NetBAS for tumor genes HBG 11/4/2018

This script perform GO enrichment using NetBAS

for 51 PDAC cell high BF genes

```
panc.file <- read.csv("../rnf43.csv",header=TRUE,stringsAsFactors=F)</pre>
panc.gene <- panc.file$gene</pre>
panc.panc <- panc.file$panc.mean</pre>
panc.np <- panc.file$nonpanc.mean</pre>
quant <- quantile(panc.panc, probs = seq(0,1,1/20))
quant.np <- quantile(panc.np, probs=seq(0,1,1/20))</pre>
# 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])]</pre>
length(np.bottom50)
## [1] 8458
# the overlap genes between top 10% tumor and (0-50%) normal cells
gene.list <- panc.gene[which(panc.top10 %in% np.bottom50)]</pre>
length(gene.list)
## [1] 51
gene.list
## [1] "FZD5"
                    "WLS"
                               "HNF1A"
                                           "RBM15"
                                                      "PPCS"
                                                                  "PORCN"
                    "ADRBK1"
                               "NANS"
                                                      "SLC2A1"
                                                                  "STK40"
## [7] "TPK1"
                                           "PPARG"
## [13] "MPI"
                    "ALG3"
                               "WNT3"
                                           "TRIB1"
                                                      "STX4"
                                                                  "SLC25A1"
## [19] "WDR26"
                    "VPS4B"
                               "MMACHC"
                                           "PPP5C"
                                                      "KATNA1"
                                                                  "ARHGEF39"
## [25] "GALE"
                               "PCBD1"
                                           "PGM3"
                                                      "PRR12"
                                                                  "AP2M1"
                    "MOGS"
## [31] "PTCH2"
                    "RIF1"
                               "TFB1M"
                                           "DHX29"
                                                      "MTR"
                                                                  "NPC1"
## [37] "PPP2R4"
                               "PPCDC"
                                                                  "EDC3"
                    "ALG9"
                                           "GOLGA7"
                                                      "KLK3"
## [43] "LAMA3"
                                           "RANBP17" "NDE1"
                    "ACO2"
                               "MTRR"
                                                                  "FAM221B"
## [49] "FDFT1"
                    "DPF2"
                               "ELMSAN1"
## read the original network
network <- read.csv("../Data/human.pin.csv", header=T, stringsAsFactors=F)</pre>
geneA <- network$geneA
geneB <- network$geneB
GOcategory.file <- read.csv("../Data/human.bp.term.csv",header=TRUE, stringsAsFactors=F)
bp.go.cat <- GOcategory.file$GO.term</pre>
```

```
bp.dim <- length(bp.go.cat)</pre>
GOterm.file <- read.csv("../Data/human.bp.gene.term.csv", header=T, stringsAsFactors=F)
bp.GO.gene <- GOterm.file$gene #it should be changed to System for yeast pin
bp.GO.term <- GOterm.file$GO.term</pre>
vec <- numeric(length=bp.dim)</pre>
for (i in 1:length(gene.list)) {
    orf <- as.character(gene.list[i])</pre>
    intA <- geneB[which(geneA %in% orf)]</pre>
    for (j in 1:length(intA)) {
        bpA <- bp.GO.term[which(bp.GO.gene %in% intA[j])]</pre>
        for (k in 1:length(bpA)) {
            na <- which(bp.go.cat %in% bpA[k])</pre>
            vec[na] \leftarrow vec[na] + 1
        }
    }
    intB <- geneA[which(geneB %in% orf)]</pre>
    for (s in 1:length(intB)) {
        bpB <- bp.GO.term[which(bp.GO.gene %in% intB[s])]</pre>
        for (t in 1:length(bpB)) {
            nb <- which(bp.go.cat %in% bpB[t])</pre>
            vec[nb] \leftarrow vec[nb] + 1
        }
    }
}
write.table(vec, file="hs.rnf43.list.bp.txt", col.names=F, row.names=F, quote=F)
# Now the ms02star permutations
for (p in 1:100) {
permutation.file <- paste("../ms02star/human/", "ms02.", p, ".csv", sep="")</pre>
permutation <- read.csv(permutation.file, header=T, stringsAsFactors = F)</pre>
geneA <- permutation$id1</pre>
geneB <- permutation$id2</pre>
vecp <- numeric(length = bp.dim)</pre>
for (i in 1:length(gene.list)) {
    orf <- as.character(gene.list[i])</pre>
    intA <- geneB[which(geneA %in% orf)]</pre>
    for (j in 1:length(intA)) {
        bpA <- bp.GO.term[which(bp.GO.gene %in% intA[j])]</pre>
        for (k in 1:length(bpA)) {
            na <- which(bp.go.cat %in% bpA[k])</pre>
            vecp[na] \leftarrow vecp[na] + 1
        }
    }
    intB <- geneA[which(geneB %in% orf)]</pre>
    for (s in 1:length(intB)) {
        bpB <- bp.GO.term[which(bp.GO.gene %in% intB[s])]</pre>
        for (t in 1:length(bpB)) {
```

```
nb <- which(bp.go.cat %in% bpB[t])</pre>
           vecp[nb] \leftarrow vecp[nb] + 1
        }
    }
}
output <- paste("ms02.human", "/", "rnf43.list", "/", "ms02.", p, ".bp.matrix.csv", sep="")
write.table(vecp, file = output, col.names=F, row.names=F, quote=F)
library("microbenchmark")
library("matrixStats")
conn.dim <- 1
hspin <- matrix(as.numeric(unlist(read.table("hs.rnf43.list.bp.txt", header=F, sep=","))), nrow=bp.dim,
obs <- c(hspin)
perm <- c()
for (i in 1:100) {
    name <- paste("ms02.human", "/", "rnf43.list", "/", "ms02.", i, ".bp.matrix.csv", sep="")</pre>
    mat <- matrix(as.numeric(unlist(read.table(name, header=F, sep=","))), nrow=bp.dim, ncol=conn.dim)</pre>
    perm <- rbind(perm, c(mat))</pre>
}
mean <- colMeans(perm)</pre>
std <- colSds(perm)</pre>
zscore <- round((obs - mean)/std, 3)</pre>
z <- matrix(zscore, nrow=bp.dim, ncol=conn.dim)</pre>
write.table(z, file="hs.rnf43.list.bp.z.csv", sep=",", row.names=F, col.names=F, quote=F)
library('gplots')
## Attaching package: 'gplots'
## The following object is masked from 'package:stats':
##
##
       lowess
library('GO.db')
## Loading required package: AnnotationDbi
## Loading required package: stats4
## 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: Biobase
## Welcome to Bioconductor
##
##
       Vignettes contain introductory material; view with
##
       'browseVignettes()'. To cite Bioconductor, see
##
       'citation("Biobase")', and for packages 'citation("pkgname")'.
##
## Attaching package: 'Biobase'
## The following objects are masked from 'package:matrixStats':
##
##
       anyMissing, rowMedians
## Loading required package: IRanges
## Loading required package: S4Vectors
##
## Attaching package: 'S4Vectors'
## The following object is masked from 'package:gplots':
##
##
       space
## The following object is masked from 'package:base':
##
##
       expand.grid
##
order <- order(z)
bp.go.cat <- bp.go.cat[order]</pre>
z <- z[order]
z \leftarrow t(z)
colnames(z) <- bp.go.cat</pre>
enriched.list <- bp.go.cat[which(z >= 3)]
```

```
enriched <- c("GO.ID", "GO.Term", "Z-score")</pre>
for (i in 1:length(enriched.list)) {
  term <- Term(GOID(as.character(enriched.list[i])))</pre>
  enriched <- rbind(enriched, c(as.character(enriched.list[i]),term,</pre>
                                   z[which(bp.go.cat %in% enriched.list[i])]))
}
print(enriched)
##
## enriched "GO.ID"
##
             "GO:0044331"
##
             "GO:0000186"
             "GO:0034614"
##
             "GO:0035116"
##
##
             "GD:0060348"
##
             "GO:0007059"
##
             "GD:0030301"
             "GO:0017156"
##
             "GD:0050807"
##
             "GD:0007420"
##
##
             "GO:0071277"
##
             "GD:0060349"
##
             "GO:0045892"
             "GO:0042472"
##
##
             "GD:0006338"
##
             "GD:0007623"
##
             "GO:0030522"
             "GO:0001658"
##
##
             "GD:0006898"
##
             "GD:0006986"
             "GO:0045599"
##
##
             "GD:0050873"
##
             "GD:0034446"
##
             "GD:0030324"
             "GO:0042026"
##
             "GO:0007155"
##
##
             "GO:0010389"
##
             "GO:0043401"
             "GO:0042730"
##
             "GD:0060384"
##
             "GO:0051591"
##
```

##

##

##

##

##

##

##

##

##

##

"GD:0006418"
"GD:0007052"

"GD:0050767"

"GD:0030521"

"GD:0034605"
"GD:0006953"

"GO:0007018"

"GO:0048146"

"GD:0002576"

"GD:0022008"

"GO:1901673"

"GO:0001934"

```
"GD:0030514"
##
             "GD:0007223"
##
             "GO:0006812"
##
##
             "GD:0001843"
##
             "GD:0030879"
##
             "GO:0043547"
##
             "GO:0016192"
             "GD:0008203"
##
##
             "GD:0003341"
             "GO:0042632"
##
##
             "GO:0051085"
             "GD:0048706"
##
             "GD:0001523"
##
             "GO:1900034"
##
##
             "GD:0046718"
##
             "GO:0033627"
##
             "GD:0006508"
             "GO:0031122"
##
             "GD:0007224"
##
             "GO:0045669"
##
##
             "GD:0034620"
##
             "GD:0006997"
             "GD:0030198"
##
##
             "GD:0034613"
             "GD:0016079"
##
##
             "GO:0019216"
##
             "GO:0030574"
##
             "GD:0045600"
             "GO:0030216"
##
             "GD:0090263"
##
             "GO:0070268"
##
##
             "GD:0006629"
             "GD:0006303"
##
             "GO:0036258"
##
             "GO:0016042"
##
             "GD:0071300"
##
##
             "GO:0015721"
##
             "GD:0097711"
##
             "GO:0008202"
             "GO:0061024"
##
##
             "GO:0022617"
             "GO:0090103"
##
##
             "GD:0060071"
             "GO:0045893"
##
##
             "GD:0016055"
             "GD:0090090"
##
             "GD:0039702"
##
##
             "GO:0048843"
             "GD:0009235"
##
             "GO:1904886"
##
             "GD:0030182"
##
             "GO:0045165"
##
             "GO:0035567"
##
             "GD:0060070"
##
```

```
##
            GO:0044331
## enriched "GO.Term"
            "cell-cell adhesion mediated by cadherin"
##
            "activation of MAPKK activity"
##
##
            "cellular response to reactive oxygen species"
##
            "embryonic hindlimb morphogenesis"
            "bone development"
##
##
            "chromosome segregation"
##
            "cholesterol transport"
##
            "calcium ion regulated exocytosis"
##
            "regulation of synapse organization"
##
            "brain development"
##
            "cellular response to calcium ion"
##
            "bone morphogenesis"
##
            "negative regulation of transcription, DNA-templated"
##
            "inner ear morphogenesis"
##
            "chromatin remodeling"
##
            "circadian rhythm"
##
            "intracellular receptor signaling pathway"
##
            "branching involved in ureteric bud morphogenesis"
##
            "receptor-mediated endocytosis"
##
            "response to unfolded protein"
            "negative regulation of fat cell differentiation"
##
            "brown fat cell differentiation"
##
##
            "substrate adhesion-dependent cell spreading"
##
            "lung development"
##
            "protein refolding"
##
            "cell adhesion"
##
            "regulation of G2/M transition of mitotic cell cycle"
##
            "steroid hormone mediated signaling pathway"
##
            "fibrinolysis"
##
            "innervation"
            "response to cAMP"
##
##
            "tRNA aminoacylation for protein translation"
##
            "mitotic spindle organization"
##
            "regulation of neurogenesis"
##
            "androgen receptor signaling pathway"
##
            "cellular response to heat"
##
            "acute-phase response"
##
            "microtubule-based movement"
##
            "positive regulation of fibroblast proliferation"
##
            "platelet degranulation"
##
            "neurogenesis"
##
            "regulation of mitotic spindle assembly"
##
            "positive regulation of protein phosphorylation"
##
            "negative regulation of BMP signaling pathway"
##
            "Wnt signaling pathway, calcium modulating pathway"
##
            "cation transport"
##
            "neural tube closure"
##
            "mammary gland development"
##
            "positive regulation of GTPase activity"
##
            "vesicle-mediated transport"
##
            "cholesterol metabolic process"
            "cilium movement"
##
```

```
"cholesterol homeostasis"
##
##
            "chaperone cofactor-dependent protein refolding"
            "embryonic skeletal system development"
##
            "retinoid metabolic process"
##
##
            "regulation of cellular response to heat"
##
            "viral entry into host cell"
            "cell adhesion mediated by integrin"
##
##
            "proteolysis"
##
            "cytoplasmic microtubule organization"
##
            "smoothened signaling pathway"
##
            "positive regulation of osteoblast differentiation"
            "cellular response to unfolded protein"
##
##
            "nucleus organization"
##
            "extracellular matrix organization"
##
            "cellular protein localization"
##
            "synaptic vesicle exocytosis"
##
            "regulation of lipid metabolic process"
##
            "collagen catabolic process"
##
            "positive regulation of fat cell differentiation"
##
            "keratinocyte differentiation"
##
            "positive regulation of canonical Wnt signaling pathway"
##
            "cornification"
##
            "lipid metabolic process"
            "double-strand break repair via nonhomologous end joining"
##
##
            "multivesicular body assembly"
##
            "lipid catabolic process"
##
            "cellular response to retinoic acid"
##
            "bile acid and bile salt transport"
##
            "ciliary basal body-plasma membrane docking"
##
            "steroid metabolic process"
##
            "membrane organization"
##
            "extracellular matrix disassembly"
##
            "cochlea morphogenesis"
##
            "Wnt signaling pathway, planar cell polarity pathway"
##
            "positive regulation of transcription, DNA-templated"
##
            "Wnt signaling pathway"
##
            "negative regulation of canonical Wnt signaling pathway"
##
            "viral budding via host ESCRT complex"
            "negative regulation of axon extension involved in axon guidance"
##
##
            "cobalamin metabolic process"
##
            "beta-catenin destruction complex disassembly"
##
            "neuron differentiation"
            "cell fate commitment"
##
            "non-canonical Wnt signaling pathway"
##
            "canonical Wnt signaling pathway"
##
##
## enriched "Z-score"
            "3.01"
##
##
            "3.014"
            "3.029"
##
##
            "3.068"
##
            "3.069"
##
            "3.078"
            "3.083"
##
```

```
"3.103"
##
             "3.117"
##
             "3.122"
##
##
             "3.128"
             "3.129"
##
##
             "3.131"
##
             "3.16"
             "3.182"
##
##
             "3.214"
##
             "3.285"
##
             "3.303"
             "3.315"
##
##
             "3.37"
             "3.385"
##
##
             "3.388"
             "3.396"
##
##
             "3.42"
             "3.451"
##
             "3.458"
##
             "3.482"
##
             "3.484"
##
##
             "3.494"
##
             "3.499"
             "3.514"
##
             "3.553"
##
##
             "3.565"
##
             "3.575"
##
             "3.598"
             "3.604"
##
##
             "3.605"
             "3.613"
##
##
             "3.619"
##
             "3.638"
             "3.669"
##
             "3.689"
##
             "3.7"
##
##
             "3.733"
##
             "3.739"
             "3.747"
##
##
             "3.75"
##
             "3.793"
             "3.797"
##
##
             "3.803"
##
             "3.823"
##
             "3.843"
             "3.843"
##
##
             "3.995"
##
             "4.011"
##
             "4.038"
##
             "4.059"
             "4.109"
##
             "4.187"
##
##
             "4.199"
##
             "4.218"
```

```
"4.234"
##
##
             "4.286"
##
             "4.292"
             "4.297"
##
##
             "4.333"
             "4.536"
##
##
             "4.55"
             "4.573"
##
##
             "4.585"
##
             "4.643"
##
             "4.703"
             "4.769"
##
##
             "4.836"
             "4.878"
##
##
             "4.953"
##
             "4.984"
##
             "5.043"
##
             "5.086"
##
             "5.111"
##
             "5.467"
             "5.555"
##
##
             "6.017"
             "6.113"
##
##
             "6.224"
             "6.25"
##
##
             "6.703"
             "6.911"
##
##
             "7.163"
##
             "7.918"
##
             "8.259"
             "9.218"
##
##
             "9.266"
##
             "10.133"
##
             "10.469"
##
             "12.956"
##
             "17.879"
write.table(enriched, file="hs.rnf43.list.bp.enriched.csv", row.names=F, col.names=F, quote=F, sep="\t"
###No suppressed terms have been found
\#sup.list \leftarrow bp.go.cat[which(z \leftarrow -3)]
#sup <- c("GO.ID", "GO.Term", "Z-score")
#for (i in 1:length(sup.list)) {
# term <- Term(GOID(as.character(sup.list[i])))</pre>
# sup <- rbind(sup, c(as.character(sup.list[i]), term,</pre>
#
                 z[which(bp.go.cat %in% sup.list[i])]))
#}
#print(sup)
\#write.table(sup, file="human.pdcd1.bp.suppressed.csv", row.names=F, col.names=F, quote=F, sep="\t")
###
#Note that there may be "inf" Z-scores owing to lack of sampling (i.e., zero in standard deviations)
```

#We can also extract the GO-terms for the gene for comparison