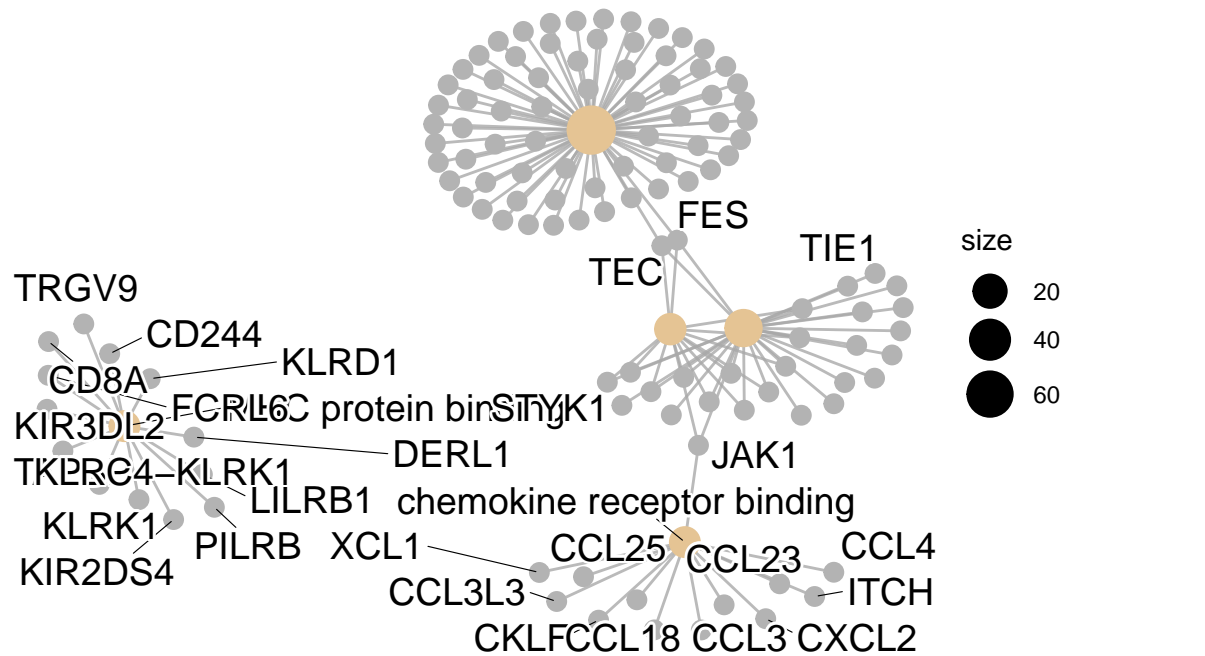


```
enrichplot::dotplot(GO_NK_Th, orderBy = "p.adjust")
```



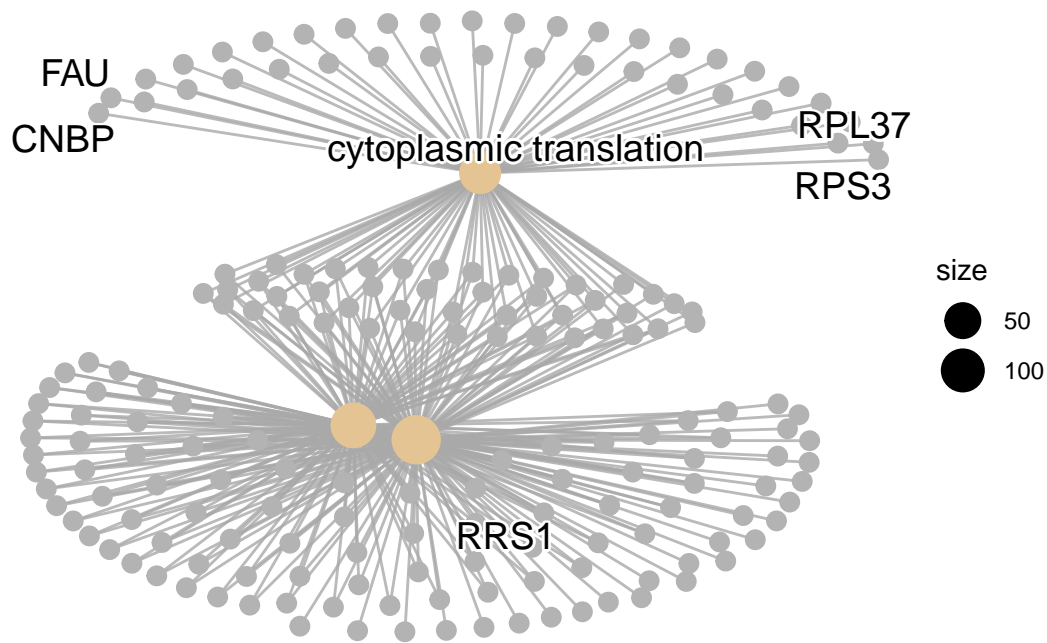
```
cnetplot(GO_enrich, categorySize = "pvalue")
```

Warning: ggrepel: 95 unlabeled data points (too many overlaps). Consider increasing max.overlaps



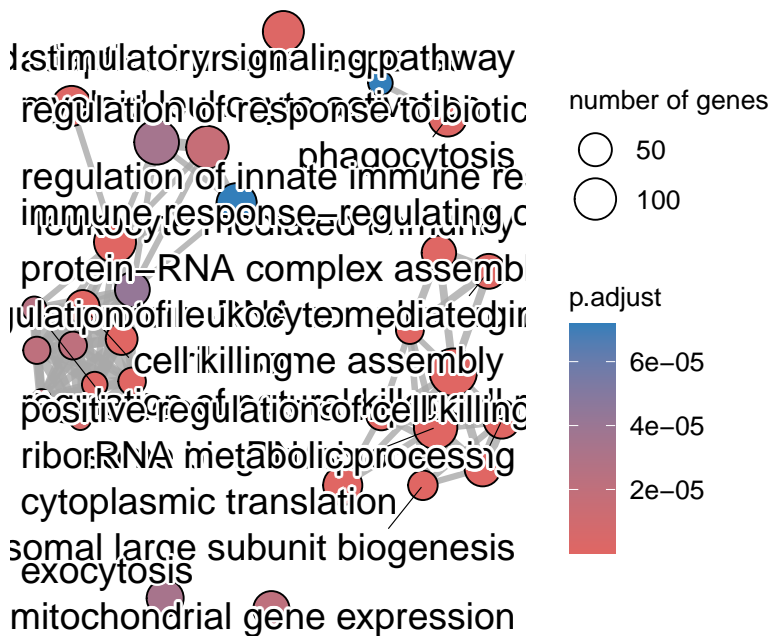
```
cnetplot(GO_NK_Th, showCategory = 3)
```

Warning: ggrepel: 185 unlabeled data points (too many overlaps). Consider increasing max.overlaps



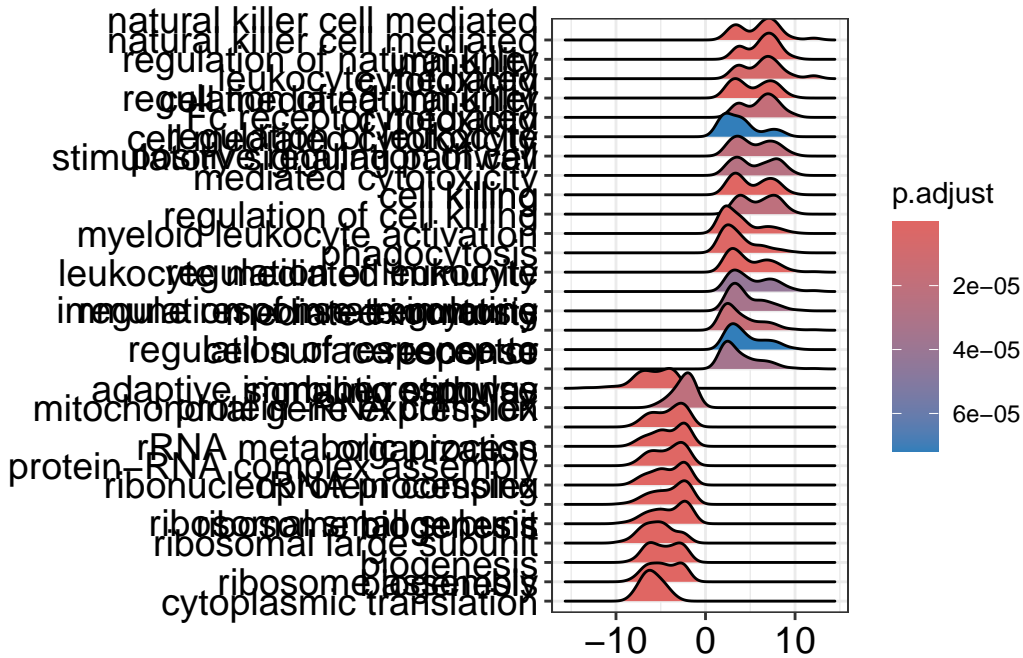
```
ego2 <- pairwise_termsim(GO_NK_Th)
emapplot(ego2, color = "p.adjust")
```

Warning: ggrepel: 7 unlabeled data points (too many overlaps). Consider increasing max.overlaps



```
# Distribution of t-statistic for genes included in significant gene sets or in selected gene
# ar(mar = c(15, 30, 3, 3))
ridgeplot(GO_NK_Th)
```

Picking joint bandwidth of 0.787

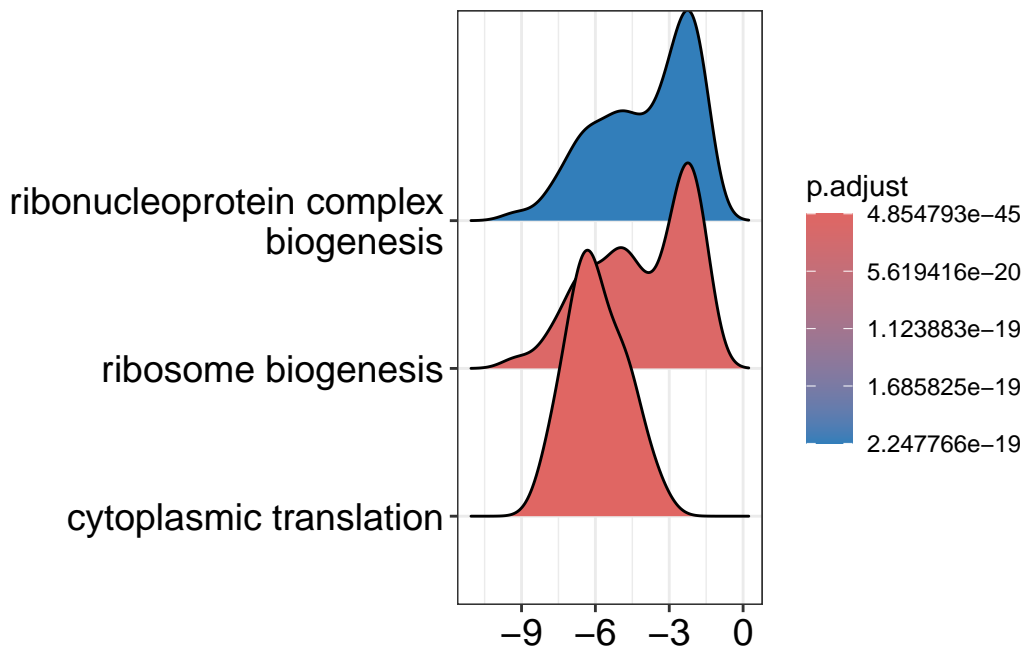


```
# What is the difference with core_enrichment =F?
ridgeplot(GO_NK_Th, core_enrichment = FALSE)
```

Picking joint bandwidth of 0.975

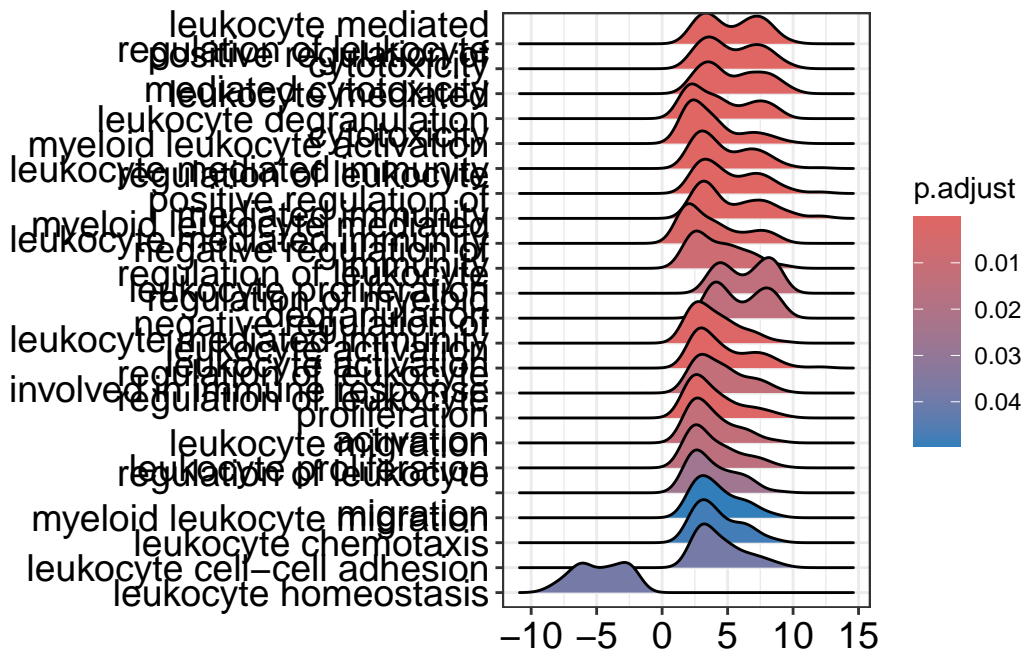

```
GO_NK_Th_selection <- GO_NK_Th[
  GO_NK_Th$ID %in% c(
    "GO:0002181", "GO:0022613",
    "GO:0042254"
  ),
  asis = TRUE
]
ridgeplot(GO_NK_Th_selection)
```

Picking joint bandwidth of 0.589



```
# Terms that contain the keyword "leukocyte"
GO_NK_Th_selection <- GO_NK_Th[grep("leukocyte", GO_NK_Th@result$Description), asis = TRUE]
ridgeplot(GO_NK_Th_selection)
```

Picking joint bandwidth of 0.839



Exercise 4 - Enrichment of other collections of gene sets

```
keytypes(org.Hs.eg.db)
```

```
[1] "ACCNUM"      "ALIAS"       "ENSEMBL"     "ENSEMBLPROT" "ENSEMBLTRANS"
[6] "ENTREZID"    "ENZYME"      "EVIDENCE"    "EVIDENCEALL"  "GENENAME"
[11] "GENETYPE"    "GO"          "GOALL"       "IPI"          "MAP"
[16] "OMIM"        "ONTOLOGY"    "ONTOLOGYALL" "PATH"         "PFAM"
[21] "PMID"        "PROSITE"     "REFSEQ"      "SYMBOL"       "UCSCKG"
[26] "UNIPROT"
```

```
# convert from= "ENSEMBL" to "SYMBOL" and "ENTREZID"
gene_convert <- bitr(as.character(NK_vs_Th$ensembl_gene_id),
  fromType = "ENSEMBL",
  toType = c("SYMBOL", "ENTREZID"), OrgDb = "org.Hs.eg.db"
)
```

'select()' returned 1:many mapping between keys and columns

Warning in bitr(as.character(NK_vs_Th\$ensembl_gene_id), fromType = "ENSEMBL", :
 18.73% of input gene IDs are fail to map...

```
# Check the format of the data frame obtained after conversion:
head(gene_convert)
```

```
      ENSEMBL SYMBOL ENTREZID
1 ENSG00000000003 TSPAN6      7105
2 ENSG000000000419 DPM1       8813
3 ENSG000000000457 SCYL3     57147
4 ENSG000000000460 FIRRM     55732
5 ENSG000000000938 FGR       2268
6 ENSG000000000971 CFH       3075
```

```
dim(gene_convert)
```

```
[1] 16794      3
```

```
# Create a vector of genes that are coded with the EntrezID:
# use the sorted gene list gl previously created:
gl_kegg <- cbind(SYMBOL = names(gl), t = gl)

# merge with converted gene symbols to combine both:
# by default the data frames are merged on the columns with names they both have
gl_kegg <- merge(gl_kegg, gene_convert)
head(gl_kegg)
```

```
      SYMBOL      t      ENSEMBL ENTREZID
1  A1BG  1.129187394 ENSG00000121410      1
2   A2M -0.382294217 ENSG00000175899      2
3 A4GALT 0.808365644 ENSG00000128274    53947
4  AAAS  0.749990903 ENSG00000094914     8086
5  AACS  2.172253591 ENSG00000081760    65985
6 AADAT 3.038354213 ENSG00000109576    51166
```

```
gl_kegg_list <- as.numeric(as.character(gl_kegg$t))
names(gl_kegg_list) <- as.character(gl_kegg$ENTREZID)
gl_kegg_list <- sort(gl_kegg_list, decreasing = T)
```

```
# run GSEA of KEGG (please note that requires internet connection to download the KEGG annotations)
KEGG_NK_Th <- gseKEGG(gl_kegg_list,
  organism = "hsa", "ncbi-geneid",
  minGSSize = 30,
```

```

    eps = 0,
    seed = T
)

```

Reading KEGG annotation online: "https://rest.kegg.jp/link/hsa/pathway"...

Reading KEGG annotation online: "https://rest.kegg.jp/list/pathway/hsa"...

Reading KEGG annotation online: "https://rest.kegg.jp/conv/ncbi-geneid/hsa"...

preparing geneSet collections...

GSEA analysis...

Warning in preparePathwaysAndStats(pathways, stats, minSize, maxSize, gseaParam, : There are
The order of those tied genes will be arbitrary, which may produce unexpected results.

Warning in preparePathwaysAndStats(pathways, stats, minSize, maxSize,
gseaParam, : There are duplicate gene names, fgsea may produce unexpected
results.

leading edge analysis...

done...

```

# What does it contain?
str(KEGG_NK_Th)

```

Formal class 'gseaResult' [package "DOSE"] with 13 slots

..@ result : 'data.frame': 24 obs. of 11 variables:

```

.. ..$ ID          : chr [1:24] "hsa03010" "hsa05171" "hsa04650" "hsa04666" ...
.. ..$ Description  : chr [1:24] "Ribosome" "Coronavirus disease - COVID-19" "Natural ki
.. ..$ setSize      : int [1:24] 130 185 98 86 163 181 93 77 118 220 ...
.. ..$ enrichmentScore: num [1:24] -0.813 -0.678 0.62 0.525 0.426 ...
.. ..$ NES          : num [1:24] -3.46 -3.02 2.49 2.07 1.87 ...
.. ..$ pvalue       : num [1:24] 3.32e-46 5.30e-32 2.03e-12 1.23e-06 2.91e-06 ...
.. ..$ p.adjust     : num [1:24] 8.90e-44 7.11e-30 1.81e-10 8.21e-05 1.56e-04 ...
.. ..$ qvalue       : num [1:24] 7.27e-44 5.81e-30 1.48e-10 6.71e-05 1.27e-04 ...

```

```

.. ..$ rank          : num [1:24] 1852 1168 1873 1924 2216 ...
.. ..$ leading_edge   : chr [1:24] "tags=72%, list=13%, signal=63%" "tags=48%, list=8%, si
.. ..$ core_enrichment: chr [1:24] "63875/140032/51121/64983/6139/9553/51021/51116/6133/51
..@ organism          : chr "hsa"
..@ setType           : chr "KEGG"
..@ geneSets          :List of 365
.. ..$ hsa00010: chr [1:67] "10327" "124" "125" "126" ...
.. ..$ hsa00020: chr [1:30] "1431" "1737" "1738" "1743" ...
.. ..$ hsa00030: chr [1:31] "132158" "2203" "221823" "226" ...
.. ..$ hsa00040: chr [1:36] "10327" "10720" "10941" "231" ...
.. ..$ hsa00051: chr [1:34] "197258" "2203" "226" "229" ...
.. ..$ hsa00052: chr [1:32] "130589" "231" "2538" "2548" ...
.. ..$ hsa00053: chr [1:30] "10327" "10720" "10941" "217" ...
.. ..$ hsa00061: chr [1:18] "109703458" "197322" "2180" "2181" ...
.. ..$ hsa00062: chr [1:27] "10449" "10965" "11332" "117145" ...
.. ..$ hsa00071: chr [1:43] "10449" "10455" "113612" "124" ...
.. ..$ hsa00100: chr [1:20] "1056" "10682" "120227" "1591" ...
.. ..$ hsa00120: chr [1:17] "10005" "10858" "10998" "1109" ...
.. ..$ hsa00130: chr [1:12] "10229" "154807" "1728" "2677" ...
.. ..$ hsa00140: chr [1:63] "100861540" "10720" "10941" "1109" ...
.. ..$ hsa00190: chr [1:138] "100532726" "10063" "101927180" "10312" ...
.. ..$ hsa00220: chr [1:23] "100526760" "1373" "137362" "162417" ...
.. ..$ hsa00230: chr [1:128] "100" "100526794" "10201" "102157402" ...
.. ..$ hsa00232: chr [1:6] "10" "1544" "1548" "1549" ...
.. ..$ hsa00240: chr [1:58] "100526794" "10201" "115024" "124583" ...
.. ..$ hsa00250: chr [1:37] "122622" "1373" "137362" "158" ...
.. ..$ hsa00260: chr [1:41] "102724560" "10993" "113675" "124908081" ...
.. ..$ hsa00270: chr [1:52] "102724560" "1036" "10768" "10993" ...
.. ..$ hsa00280: chr [1:48] "10449" "11112" "1629" "1738" ...
.. ..$ hsa00290: chr [1:4] "10993" "113675" "586" "587"
.. ..$ hsa00310: chr [1:63] "10157" "10919" "11105" "123688" ...
.. ..$ hsa00330: chr [1:50] "112483" "112817" "112849" "113451" ...
.. ..$ hsa00340: chr [1:22] "10841" "131669" "138199" "144193" ...
.. ..$ hsa00350: chr [1:36] "124" "125" "126" "127" ...
.. ..$ hsa00360: chr [1:16] "137362" "1644" "218" "221" ...
.. ..$ hsa00380: chr [1:42] "11185" "121278" "125061" "130013" ...
.. ..$ hsa00400: chr [1:6] "137362" "259307" "2805" "2806" ...
.. ..$ hsa00410: chr [1:31] "18" "1806" "1807" "1892" ...
.. ..$ hsa00430: chr [1:17] "102724197" "1036" "124975" "2326" ...
.. ..$ hsa00440: chr [1:6] "10390" "5130" "56994" "5833" ...
.. ..$ hsa00450: chr [1:17] "10587" "11185" "114112" "118672" ...
.. ..$ hsa00470: chr [1:6] "1610" "27165" "2744" "63826" ...
.. ..$ hsa00480: chr [1:59] "102724197" "10314" "119391" "124975" ...

```

```

.. ..$ hsa00500: chr [1:40] "11181" "124905666" "124905668" "128966568" ...
.. ..$ hsa00510: chr [1:54] "10195" "10905" "11253" "11282" ...
.. ..$ hsa00511: chr [1:18] "10825" "129807" "175" "23324" ...
.. ..$ hsa00512: chr [1:36] "100528030" "10331" "10610" "11226" ...
.. ..$ hsa00513: chr [1:42] "10195" "10905" "11253" "11282" ...
.. ..$ hsa00514: chr [1:47] "100528030" "10585" "11226" "11227" ...
.. ..$ hsa00515: chr [1:23] "10329" "10585" "10690" "11041" ...
.. ..$ hsa00520: chr [1:38] "10007" "10020" "1118" "132789" ...
.. ..$ hsa00524: chr [1:5] "2645" "3098" "3099" "3101" ...
.. ..$ hsa00531: chr [1:19] "10855" "138050" "23553" "2588" ...
.. ..$ hsa00532: chr [1:21] "10090" "11285" "113189" "126792" ...
.. ..$ hsa00533: chr [1:14] "10164" "10678" "2530" "2683" ...
.. ..$ hsa00534: chr [1:24] "11285" "126792" "2131" "2132" ...
.. ..$ hsa00541: chr [1:16] "10020" "123956252" "140838" "1727" ...
.. ..$ hsa00561: chr [1:65] "10327" "10554" "10555" "1056" ...
.. ..$ hsa00562: chr [1:73] "10423" "113026" "138429" "200576" ...
.. ..$ hsa00563: chr [1:30] "10026" "128869" "23556" "27315" ...
.. ..$ hsa00564: chr [1:103] "100137049" "10162" "10390" "1040" ...
.. ..$ hsa00565: chr [1:50] "100137049" "10390" "11145" "122618" ...
.. ..$ hsa00590: chr [1:63] "100137049" "102724197" "10728" "11145" ...
.. ..$ hsa00591: chr [1:30] "100137049" "11145" "123745" "151056" ...
.. ..$ hsa00592: chr [1:26] "100137049" "11145" "123745" "151056" ...
.. ..$ hsa00600: chr [1:54] "10558" "10715" "10825" "123099" ...
.. ..$ hsa00601: chr [1:28] "10317" "10331" "10402" "10678" ...
.. ..$ hsa00603: chr [1:16] "10317" "10690" "127550" "2523" ...
.. ..$ hsa00604: chr [1:15] "256435" "2583" "27090" "2720" ...
.. ..$ hsa00620: chr [1:47] "10327" "10873" "124" "125" ...
.. ..$ hsa00630: chr [1:31] "112817" "124908081" "125061" "132158" ...
.. ..$ hsa00640: chr [1:32] "160287" "1629" "1738" "18" ...
.. ..$ hsa00650: chr [1:27] "116285" "123876" "142827" "18" ...
.. ..$ hsa00670: chr [1:39] "100528021" "102724560" "10588" "10768" ...
.. ..$ hsa00730: chr [1:15] "122481" "158067" "203" "204" ...
.. ..$ hsa00740: chr [1:8] "5167" "5169" "52" "53" ...
.. ..$ hsa00750: chr [1:6] "29968" "316" "493911" "55163" ...
.. ..$ hsa00760: chr [1:38] "100526794" "10135" "133686" "22933" ...
.. ..$ hsa00770: chr [1:21] "1806" "1807" "217" "219" ...
.. ..$ hsa00780: chr [1:3] "3141" "54995" "686"
.. ..$ hsa00785: chr [1:20] "11019" "116285" "124908081" "1629" ...
.. ..$ hsa00790: chr [1:28] "10243" "121278" "1719" "200895" ...
.. ..$ hsa00830: chr [1:68] "100861540" "10170" "10720" "10901" ...
.. ..$ hsa00860: chr [1:46] "10720" "10941" "124454" "1352" ...
.. ..$ hsa00900: chr [1:23] "100529261" "10269" "10654" "116150" ...
.. ..$ hsa00910: chr [1:17] "11238" "1373" "23632" "2746" ...

```

```

.. ..$ hsa00920: chr [1:10] "10380" "23474" "4357" "54928" ...
.. ..$ hsa00970: chr [1:66] "10056" "10352" "10667" "118672" ...
.. ..$ hsa00980: chr [1:79] "10720" "107987478" "107987479" "10941" ...
.. ..$ hsa00982: chr [1:73] "10720" "107987478" "107987479" "10941" ...
.. ..$ hsa00983: chr [1:81] "10" "10201" "1066" "10720" ...
.. ..$ hsa01040: chr [1:27] "10965" "11332" "122970" "201562" ...
.. ..$ hsa01100: chr [1:1570] "10" "100" "10005" "10007" ...
.. ..$ hsa01200: chr [1:116] "10873" "10993" "113675" "124908081" ...
.. ..$ hsa01210: chr [1:33] "100526760" "137362" "1431" "162417" ...
.. ..$ hsa01212: chr [1:57] "10449" "109703458" "126129" "1374" ...
.. ..$ hsa01230: chr [1:75] "100526760" "102724560" "10993" "113675" ...
.. ..$ hsa01232: chr [1:85] "100" "100526794" "10201" "102157402" ...
.. ..$ hsa01240: chr [1:154] "10201" "102157402" "10229" "10243" ...
.. ..$ hsa01250: chr [1:37] "10020" "140838" "197258" "23483" ...
.. ..$ hsa01320: chr [1:2] "9060" "9061"
.. ..$ hsa01521: chr [1:80] "10000" "10018" "110117499" "1950" ...
.. ..$ hsa01522: chr [1:99] "10000" "1019" "1026" "1027" ...
.. ..$ hsa01523: chr [1:30] "10057" "10257" "113235" "1147" ...
.. ..$ hsa01524: chr [1:75] "10000" "1026" "1029" "110117499" ...
.. .. [list output truncated]
..@ geneList : Named num [1:14284] 19 13.1 12.1 12 10.7 ...
..- attr(*, "names")= chr [1:14284] "2693" "23209" "117157" "6983" ...
..@ keytype : chr "ncbi-geneid"
..@ permScores : num[0 , 0 ]
..@ params :List of 6
.. ..$ pvalueCutoff : num 0.05
.. ..$ eps : num 0
.. ..$ pAdjustMethod: chr "BH"
.. ..$ exponent : num 1
.. ..$ minGSSize : num 30
.. ..$ maxGSSize : num 500
..@ gene2Symbol: chr(0)
..@ readable : logi FALSE
..@ termsim : num[0 , 0 ]
..@ method : chr(0)
..@ dr : list()

```

```

# How many gene sets are up-regulated?
sum(KEGG_NK_Th@result$NES > 0) # 17

```

```
[1] 17
```

```
#|
grep_kegg_description <- function(pattern) {
  return(grep(pattern, tolower((KEGG_NK_Th@result$Description))))
}
```

```
# Is there an immune-related gene set significant?
grep_kegg_description("immune")
```

```
integer(0)
```

```
# Is there an NK gene set significant?
grep_kegg_description("natural killer") # 3
```

```
[1] 3
```

```
# What is the total number of built-in KEGG gene sets?
length(KEGG_NK_Th@geneSets) # 265
```

```
[1] 365
```

```
KEGG_NK_Th[grep_kegg_description("natural killer"), ] |>
  select(ID, Description) # hsa04650
```

	ID	Description
hsa04650	hsa04650	Natural killer cell mediated cytotoxicity

```
KEGG_NK_Th@geneSets$hsa04650
```

[1]	"100132285"	"100507436"	"100528032"	"102723407"	"10451"	"10870"
[7]	"110117499"	"117157"	"124905743"	"135250"	"1437"	"154064"
[13]	"2185"	"2207"	"2214"	"2215"	"22914"	"2534"
[19]	"25759"	"259197"	"27040"	"2885"	"3002"	"3105"
[25]	"3106"	"3107"	"3133"	"3135"	"3265"	"3383"
[31]	"3384"	"3439"	"3440"	"3441"	"3442"	"3443"
[37]	"3444"	"3445"	"3446"	"3447"	"3448"	"3449"
[43]	"3451"	"3452"	"3454"	"3455"	"3456"	"3458"
[49]	"3459"	"3460"	"353091"	"355"	"356"	"3683"
[55]	"3689"	"369"	"3802"	"3803"	"3804"	"3805"

[61]	"3806"	"3808"	"3809"	"3810"	"3811"	"3812"
[67]	"3821"	"3822"	"3823"	"3824"	"3845"	"3932"
[73]	"3937"	"399694"	"4068"	"4277"	"4772"	"4773"
[79]	"4893"	"5058"	"51744"	"5290"	"5291"	"5293"
[85]	"5295"	"5296"	"5335"	"53358"	"5336"	"5530"
[91]	"5532"	"5533"	"5534"	"5535"	"5551"	"5578"
[97]	"5579"	"5582"	"5594"	"5595"	"5604"	"5605"
[103]	"57292"	"5777"	"5781"	"5879"	"5880"	"5881"
[109]	"5894"	"637"	"6452"	"6464"	"6654"	"6655"
[115]	"673"	"6850"	"7124"	"7305"	"7409"	"7410"
[121]	"7462"	"7535"	"79465"	"80328"	"80329"	"836"
[127]	"8503"	"8743"	"8795"	"8797"	"919"	"9436"
[133]	"9437"	"962"				

```
# pathview map with non-significant genes in grey:
# set log fold change of non-significant genes to 0:
NK_vs_Th$logFC_0 <- ifelse(NK_vs_Th$p.adj > 0.05, 0, NK_vs_Th$logFC)

# create named vector of fold change values:
genePW <- NK_vs_Th$logFC_0
names(genePW) <- NK_vs_Th$symbol

# Create pathview map for Ribosome = hsa03010
pathview(
  gene.data = genePW,
  pathway.id = "hsa03010",
  species = "hsa",
  gene.idtype = "SYMBOL"
)
```

'select()' returned 1:many mapping between keys and columns

```
[1] "Note: 4806 of 20411 unique input IDs unmapped."
```

'select()' returned 1:1 mapping between keys and columns

Info: Working in directory /var/home/artur/Documents/10-19_PhD/11_Education/11.22-sib-enrich

Info: Writing image file hsa03010.pathview.png


```
# Create pathview map of Natural killer cell mediated cytotoxicity = hsa04650
pathview(
  gene.data = genePW,
  pathway.id = "hsa04650",
  species = "hsa",
  gene.idtype = "SYMBOL"
)
```

'select()' returned 1:many mapping between keys and columns

```
[1] "Note: 4806 of 20411 unique input IDs unmapped."
```

'select()' returned 1:1 mapping between keys and columns

Info: Working in directory /var/home/artur/Documents/10-19_PhD/11_Education/11.22-sib-enrich

Info: Writing image file hsa04650.pathview.png

```
# Import hallmark, convert to term2gene and run GSEA:
term2gene_h <- msigdb(species = "Homo sapiens", category = "H")
# Or alternatively:
# term2gene_h<-read.gmt("h.all.v2023.2.Hs.symbols.gmt")

head(term2gene_h)
```

```
# A tibble: 6 x 15
  gs_cat gs_subcat gs_name          gene_symbol entrez_gene ensembl_gene
  <chr>  <chr>      <chr>          <chr>          <int> <chr>
1 H      ""        HALLMARK_ADIPOGENESIS ABCA1             19 ENSG00000165029
2 H      ""        HALLMARK_ADIPOGENESIS ABCB8            11194 ENSG00000197150
3 H      ""        HALLMARK_ADIPOGENESIS ACAA2            10449 ENSG00000167315
4 H      ""        HALLMARK_ADIPOGENESIS ACADL             33 ENSG00000115361
5 H      ""        HALLMARK_ADIPOGENESIS ACADM             34 ENSG00000117054
6 H      ""        HALLMARK_ADIPOGENESIS ACADS             35 ENSG00000122971
# i 9 more variables: human_gene_symbol <chr>, human_entrez_gene <int>,
#   human_ensembl_gene <chr>, gs_id <chr>, gs_pmid <chr>, gs_geoid <chr>,
#   gs_exact_source <chr>, gs_url <chr>, gs_description <chr>
```

```
length(unique(term2gene_h$gs_name)) # 50
```

```
[1] 50
```

```
# Run GSEA with the function that allows to use custom gene sets,  
# provide the named vector of t statistics  
h_NK_vs_Th <- GSEA(gl,  
  TERM2GENE = term2gene_h[, c("gs_name", "gene_symbol")],  
  eps = 0,  
  seed = T  
)
```

preparing geneSet collections...

GSEA analysis...

Warning in preparePathwaysAndStats(pathways, stats, minSize, maxSize, gseaParam, : There are
The order of those tied genes will be arbitrary, which may produce unexpected results.

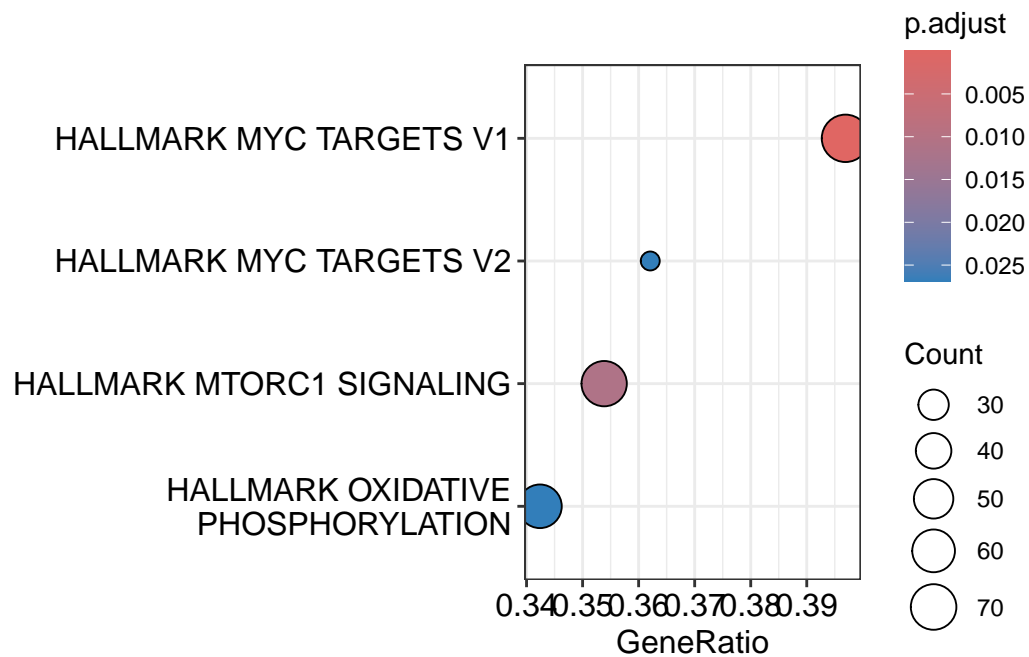
leading edge analysis...

done...

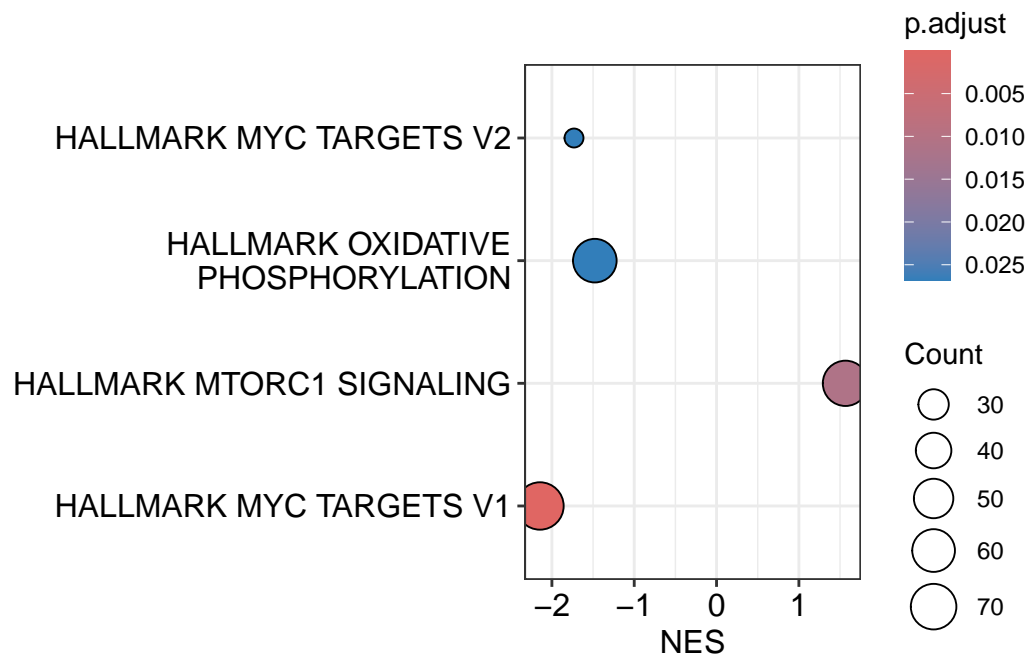
```
# Number of significant gene sets:  
length(which(h_NK_vs_Th@result$p.adjust <= 0.05))
```

```
[1] 4
```

```
# A dotplot with geneRatio or NES on the x-axis:  
dotplot(h_NK_vs_Th)
```



```
dotplot(h_NK_vs_Th, x = "NES", orderBy = "p.adjust")
```



```
# A barcode plot:  
gseaplot2(h_NK_vs_Th,
```

```
geneSetID = "HALLMARK_MTORC1_SIGNALING",  
title = "HALLMARK_MTORC1_SIGNALING"  
)
```

