

GO enrichment

HBG

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This script perform GO enrichment using topGO for a set of genes that show high (top 10%)

Bayes factors in tumor cells, but low or moderate (0%-50%) BF in normal cells

topGO analysis

```
# loading the libraries
library(topGO)
```

```
## Loading required package: BiocGenerics
## Loading required package: parallel
##
## Attaching package: 'BiocGenerics'
## The following objects are masked from 'package:parallel':
##
##   clusterApply, clusterApplyLB, clusterCall, clusterEvalQ,
##   clusterExport, clusterMap, parApply, parCapply, parLapply,
##   parLapplyLB, parRapply, parSapply, parSapplyLB
## The following objects are masked from 'package:stats':
##
##   IQR, mad, sd, var, xtabs
## The following objects are masked from 'package:base':
##
##   anyDuplicated, append, as.data.frame, basename, cbind,
##   colMeans, colnames, colSums, dirname, do.call, duplicated,
##   eval, evalq, Filter, Find, get, grep, grepl, intersect,
##   is.unsorted, lapply, lengths, Map, mapply, match, mget, order,
##   paste, pmax, pmax.int, pmin, pmin.int, Position, rank, rbind,
##   Reduce, rowMeans, rownames, rowSums, sapply, setdiff, sort,
##   table, tapply, union, unique, unsplit, which, which.max,
##   which.min
## Loading required package: graph
## Loading required package: Biobase
## Welcome to Bioconductor
##
##   Vignettes contain introductory material; view with
##   'browseVignettes()'. To cite Bioconductor, see
##   'citation("Biobase)"', and for packages 'citation("pkgname)".
## Loading required package: GO.db
```

```

## Loading required package: AnnotationDbi
## Loading required package: stats4
## Loading required package: IRanges
## Loading required package: S4Vectors
##
## Attaching package: 'S4Vectors'
## The following object is masked from 'package:base':
##
##     expand.grid
##
## Loading required package: SparseM
##
## Attaching package: 'SparseM'
## The following object is masked from 'package:base':
##
##     backsolve
##
## groupGOTerms:      GOBPTerm, GOMFTerm, GOCCTerm environments built.
##
## Attaching package: 'topGO'
## The following object is masked from 'package:IRanges':
##
##     members
library(biomaRt)
library(ensemblDb)

## Loading required package: GenomicRanges
## Loading required package: GenomeInfoDb
## Loading required package: GenomicFeatures
##
## Attaching package: 'GenomicFeatures'
## The following object is masked from 'package:topGO':
##
##     genes
## Loading required package: AnnotationFilter
##
## Attaching package: 'ensemblDb'
## The following object is masked from 'package:stats':
##
##     filter
#library(mygene) ## it may be useful for gene name conversion
library(EnsDb.Hsapiens.v86)
library(GenomicFeatures)
library(GO.db)

```

```
## some packages are installed using the package 'BioManager'
## the package 'GenomicFeatures' is needed to install 'EnsDb.Hsapiens.v86'
## install_github('https://github.com/Bioconductor/GenomicFeatures')
## the library 'devtools' is needed to install GenomicFeatures from GitHub repository
```

```
# the BF data file
```

```
panc.file <- read.csv("../rnf43.csv",header=TRUE,stringsAsFactors=F)
panc.gene <- panc.file$gene
panc.panc <- panc.file$panc.mean
panc.np <- panc.file$nonpanc.mean
quant <- quantile(panc.panc, probs = seq(0,1,1/20))
quant.np <- quantile(panc.np, probs=seq(0,1,1/20))
```

```
# top 10% BF genes (1718) in tumor cells
```

```
panc.top10 <- panc.gene[which(panc.panc > quant[19])]
length(panc.top10)
```

```
## [1] 1718
```

```
# genes (8458) of 0% to 50% BF factors in normal cells
```

```
np.bottom50 <- panc.gene[which(panc.np < quant.np[11])]
length(np.bottom50)
```

```
## [1] 8458
```

```
# the overlap genes between top 10% tumor and (0-50%) normal cells
```

```
genes=panc.gene[which(panc.top10 %in% np.bottom50)]
length(genes)
```

```
## [1] 51
```

```
GeneNameFilter(genes)
```

```
## class: GeneNameFilter
```

```
## condition: ==
```

```
## value: FZD5 WLS HNF1A RBM15 PPCS PORCN TPK1 ADRBK1 NANS PPARG SLC2A1 STK40 MPI ALG3 WNT3 TRIB1 STX4
```

```
# Gene names need be converted to Ensembl IDs
```

```
library(EnsDb.Hsapiens.v86)
```

```
ensembl.id <- genes(EnsDb.Hsapiens.v86,
```

```
  filter=list(GeneNameFilter(genes),GeneIdFilter("ENSG", "startsWith")),
```

```
  return.type="data.frame", columns=c("gene_id"))
```

```
write.table(ensembl.id, file="hs.rnf43.gene.set.csv", sep=",", row.names = F, col.names = F, quote = F)
```

```
ensembl.id
```

```
##           gene_id gene_name
## 1  ENSG00000100412      ACO2
## 2  ENSG00000214160      ALG3
## 3  ENSG00000086848      ALG9
## 4  ENSG00000161203     AP2M1
## 5  ENSG00000137135   ARHGEF39
## 6  ENSG00000067248     DHX29
## 7  ENSG00000133884     DPF2
## 8  ENSG00000179151     EDC3
## 9  ENSG00000156030   ELMSAN1
## 10 ENSG00000204930   FAM221B
```

## 11	ENSG00000079459	FDFT1
## 12	ENSG00000163251	FZD5
## 13	ENSG00000117308	GALE
## 14	ENSG00000147533	GOLGA7
## 15	ENSG00000135100	HNF1A
## 16	ENSG00000186625	KATNA1
## 17	ENSG00000142515	KLK3
## 18	ENSG00000053747	LAMA3
## 19	ENSG00000132763	MMACHC
## 20	ENSG00000115275	MOGS
## 21	ENSG00000178802	MPI
## 22	ENSG00000116984	MTR
## 23	ENSG00000124275	MTRR
## 24	ENSG00000095380	NANS
## 25	ENSG00000072864	NDE1
## 26	ENSG00000275911	NDE1
## 27	ENSG00000141458	NPC1
## 28	ENSG00000166228	PCBD1
## 29	ENSG00000013375	PGM3
## 30	ENSG00000102312	PORCN
## 31	ENSG00000132170	PPARG
## 32	ENSG00000138621	PPCDC
## 33	ENSG00000127125	PPCS
## 34	ENSG00000011485	PPP5C
## 35	ENSG00000126464	PRR12
## 36	ENSG00000117425	PTCH2
## 37	ENSG00000204764	RANBP17
## 38	ENSG00000162775	RBM15
## 39	ENSG00000080345	RIF1
## 40	ENSG00000100075	SLC25A1
## 41	ENSG00000117394	SLC2A1
## 42	ENSG00000196182	STK40
## 43	ENSG00000103496	STX4
## 44	ENSG00000029639	TFB1M
## 45	ENSG00000196511	TPK1
## 46	ENSG00000173334	TRIB1
## 47	ENSG00000119541	VPS4B
## 48	ENSG00000162923	WDR26
## 49	ENSG00000116729	WLS
## 50	ENSG00000108379	WNT3
## 51	ENSG00000277626	WNT3
## 52	ENSG00000277641	WNT3

ensembl.id\$gene_id

```
## [1] "ENSG00000100412" "ENSG00000214160" "ENSG00000086848"
## [4] "ENSG00000161203" "ENSG00000137135" "ENSG00000067248"
## [7] "ENSG00000133884" "ENSG00000179151" "ENSG00000156030"
## [10] "ENSG00000204930" "ENSG00000079459" "ENSG00000163251"
## [13] "ENSG00000117308" "ENSG00000147533" "ENSG00000135100"
## [16] "ENSG00000186625" "ENSG00000142515" "ENSG00000053747"
## [19] "ENSG00000132763" "ENSG00000115275" "ENSG00000178802"
## [22] "ENSG00000116984" "ENSG00000124275" "ENSG00000095380"
## [25] "ENSG00000072864" "ENSG00000275911" "ENSG00000141458"
## [28] "ENSG00000166228" "ENSG00000013375" "ENSG00000102312"
```

```

## [31] "ENSG000000132170" "ENSG000000138621" "ENSG000000127125"
## [34] "ENSG000000011485" "ENSG000000126464" "ENSG000000117425"
## [37] "ENSG000000204764" "ENSG000000162775" "ENSG000000080345"
## [40] "ENSG000000100075" "ENSG000000117394" "ENSG000000196182"
## [43] "ENSG000000103496" "ENSG000000029639" "ENSG000000196511"
## [46] "ENSG000000173334" "ENSG000000119541" "ENSG000000162923"
## [49] "ENSG000000116729" "ENSG000000108379" "ENSG000000277626"
## [52] "ENSG000000277641"

# Enrichment analysis
mart <- useDataset("hsapiens_gene_ensembl", mart=useMart("ensembl"))
all_ensembl_gene_id <- getBM(attributes = "ensembl_gene_id",
                             values = "*", mart = mart)
all <- factor(as.integer(all_ensembl_gene_id[,1] %in% ensembl.id$gene_id))
names(all) <- all_ensembl_gene_id[,1]

GOdata <- new("topGOdata", ontology="BP",
              allGenes = all, geneSel=function(p) p == 1,
              description = "P.not.NP", annot=annFUN.org, mapping="org.Hs.eg.db", ID="Ensembl")

##
## Building most specific GOs .....
## Loading required package: org.Hs.eg.db
##
## ( 12078 GO terms found. )
##
## Build GO DAG topology .....
## ( 16113 GO terms and 38254 relations. )
##
## Annotating nodes .....
## ( 20504 genes annotated to the GO terms. )
result.test <- runTest(GOdata, algorithm = "classic", statistic = "fisher")

##
## -- Classic Algorithm --
##
## the algorithm is scoring 1966 nontrivial nodes
## parameters:
## test statistic: fisher
resultKS.test <- runTest(GOdata, algorithm = "classic", statistic = "ks")

##
## -- Classic Algorithm --
##
## the algorithm is scoring 16113 nontrivial nodes
## parameters:
## test statistic: ks
## score order: increasing
resultKS.elim.test <- runTest(GOdata, algorithm = "elim", statistic = "ks")

```

```

##
##      -- Elim Algorithm --
##
##      the algorithm is scoring 16113 nontrivial nodes
##      parameters:
##          test statistic: ks
##          cutOff: 0.01
##          score order: increasing
##
## Level 20:  1 nodes to be scored    (0 eliminated genes)
##
## Level 19:  5 nodes to be scored    (0 eliminated genes)
##
## Level 18: 22 nodes to be scored    (0 eliminated genes)
##
## Level 17: 52 nodes to be scored    (13 eliminated genes)
##
## Level 16: 120 nodes to be scored   (31 eliminated genes)
##
## Level 15: 255 nodes to be scored   (195 eliminated genes)
##
## Level 14: 499 nodes to be scored   (354 eliminated genes)
##
## Level 13: 919 nodes to be scored   (1156 eliminated genes)
##
## Level 12: 1374 nodes to be scored  (3460 eliminated genes)
##
## Level 11: 1786 nodes to be scored  (6096 eliminated genes)
##
## Level 10: 2092 nodes to be scored  (7714 eliminated genes)
##
## Level 9:  2191 nodes to be scored  (9857 eliminated genes)
##
## Level 8:  2088 nodes to be scored  (12111 eliminated genes)
##
## Level 7:  1942 nodes to be scored  (13740 eliminated genes)
##
## Level 6:  1453 nodes to be scored  (15133 eliminated genes)
##
## Level 5:   779 nodes to be scored  (16447 eliminated genes)
##
## Level 4:   375 nodes to be scored  (17342 eliminated genes)
##
## Level 3:   135 nodes to be scored  (17551 eliminated genes)

```

```
##
## Level 2: 24 nodes to be scored (17895 eliminated genes)

##
## Level 1: 1 nodes to be scored (17905 eliminated genes)
allRes <- GenTable(GOdata, classicFisher = result.test, classicKS = resultKS.test, elimKS = resultKS.elim)
allRes
```

##	GO.ID	Term	Annotated
## 1	GO:0045944	positive regulation of transcription by ...	1273
## 2	GO:0000122	negative regulation of transcription by ...	904
## 3	GO:0008283	cell proliferation	2322
## 4	GO:0043687	post-translational protein modification	401
## 5	GO:0051301	cell division	625
## 6	GO:0045893	positive regulation of transcription, DN...	1621
## 7	GO:0008584	male gonad development	137
## 8	GO:1990830	cellular response to leukemia inhibitory...	99
## 9	GO:0035556	intracellular signal transduction	3075
## 10	GO:0046777	protein autophosphorylation	245
## 11	GO:0007411	axon guidance	267
## 12	GO:0001701	in utero embryonic development	338
## 13	GO:0006406	mRNA export from nucleus	123
## 14	GO:0043547	positive regulation of GTPase activity	456
## 15	GO:0018105	peptidyl-serine phosphorylation	294
## 16	GO:0007601	visual perception	226
## 17	GO:0045471	response to ethanol	132
## 18	GO:0010628	positive regulation of gene expression	2104
## 19	GO:0043627	response to estrogen	74
## 20	GO:0035735	intraciliary transport involved in cilium...	40

##	Significant	Expected Rank in classicFisher	classicFisher	classicKS
## 1	5	3.04	1101	0.186 5.9e-23
## 2	2	2.16	1810	0.643 2.7e-13
## 3	9	5.55	795	0.097 3.2e-18
## 4	0	0.96	1966	1.000 1.2e-11
## 5	4	1.49	631	0.062 7.3e-18
## 6	6	3.87	1105	0.188 < 1e-30
## 7	0	0.33	1967	1.000 1.5e-10
## 8	1	0.24	1176	0.211 3.6e-09
## 9	6	7.35	1888	0.764 < 1e-30
## 10	0	0.59	1968	1.000 8.0e-09
## 11	3	0.64	386	0.026 8.9e-10
## 12	1	0.81	1745	0.558 6.3e-11
## 13	0	0.29	1969	1.000 7.9e-08
## 14	0	1.09	1970	1.000 8.6e-08
## 15	0	0.70	1971	1.000 4.7e-11
## 16	0	0.54	1972	1.000 1.3e-07
## 17	0	0.32	1973	1.000 1.3e-07
## 18	10	5.03	381	0.025 < 1e-30
## 19	1	0.18	1029	0.163 2.3e-07
## 20	0	0.10	1974	1.000 2.5e-07

##	elimKS
## 1	6.6e-22
## 2	2.7e-13

```
## 3  1.1e-12
## 4  1.2e-11
## 5  3.2e-10
## 6  9.1e-10
## 7  1.4e-09
## 8  3.6e-09
## 9  5.0e-09
## 10 8.0e-09
## 11 1.6e-08
## 12 5.4e-08
## 13 7.9e-08
## 14 8.6e-08
## 15 1.1e-07
## 16 1.3e-07
## 17 1.3e-07
## 18 1.4e-07
## 19 2.3e-07
## 20 2.5e-07
```