## BioinformaticsHW\_IsabelMetzger

ISABEL METZGER BIOINFORMATICS FALL 2017 LOADING EDGER

```
library(limma)
## Warning: package 'limma' was built under R version 3.4.2
library(edgeR)
library(car)
Read in count table and experimental design:
ovarian df <- read.delim("GSE52695 Ovarian RefSeq RPKM Values.txt",
                        row.names=1. header=T)
head(ovarian_df)
##
           Library_1 Library_10 Library_11 Library_12 Library_2 Library_3
## A1BG
           0.2621930 0.9163250 0.90431500 1.03555000 0.30342000 0.3397060
## A1BG-AS1 0.9704190 0.8373180 1.38764000 1.51185000 0.66485800 1.0530700
## A1CF 0.0103266 0.0212174 0.03549150 0.00425699 0.01229490 0.0111711
## A2LD1
           0.5142540 1.5253800 0.92915700 1.10681000 0.41714800 0.4988250
## A2M
           0.0000000 \quad 0.0283904 \quad 0.00647592 \quad 0.00854420 \quad 0.00822571 \quad 0.0000000
## A2ML1 0.0000000 0.0127313 0.01161620 0.00766310 0.00737746 0.0000000
##
           Library_4 Library_5 Library_6 Library_7 Library_8 Library_9
## A1BG 0.44528100 0.51423700 0.30590300 0.6488740 0.65693400 0.81654700
## A1BG-AS1 0.68563300 0.64529800 0.99422200 1.0894200 0.99634600 0.86571700
## A1CF 0.00902164 0.01051260 0.00925534 0.0214581 0.01211950 0.02153980
           0.65327200 0.40458300 0.50209300 0.5161380 0.42464100 1.39639000
## A2LD1
           0.37421800 0.00527496 0.00928820 0.0000000 0.00608129 0.00000000
## A2M
## A2ML1
           the dataframe contains 22446 rows (genes) and a total of 12 cols (12 libraries)
dim(ovarian_df)
## [1] 22446
               12
Reordering column names
reorder.colnames <- paste("Library", 1:12, sep="_")
reorder.colnames
  [1] "Library 1" "Library 2" "Library 3" "Library 4"
                                                          "Library 5"
  [6] "Library_6" "Library_7" "Library_8" "Library_9" "Library_10"
## [11] "Library_11" "Library_12"
ovarian_reordered_df <- ovarian_df[,reorder.colnames]</pre>
head(ovarian_reordered_df)
           Library_1 Library_2 Library_3 Library_4 Library_5 Library_6
```

0.2621930 0.30342000 0.3397060 0.44528100 0.51423700 0.30590300

## A1BG-AS1 0.9704190 0.66485800 1.0530700 0.68563300 0.64529800 0.99422200

## A1BG

```
0.0103266 0.01229490 0.0111711 0.00902164 0.01051260 0.00925534
## A1CF
           0.5142540 0.41714800 0.4988250 0.65327200 0.40458300 0.50209300
## A2LD1
## A2M
           0.0000000 0.00822571 0.0000000 0.37421800 0.00527496 0.00928820
           ## A2ML1
           Library_7 Library_8 Library_9 Library_10 Library_11 Library_12
           0.6488740 0.65693400 0.81654700 0.9163250 0.90431500 1.03555000
## A1BG
## A1BG-AS1 1.0894200 0.99634600 0.86571700 0.8373180 1.38764000 1.51185000
           0.0214581 0.01211950 0.02153980 0.0212174 0.03549150 0.00425699
## A1CF
## A2LD1
           0.5161380 0.42464100 1.39639000 1.5253800 0.92915700 1.10681000
           0.0000000 0.00608129 0.00000000 0.0283904 0.00647592 0.00854420
## A2M
## A2ML1
           0.0000000 0.00000000 0.00646238 0.0127313 0.01161620 0.00766310
Creating new colnames
newcolnames <- paste("GSM12742", 53:64, sep="") #2 or more vectors pasted element for element.
newcolnames
  [1] "GSM1274253" "GSM1274254" "GSM1274255" "GSM1274256" "GSM1274257"
   [6] "GSM1274258" "GSM1274259" "GSM1274260" "GSM1274261" "GSM1274262"
## [11] "GSM1274263" "GSM1274264"
library(data.table)
## Warning: package 'data.table' was built under R version 3.4.2
new_ovarian_df <- setnames(ovarian_reordered_df,</pre>
                         old=reorder.colnames, new=newcolnames)
head(new_ovarian_df)
           GSM1274253 GSM1274254 GSM1274255 GSM1274256 GSM1274257 GSM1274258
##
## A1BG
            0.2621930 0.30342000 0.3397060 0.44528100 0.51423700 0.30590300
## A1BG-AS1 0.9704190 0.66485800 1.0530700 0.68563300 0.64529800 0.99422200
            0.0103266 0.01229490 0.0111711 0.00902164 0.01051260 0.00925534
## A1CF
## A2LD1
            0.5142540 0.41714800 0.4988250 0.65327200 0.40458300 0.50209300
            0.0000000 0.00822571 0.0000000 0.37421800 0.00527496 0.00928820
## A2M
## A2ML1
           ##
           GSM1274259 GSM1274260 GSM1274261 GSM1274262 GSM1274263 GSM1274264
           0.6488740 0.65693400 0.81654700 0.9163250 0.90431500 1.03555000
## A1BG
## A1BG-AS1 1.0894200 0.99634600 0.86571700 0.8373180 1.38764000 1.51185000
## A1CF
           0.0214581 0.01211950 0.02153980 0.0212174 0.03549150 0.00425699
## A2LD1
            0.5161380 0.42464100 1.39639000 1.5253800 0.92915700 1.10681000
## A2M
            0.0000000 0.00608129 0.00000000 0.0283904 0.00647592 0.00854420
            0.0000000 0.00000000 0.00646238 0.0127313 0.01161620 0.00766310
Genes are now row names and what is left is the counts/ expression levels. See below for the renamed
columns.
length(reorder.colnames)
## [1] 12
length(newcolnames)
## [1] 12
treatment <- c("Normoxia_6hr (total RNA)_1", "Normoxia_6hr (total RNA)_2",
              "Hypoxia_6hr (total RNA)_3","Hypoxia_6hr (total RNA)_4","Normoxia_48hr (total RNA)_5","
```

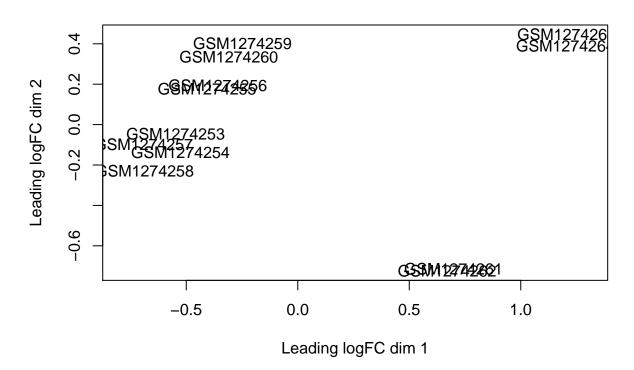
```
"Hypoxia_48hr (total RNA)_8","Normoxia_6days (total RNA)_9","Normoxia_6days (total RNA)_
length(treatment)
## [1] 12
keydf <- data.frame(reorder.colnames, newcolnames, treatment)</pre>
keydf$treatmentsimplified <- c("Normoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", "Hypoxia", "Normoxia", "Hypoxia", 
keydf
##
           reorder.colnames newcolnames
                                                                                                        treatment
                        Library_1 GSM1274253
## 1
                                                                       Normoxia_6hr (total RNA)_1
## 2
                       Library_2 GSM1274254
                                                                       Normoxia_6hr (total RNA)_2
## 3
                       Library_3 GSM1274255
                                                                      Hypoxia_6hr (total RNA)_3
## 4
                       Library_4 GSM1274256
                                                                      Hypoxia_6hr (total RNA)_4
                       Library_5 GSM1274257
## 5
                                                                     Normoxia_48hr (total RNA)_5
## 6
                       Library_6 GSM1274258
                                                                   Normoxia_48hr (total RNA)_6
## 7
                       Library 7 GSM1274259
                                                                   Hypoxia 48hr (total RNA) 7
                       Library_8 GSM1274260
## 8
                                                                       Hypoxia_48hr (total RNA)_8
                       Library_9 GSM1274261 Normoxia_6days (total RNA)_9
## 9
## 10
                      Library_10 GSM1274262 Normoxia_6days (total RNA)_10
## 11
                      Library_11 GSM1274263 Hypoxia_6days (total RNA)_11
                      Library_12 GSM1274264 Hypoxia_6days (total RNA)_12
## 12
##
           treatmentsimplified
## 1
                                Normoxia
## 2
                                Normoxia
## 3
                                 Hypoxia
## 4
                                 Hypoxia
                                Normoxia
## 5
                                Normoxia
## 6
## 7
                                 Hypoxia
## 8
                                Hypoxia
## 9
                                Normoxia
## 10
                                Normoxia
## 11
                                 Hypoxia
                                 Hypoxia
group <- keydf$treatmentsimplified</pre>
group
## [1] "Normoxia" "Normoxia" "Hypoxia" "Hypoxia" "Normoxia" "Normoxia"
    [7] "Hypoxia" "Hypoxia" "Normoxia" "Normoxia" "Hypoxia" "Hypoxia"
head(new_ovarian_df)
                      GSM1274253 GSM1274254 GSM1274255 GSM1274256 GSM1274257 GSM1274258
                       0.2621930\ 0.30342000\ 0.3397060\ 0.44528100\ 0.51423700\ 0.30590300
## A1BG
## A1BG-AS1 0.9704190 0.66485800 1.0530700 0.68563300 0.64529800 0.99422200
                        0.0103266 0.01229490 0.0111711 0.00902164 0.01051260 0.00925534
## A1CF
## A2LD1
                        0.5142540\ 0.41714800\ 0.4988250\ 0.65327200\ 0.40458300\ 0.50209300
                        0.0000000 0.00822571 0.0000000 0.37421800 0.00527496 0.00928820
## A2M
## A2ML1
                        GSM1274259 GSM1274260 GSM1274261 GSM1274262 GSM1274263 GSM1274264
##
## A1BG
                0.6488740 0.65693400 0.81654700 0.9163250 0.90431500 1.03555000
## A1BG-AS1 1.0894200 0.99634600 0.86571700 0.8373180 1.38764000 1.51185000
```

```
## A1CF
             0.0214581 0.01211950 0.02153980 0.0212174 0.03549150 0.00425699
## A2LD1
             0.5161380 0.42464100 1.39639000
                                                1.5253800 0.92915700 1.10681000
             0.0000000 0.00608129 0.00000000
                                                0.0283904 0.00647592 0.00854420
## A2M
## A2ML1
             0.0000000 0.00000000 0.00646238 0.0127313 0.01161620 0.00766310
#Create DGEList object with groups for the treatments N and H
cds <- DGEList(new_ovarian_df, group = group )</pre>
dim(cds)
## [1] 22446
#Filter out genes with low counts, keeping those rows where the count
#per million (cpm) is at least 1 in at least three samples:
keep \leftarrow rowSums(cpm(cds)>1) >=3
cds <- cds[keep,]</pre>
dim(cds)
                         # How many genes are left?
## [1] 12162
                12
cds <- calcNormFactors(cds) #normalize number of reads per sample</pre>
?plotMDS
```

Usage ## Default S3 method: plotMDS(x, top = 500, labels = NULL, pch = NULL, cex = 1, dim.plot = c(1,2), ndim = max(dim.plot), gene.selection = "pairwise", xlab = NULL, ylab = NULL, plot = TRUE, ...)
## S3 method for class 'MDS' plotMDS(x, labels = NULL, pch = NULL, cex = 1, dim.plot = NULL, xlab = NULL, ylab = NULL, ...)

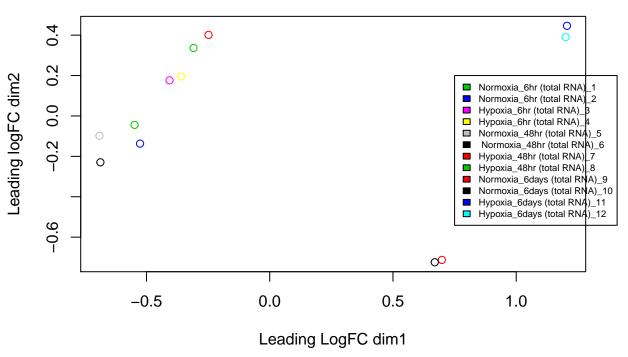
```
#plot the data (MDS is like PCA)
mds <- plotMDS(cds, main = "MDS Plot for Count Data", labels = colnames(cds$counts))</pre>
```

## **MDS Plot for Count Data**



```
plot(mds, main="MDS Plot for Count Data, clear view", xlab="Leading LogFC dim1", ylab="Leading logFC dim1", par(xpd=TRUE)
legend(0.75,0.2,fill=keydf$treatment,legend=keydf$treatment,cex=0.6)
```

## MDS Plot for Count Data, clear view



```
# title("Tx Status")
# library(car)
cds
```

```
## An object of class "DGEList"
   $counts
##
            GSM1274253 GSM1274254 GSM1274255 GSM1274256 GSM1274257 GSM1274258
##
## A1BG
              0.262193
                          0.303420
                                      0.339706
                                                 0.445281
                                                             0.514237
                                                                        0.305903
## A1BG-AS1
              0.970419
                          0.664858
                                      1.053070
                                                 0.685633
                                                             0.645298
                                                                        0.994222
## A2LD1
              0.514254
                          0.417148
                                     0.498825
                                                 0.653272
                                                             0.404583
                                                                        0.502093
## AAAS
              8.223290
                          7.361950
                                     9.427670
                                                 7.773020
                                                            10.328000
                                                                       10.390200
## AACS
              5.169470
                          4.668560
                                      5.389730
                                                 4.583680
                                                             4.632230
                                                                        4.953550
##
            GSM1274259 GSM1274260 GSM1274261 GSM1274262 GSM1274263 GSM1274264
## A1BG
              0.648874
                          0.656934
                                      0.816547
                                                 0.916325
                                                             0.904315
                                                                          1.03555
## A1BG-AS1
              1.089420
                          0.996346
                                     0.865717
                                                 0.837318
                                                             1.387640
                                                                          1.51185
## A2LD1
              0.516138
                          0.424641
                                      1.396390
                                                 1.525380
                                                             0.929157
                                                                          1.10681
## AAAS
              10.274000
                          9.328310
                                                 8.270740
                                                                          6.94087
                                      7.662060
                                                             6.487310
## AACS
              5.533260
                          5.850510
                                      3.634350
                                                 3.553840
                                                             5.642410
                                                                          5.87783
##
  12157 more rows ...
##
##
   $samples
##
                  group lib.size norm.factors
## GSM1274253 Normoxia 772461.6
                                     0.8880870
## GSM1274254 Normoxia 603231.3
                                     1.0442287
## GSM1274255 Hypoxia 618097.6
                                     1.1320669
```

```
## GSM1274256 Hypoxia 615240.6
                                    1.0903608
## GSM1274257 Normoxia 788218.0
                                    0.8841033
## 7 more rows ...
                #estimate variance, compute diff. exp
cds <- estimateCommonDisp( cds )</pre>
cds <- estimateTagwiseDisp( cds , prior.df = 10 )</pre>
                                                      #shrinks variance toward common disp.
de.tgw <- exactTest( cds , pair = c( "Normoxia" , "Hypoxia" ) ) #this is the DE test</pre>
options( digits = 3 ) # print only 3 digits
topTags( de.tgw , n = 20 , sort.by = "p.value" )
                                                        # print the top 20 DE genes, by p-value
## Comparison of groups: Hypoxia-Normoxia
##
           logFC logCPM
                           PValue
## PFKFB4
            3.37
                   4.19 3.73e-23 4.54e-19
                   6.88 8.05e-17 4.90e-13
## BNIP3
            2.62
## SLC2A1
            2.28
                   4.30 7.96e-16 3.23e-12
## HK2
                   5.06 1.08e-13 3.29e-10
            1.83
## ADM
            2.50
                   4.17 1.85e-13 4.49e-10
## ANGPTL4
            3.11
                   3.75 2.43e-13 4.93e-10
## SLC16A3
            1.90
                   4.86 4.07e-13 7.07e-10
## ANKZF1
            1.77
                   4.47 2.99e-12 4.54e-09
## FAM162A 1.85
                   6.83 4.35e-12 5.88e-09
                   4.42 5.46e-12 6.63e-09
## AK4
            1.77
## PFKP
            1.87
                   5.17 1.50e-11 1.66e-08
## PGK1
            1.53
                   7.94 1.67e-11 1.69e-08
                   3.99 3.12e-11 2.91e-08
## INSIG2
            2.44
                   5.13 3.39e-11 2.95e-08
## P4HA2
            1.80
## EGLN1
            1.60
                   4.58 6.86e-11 5.56e-08
## PLOD2
            2.32
                   4.99 7.53e-11 5.73e-08
## TMEM45A
            2.25
                   4.19 4.06e-10 2.91e-07
## ENO2
                   5.72 8.25e-10 5.47e-07
            2.24
                   3.14 8.55e-10 5.47e-07
## PPFIA4
            3.33
                   4.44 2.25e-09 1.37e-06
## PFKFB3
            1.49
resultsTbl.tgw <- topTags( de.tgw , n = nrow( de.tgw$table ) )$table</pre>
de.genes.tgw <- rownames( resultsTbl.tgw )[ resultsTbl.tgw$FDR <= 0.05 ]</pre>
length( de.genes.tgw )
                             # How many genes with significant DE?
## [1] 105
summary( decideTestsDGE( de.tgw , p.value = 0.05 ) )
                                                              # show Up and Down regulated genes
##
      Normoxia+Hypoxia
## -1
                      1
## 0
                  12057
## 1
                   104
write.table(resultsTbl.tgw, file = "LiData_egeR_.csv" , sep = "," , row.names = TRUE )
# xgenes \leftarrow cat(noquote(row.names(resultsTbl.tgw)), sep="\n")
```

Took list of top DE genes and used DAVID to find Gene Ontology and KEGG enriched functional groups. Went to website: https://david.ncifcrf.gov/summary.jsp Uploaded gene list, selected Homo sapiens:

I picked the top 225 genes, selected Homo sapiens, which narrowed the Gene List to 213 genes.

Here is an image of the process: Explored various ways to import the URL tab delim file.

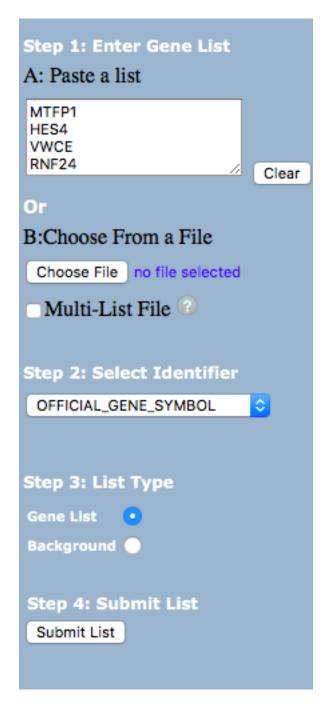


Figure 1: Pasting gene list.

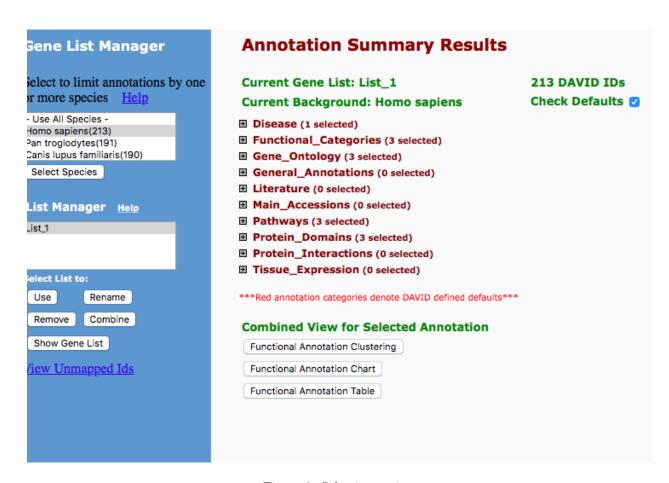


Figure 2: Selecting spcies.

```
# ![Selecting spcies.] (selectingspecies.png)
Current Gene List: List 1 Current Background: Homo sapiens 213 DAVID IDs —
functional annotation clustering for Homo sapiens
library(readr)
library(RCurl)
## Loading required package: bitops
  URL <-"https://david.ncifcrf.gov/data/download/t2t_0C32396F8F541508972082264.txt"</pre>
# Final_Clustering <- read.table(URL, header=FALSE, sep="\t")</pre>
functional.annotaton.clustering.dat <- fread(URL, header=T)</pre>
## Warning in fread(URL, header = T): Starting data input on line 2 and
## discarding line 1 because it has too few or too many items to be column
## names or data: Annotation Cluster 1 Enrichment Score: 9.178650479827342
## Warning in fread(URL, header = T): Stopped reading at empty line 15 but
## text exists afterwards (discarded): Annotation Cluster 2 Enrichment Score:
## 4.539764866201056
head(functional.annotaton.clustering.dat)
                                                                         %
##
                                                           Term Count
              Category
## 1: GOTERM_BP_DIRECT
                               GO:0061621~canonical glycolysis
                                                                   14 6.57
## 2: GOTERM BP DIRECT
                                 GO:0006096~glycolytic process
                                                                   14 6.57
## 3:
           UP KEYWORDS
                                                     Glycolysis
                                                                   12 5.63
## 4:
                        hsa00010:Glycolysis / Gluconeogenesis
                                                                   14 6.57
          KEGG_PATHWAY
              BIOCARTA h_glycolysisPathway:Glycolysis Pathway
## 5:
                                                                    7 3.29
                                    hsa01200:Carbon metabolism
## 6:
          KEGG_PATHWAY
                                                                   13 6.10
        PValue
## 1: 1.68e-19
## 2: 1.38e-17
## 3: 3.48e-15
## 4: 1.57e-12
## 5: 1.07e-09
## 6: 1.68e-08
##
## 1: ALDOA, PFKL, PFKFB4, PFKFB3, ALDOC, HK2, PGAM1, PFKP, HK1, TPI1, ENO2, PGK1, GAPDH, ENO1
       ALDOA, LDHA, PFKL, ALDOC, HK2, PGAM1, HK1, TPI1, PGM1, ENO2, PGK1, GAPDH, EDARADD, ENO1
## 2:
## 3:
                      ALDOA, TPI1, PFKL, ALDOC, ENO2, PGAM1, HK2, PFKP, HK1, PGK1, GAPDH, ENO1
## 4:
          ALDOA, LDHA, PFKL, ALDOC, HK2, PGAM1, PFKP, HK1, TPI1, PGM1, ENO2, PGK1, GAPDH, ENO1
## 5:
                                                       TPI1, PFKL, PGAM1, HK1, PGK1, GAPDH, ENO1
                ALDOA, TPI1, PFKL, ALDOC, ENO2, PGAM1, HK2, PFKP, HK1, PGK1, GPT2, GAPDH, ENO1
## 6:
##
      List Total Pop Hits Pop Total Fold Enrichment Bonferroni Benjamini
## 1:
             183
                        26
                               16792
                                               49.41
                                                        1.65e-16
                                                                  1.65e-16
## 2:
             183
                        34
                               16792
                                               37.78
                                                        1.36e-14
                                                                  6.80e-15
## 3:
             199
                        31
                               20581
                                               40.03
                                                       8.29e-13
                                                                  8.29e-13
## 4:
              90
                        67
                                6910
                                               16.04
                                                       2.22e-10
                                                                  2.22e-10
## 5:
              25
                       10
                                1625
                                               45.50
                                                        6.66e-08
                                                                  6.66e-08
## 6:
              90
                      113
                                6910
                                                8.83
                                                        2.36e-06
                                                                 7.88e-07
##
           FDR
## 1: 2.65e-16
## 2: 2.18e-14
```

## 3: 4.44e-12 ## 4: 1.86e-09 ## 5: 1.08e-06 ## 6: 1.98e-05

Annotation Cluster 1	Enrichment Score: 9.18	G	The second secon	Count	P_Value	Benjamini
GOTERM_BP_DIRECT	canonical glycolysis	RT	=	14	1.7E- 19	1.7E-16
GOTERM_BP_DIRECT	glycolytic process	RT	=	14	1.4E- 17	6.8E-15
UP_KEYWORDS	Glycolysis	RT	=	12	3.5E- 15	8.3E-13
KEGG_PATHWAY	Glycolysis / Gluconeogenesis	RT	=	14	1.6E- 12	2.2E-10
BIOCARTA	Glycolysis Pathway	<u>RT</u>	<b>=</b>	7	1.1E-9	6.7E-8
KEGG_PATHWAY	Carbon metabolism	<u>RT</u>	=	13	1.7E-8	7.9E-7
GOTERM_BP_DIRECT	gluconeogenesis	<u>RT</u>	<u>=</u>	9	2.1E-8	6.8E-6
KEGG_PATHWAY	Biosynthesis of amino acids	RT	<u> </u>	11	2.7E-8	9.7E-7
KEGG_PATHWAY	Biosynthesis of antibiotics	<u>RT</u>	<u> </u>	16	6.3E-8	1.8E-6
KEGG_PATHWAY	Metabolic pathways	RT	_	32	8.0E-5	1.4E-3
UP_SEQ_FEATURE	binding site:Substrate	<u>RT</u>	=	12	1.7E-4	2.8E-2
UP_SEQ_FEATURE	active site:Proton acceptor	<u>RT</u>	=	12	6.2E-2	9.4E-1

Figure 3: Example: Annotation Cluster 1

```
# functional.annotaton.clustering.dat <- read.csv(textConnection(myfile), header=F)
# head(functional.annotaton.clustering.dat)</pre>
```

Current Gene List: List\_1 Current Background: Homo sapiens 213 DAVID IDs ——— Importing the Functional Annotation Chart

```
URL2 <- "https://david.ncifcrf.gov/data/download/chart_0C32396F8F541508972594179.txt"
Functional_Annotation_Chart <- fread(URL2, header=T)
head(Functional_Annotation_Chart)</pre>
```

```
Term Count
                                                                           %
##
              Category
## 1: GOTERM_BP_DIRECT
                                 GO:0061621~canonical glycolysis
                                                                     14 6.57
## 2: GOTERM_BP_DIRECT
                                   GO:0006096~glycolytic process
                                                                     14 6.57
## 3:
           UP_KEYWORDS
                                                      Glycolysis
                                                                     12 5.63
## 4:
          KEGG_PATHWAY
                          hsa00010:Glycolysis / Gluconeogenesis
                                                                     14 6.57
          KEGG_PATHWAY hsa00051:Fructose and mannose metabolism
## 5:
                                                                     10 4.69
## 6:
              BIOCARTA
                         h_glycolysisPathway:Glycolysis Pathway
                                                                      7 3.29
##
        PValue
## 1: 1.68e-19
## 2: 1.38e-17
## 3: 3.48e-15
## 4: 1.57e-12
## 5: 1.42e-10
## 6: 1.07e-09
##
                                                                                           Genes
## 1: ALDOA, PFKL, PFKFB4, PFKFB3, ALDOC, HK2, PGAM1, PFKP, HK1, TPI1, ENO2, PGK1, GAPDH, ENO1
       ALDOA, LDHA, PFKL, ALDOC, HK2, PGAM1, HK1, TPI1, PGM1, ENO2, PGK1, GAPDH, EDARADD, ENO1
## 2:
                      ALDOA, TPI1, PFKL, ALDOC, ENO2, PGAM1, HK2, PFKP, HK1, PGK1, GAPDH, ENO1
## 3:
## 4:
          ALDOA, LDHA, PFKL, ALDOC, HK2, PGAM1, PFKP, HK1, TPI1, PGM1, ENO2, PGK1, GAPDH, ENO1
## 5:
                                  ALDOA, TPI1, MPI, PFKL, PFKFB4, PFKFB3, ALDOC, HK2, PFKP, HK1
## 6:
                                                      TPI1, PFKL, PGAM1, HK1, PGK1, GAPDH, ENO1
##
      List Total Pop Hits Pop Total Fold Enrichment Bonferroni Benjamini
                              16792
## 1:
             183
                       26
                                                49.4
                                                       1.65e-16 1.65e-16
```

```
## 2:
             183
                        34
                               16792
                                                37.8
                                                        1.36e-14
                                                                  6.80e-15
## 3:
             199
                        31
                               20581
                                                40.0
                                                        8.29e-13
                                                                  8.29e-13
                                                                  2.22e-10
## 4:
              90
                        67
                                6910
                                                16.0
                                                        2.22e-10
## 5:
              90
                        32
                                6910
                                                24.0
                                                        2.01e-08 1.00e-08
## 6:
              25
                        10
                                1625
                                                45.5
                                                        6.66e-08 6.66e-08
##
           FDR
## 1: 2.65e-16
## 2: 2.18e-14
## 3: 4.44e-12
## 4: 1.86e-09
## 5: 1.68e-07
## 6: 1.08e-06
Functional Annotation Table Current Gene List: List_1 Current Background: Homo sapiens 28387 DAVID
#importing functional annotation table
URL3 <- "https://david.ncifcrf.gov/data/download/tr_0C32396F8F541508973016208.txt"</pre>
myfile <- getURL(URL3, ssl.verifyhost=FALSE, ssl.verifypeer=FALSE)</pre>
Functional_Annotation_Table <- read.delim(textConnection(myfile), header=T, sep = "\t")
head(Functional_Annotation_Table)
##
          ID
## 1
        GBE1
## 2
     PFKFB3
## 3 PFKFB4
## 4 ADPRHL2
## 5
       ABCB6
## 6
       BSPRY
##
                                                                      Gene.Name
## 1
                                     1,4-alpha-glucan branching enzyme 1(GBE1)
## 2
                6-phosphofructo-2-kinase/fructose-2,6-biphosphatase 3(PFKFB3)
                6-phosphofructo-2-kinase/fructose-2,6-biphosphatase 4(PFKFB4)
## 3
## 4
                                          ADP-ribosylhydrolase like 2(ADPRHL2)
## 5 ATP binding cassette subfamily B member 6 (Langereis blood group)(ABCB6)
                                       B-box and SPRY domain containing(BSPRY)
## 6
          Species BBID BIOCARTA
##
## 1 Homo sapiens
## 2 Homo sapiens
## 3 Homo sapiens
## 4 Homo sapiens
## 5 Homo sapiens
## 6 Homo sapiens
##
                                                        COG ONTOLOGY
## 1
                             Carbohydrate transport and metabolism,
## 2
## 3
## 5 Posttranslational modification, protein turnover, chaperones,
## 6
##
## 1
## 2
                        GO:0006000~fructose metabolic process,GO:0006003~fructose 2,6-bisphosphate meta
## 3
                                                       GO:0006000~fructose metabolic process,GO:0006003~f
```

```
## 4
## 5 GO:0006779~porphyrin-containing compound biosynthetic process,GO:0006810~transport,GO:0006879~cell
##
## 1
## 2
## 3
## 4
## 5 GO:0000139~Golgi membrane,GO:0005739~mitochondrion,GO:0005740~mitochondrial envelope,GO:0005741~mi
## 6
##
                                                                                          GO:0003844~1,4-a
## 1
## 2
## 3
## 4
## 5 GO:0005524~ATP binding,GO:0015232~heme transporter activity,GO:0015439~heme-transporting ATPase ac
## 6
##
## 1 IPR004193:Glycoside hydrolase, family 13, N-terminal, IPR006047:Glycosyl hydrolase, family 13, cata
## 3
## 4
## 5
## 6
##
                                                                                                      KEGG
                                             hsa00500:Starch and sucrose metabolism, hsa01100:Metabolic p
## 2 hsa00051:Fructose and mannose metabolism, hsa04066:HIF-1 signaling pathway, hsa04152:AMPK signaling
                                       hsa00051:Fructose and mannose metabolism, hsa04152:AMPK signaling
## 3
## 4
## 5
                                                                                        hsa02010:ABC trans
## 6
##
## 1
## 2
## 3
## 4
## 5 111600~Blood group, Langereis system,609153~Pseudohyperkalemia, familial, 2, due to red cell leak,
## 6
##
                                     PIR SUPERFAMILY
                                                                          SMART
## 1 PIRSF000463:1,4-alpha-glucan branching enzyme,
                                                                  SM00642: Aamy,
                                                               SM00855:SM00855,
## 3
                                                               SM00855:SM00855,
## 4
## 5
                                                                   SM00382:AAA,
## 6
                                                     SM00449:SPRY,SM00589:PRY,
##
## 1
## 2
## 3
## 4
## 5 3D-structure, Alternative splicing, ATP-binding, Cell membrane, Complete proteome, Disease mutation, Dys.
## 6
##
## 1
```