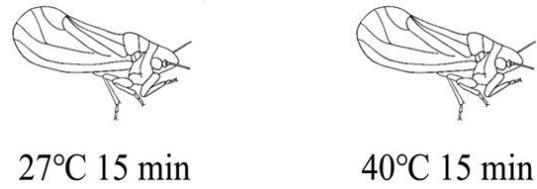


WEEK 15. PRACTICE

Differential expression analysis using DESeq2

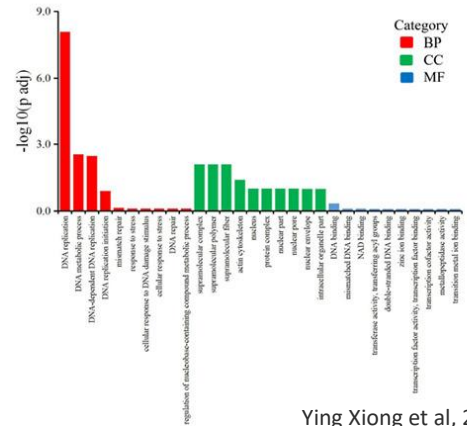
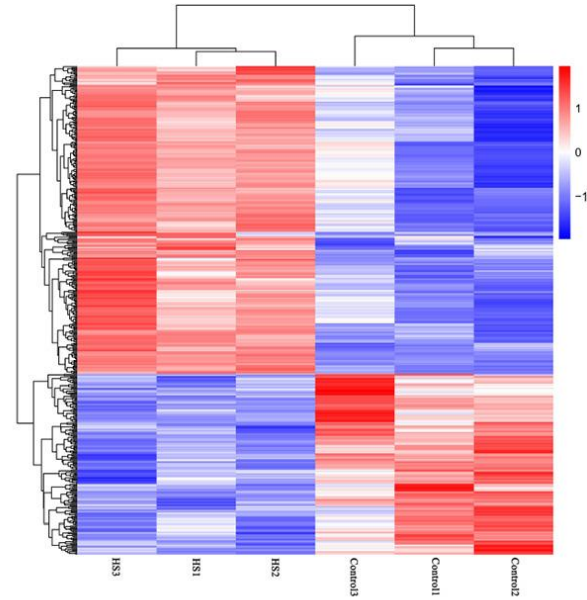
Differentially expressed gene(DEG) analysis



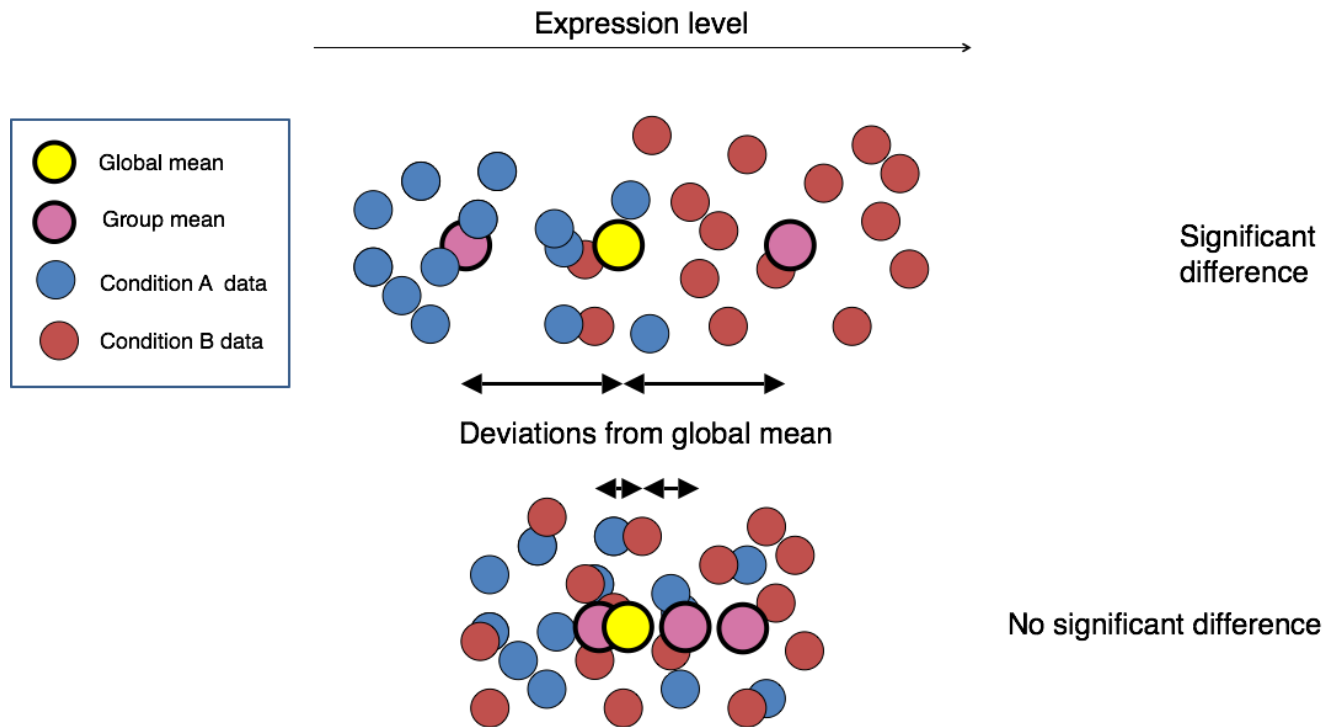
RNA-seq

Differentially expressed genes

Gene ontology analysis



Differential expression analysis with DESeq2



Input data for this week

From www.bloodjournal.org by guest on December 10, 2018. For personal use only.

Regular Article

MYELOID NEOPLASIA

GPR56 identifies primary human acute myeloid leukemia cells with high repopulating potential in vivo

Caroline Pabst,^{1,2} Anne Bergeron,³ Vincent-Philippe Lavallée,^{1,4} Jonathan Yeh,¹ Patrick Gendron,⁵ Gudmundur L. Norddahl,⁶ Jana Krosl,¹ Isabel Boivin,¹ Eric Deneault,¹ Jessica Simard,³ Suzan Imren,⁶ Geneviève Boucher,⁵ Kolja Eppert,⁷ Tobias Herold,⁸ Stefan K. Bohlander,⁹ Keith Humphries,⁶ Sébastien Lemieux,^{5,10} Josée Hébert,^{4,11,12,*} Guy Sauvageau,^{1,4,11,12,*} and Frédéric Barabé^{3,13,14,*}

- The data reported in this article have been deposited in the Gene Expression Omnibus database (accession numbers [GSE49642](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE49642), [GSE52656](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE52656), [GSE62190](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE62190), [GSE66917](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE66917), [GSE67039](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE67039), [GSE48843](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE48843), [GSE48846](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE48846), and [GSE51984](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE51984)).

- RNA-seq of T-cells and B-cells (5 replicates for each cell types)
- the genes whose variances of RNA-seq expression values (FPKM) among the samples are high were selected (501 genes)

<https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE51984>

install R studio

DOWNLOAD

RStudio Desktop

Used by millions of people weekly, the RStudio integrated development environment (IDE) is a set of tools built to help you be more productive with R and Python.

Don't want to download or install anything? Get started with RStudio on [Posit Cloud for free](#). If you're a professional data scientist looking to download RStudio and also need common enterprise features, don't hesitate to [book a call with us](#).

1: Install R

RStudio requires R 3.3.0+. Choose a version of R that matches your computer's operating system.

DOWNLOAD AND INSTALL R

2: Install RStudio

DOWNLOAD RSTUDIO DESKTOP FOR WINDOWS

Size: 214.34 MB | [SHA-256: FE62B784](#) | Version: 2023.09.1+494 | Released: 2023-10-17

```
if (!require("BiocManager", quietly = TRUE))  
  install.packages("BiocManager")
```

```
BiocManager::install("DESeq2")  
install.packages("ggplot2")  
install.packages("reshape2")
```

Identification of DEGs using DESeq2

- DESeq2

- test for differential expression based on a model using the negative binomial distribution

- Usage (in R):

```
dds <- DESeqDataSetFromMatrix(countData = <read count table>,  
                              colData = <label table>,  
                              design = ~ <group attribute in label>)  
  
dds <- DESeq(dds)
```

Input files for the DESeq2

```
cp /home/biguser/tutor/Week15/inputTables/* .
```

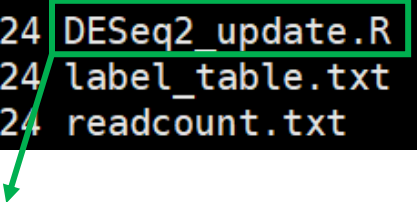
```
-rw-r--r-- 1 biguser biguser 3593 Dec 8 15:24 DESeq2 update.R  
-rw-r--r-- 1 biguser biguser 311 Dec 8 15:24 label_table.txt  
-rw-r--r-- 1 biguser biguser 39129 Dec 8 15:24 readcount.txt
```

```
Sample group paired  
B_cells_01 Bcell paired-end  
B_cells_02 Bcell paired-end  
B_cells_03 Bcell paired-end  
B_cells_04 Bcell paired-end  
B_cells_05 Bcell paired-end  
T_cells_01 Tcell paired-end  
T_cells_02 Tcell paired-end  
T_cells_03 Tcell paired-end  
T_cells_04 Tcell paired-end  
T_cells_05 Tcell paired-end
```

```
ID Symbol B_cells_01 B_cells_02 B_cells_03 B_cells_04 B_cells_05 T_cells_01 T_cells_02  
ENSG0000019582.10 CD74 164084 393381 335146 377965 179726 7248 9580 11369 11884 9673  
ENSG00000204287.9 HLA-DRA 78780 135217 105086 126555 60643 836 462 412 1231 637  
ENSG00000198712.1 MT-CO2 155975 411920 263438 152005 87205 129234 111245 141479 90318 77609  
ENSG00000112306.7 RPS12 30589 78362 47051 58940 29070 67696 69411 85748 62374 53368  
ENSG00000198804.2 MT-CO1 380232 911234 651133 468509 252012 310912 279098 372894 264151 243672  
ENSG00000177954.7 RPS27 51124 109201 70998 88332 43779 80835 76318 101795 75735 69858  
ENSG00000156508.13 EEF1A1 172420 417900 239409 379887 200479 403599 419327 527892 379702 368594  
ENSG00000198840.2 MT-ND3 47436 94410 67487 37505 26247 38020 28714 35585 20148 21662  
ENSG00000166710.13 B2M 65091 170656 80238 129395 69703 157766 164030 204816 159849 192607  
ENSG00000212907.2 MT-ND4L 28302 69691 58661 37725 19451 24929 23878 31947 23615 16736  
ENSG00000196126.6 HLA-DRB1 19952 51776 34618 42725 25455 838 625 544 2097 1370
```

R code to run DESeq2

```
-rw-r--r-- 1 biguser biguser 3593 Dec 8 15:24 DESeq2_update.R
-rw-r--r-- 1 biguser biguser 311 Dec 8 15:24 label_table.txt
-rw-r--r-- 1 biguser biguser 39129 Dec 8 15:24 readcount.txt
```



```
##install the DESeq2 package##
#source("https://bioconductor.org/biocLite.R")
#biocLite("DESeq2")

##get a DESeq2 package##

DESeq2 <- function(readCountTable, tableInfoFile, outputFile, gradeIn, gradeBk) {
  library('DESeq2', verbose = F)
  library(ggplot2, verbose = F)
  library(reshape2, verbose = F)

  ##read count matrix table and information table##
  avgReadCount <- read.delim(readCountTable, header = TRUE, sep = '\t', row.names = 1, check.names = FALSE)
  avgReadCount <- avgReadCount[c(-1)] #remove gene symbol column
  avgReadCountInfo <- read.table(tableInfoFile, header = TRUE, sep = '\t', row.names = 1, check.names = FALSE)

  ##DEG run##
  print('Running DESeq2')
  dds <- DESeqDataSetFromMatrix(countData = avgReadCount, colData = avgReadCountInfo, design = ~ group)
  dds <- dds[ rowSums(counts(dds)) > 20, ]
  dds <- DESeq(dds)
  result_05 <- results(dds, alpha = 0.05, contrast=c("group", gradeIn, gradeBk))

  outputPdf <- unlist(strsplit(outputFile, split = '.txt', fixed = TRUE))[1]
  outputMaPdf <- paste(c(outputPdf, '.maplot.pdf'), collapse = '')
  pdf(outputMaPdf, width = 4, height = 4)
  plotMA(result_05, main = paste(c(gradeIn, 'vs', gradeBk), collapse = ' '),
         alpha = 0.05)#, ylim = c(-max(abs(result_05$log2FoldChange)), max(abs(result_05$log2FoldChange))))
  abline(h=c(-2,2), col = 'dodgerblue', lwd = 2)
  dev.off()
```


R code to run DESeq2

```
##whole table##
newColumn <- c('gene','baseMean','log2FoldChange','lfcSE','stat','pvalue','padj')
writeTable <- data.frame(result_05)
writeTable <- data.frame(row.names(writeTable), writeTable)
colnames(writeTable) <- newColumn

outputFile <- strsplit(outputFile, split = 'txt')
allOutFile <- paste(c(outputFile, 'all.txt'), collapse = '')
write.table(writeTable, file = allOutFile, quote = FALSE, sep = '\t', col.names = TRUE, row.names = FALSE)

##adjusted P-value cutoff##
sig_result_05 <- subset(result_05, padj < 0.05)
df_sig_result_05 <- data.frame(sig_result_05)
df_sig_result_05 <- data.frame(row.names(df_sig_result_05), df_sig_result_05)
colnames(df_sig_result_05) <- newColumn

sigOutFile <- paste(c(outputFile, 'sig.txt'), collapse = '')
write.table(df_sig_result_05, file = sigOutFile, quote = FALSE, sep = '\t', col.names = TRUE, row.names = FALSE)
print('DESeq done')

##volcano plot
writeTable <- na.omit(writeTable)
sigDeg <- as.factor(abs(writeTable$log2FoldChange) >= 2 & writeTable$padj <= 0.05)
lgfcMax <- max(c(abs(min(writeTable$log2FoldChange)-0.5), max(writeTable$log2FoldChange)+0.5))
plt <- ggplot(writeTable, aes(log2FoldChange, -log10(padj), colour=sigDeg)) +
  geom_point(alpha=0.4, size=1) + labs(x="log fold change", y="-log10 adjusted P-value", title=paste(c(gradeIn, 'vs', gradeBk), collapse = ' ')) +
  theme_bw() +
  theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(), plot.title = element_text(hjust = 0.5),
        text = element_text(size = 16, colour = 'black'), legend.position="none") +
  scale_color_manual(values=c("#999999", "red2"))
outputVolPdf <- paste(c(outputPdf, '.volcanoplot.pdf'), collapse = '')
ggsave(outputVolPdf, units = 'cm', height = 10, width = 10)
}

args <- commandArgs(trailingOnly = TRUE)
if (args[1] == '-h' | args[1] == '--help'){
  print('Rscript DESeq2_update.R <read count> <label> <output> <group of interest> <background group>')
} else {
  DESeq2(args[1], args[2], args[3], args[4], args[5])
}
```

Running DESeq2

```
$Rscript DESeq2_update.R .readcount.txt label_table.txt  
Bcell_background_vs_Tcell Tcell Bcell
```

```
[1] "Running DESeq2"  
estimating size factors  
estimating dispersions  
gene-wise dispersion estimates  
mean-dispersion relationship  
final dispersion estimates  
fitting model and testing  
[1] "DESeq done"
```

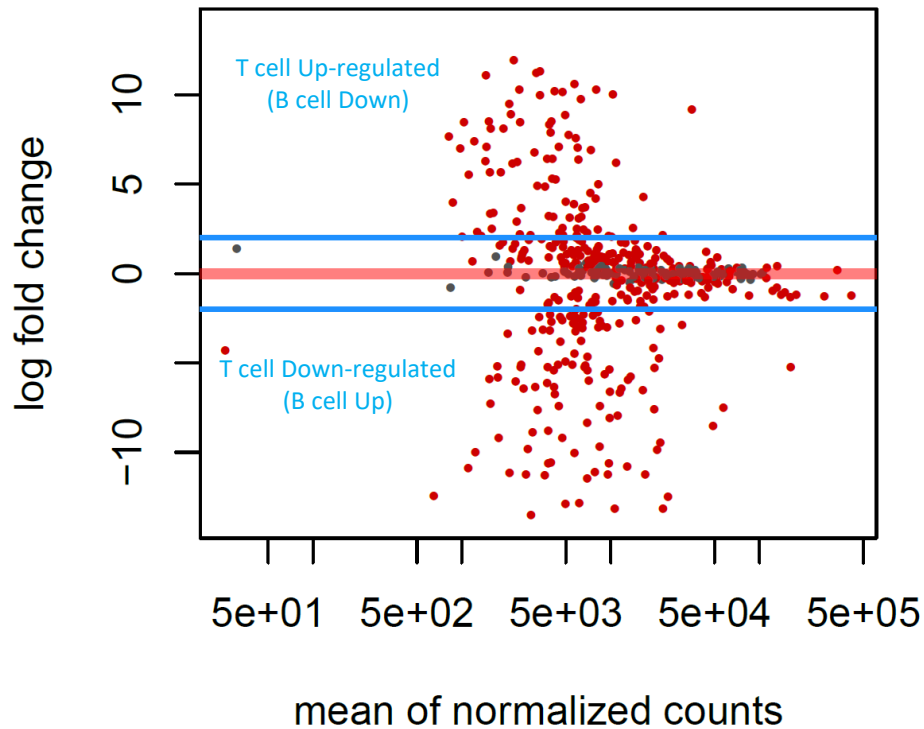
```
-rw-rw-r-- 1 biguser biguser 64380 Dec  8 16:29 Bcell_background_vs_Tcellall.txt all 501 genes  
-rw-rw-r-- 1 biguser biguser  8928 Dec  8 16:29 Bcell_background_vs_Tcell.maplot.pdf  
-rw-rw-r-- 1 biguser biguser 49492 Dec  8 16:29 Bcell_background_vs_Tcellsig.txt significant DEGs  
-rw-rw-r-- 1 biguser biguser 33708 Dec  8 16:29 Bcell_background_vs_Tcell.volcanoplot.pdf
```

```
setwd('~/')  
read_count <- './readcount.txt'  
label <- './label_table.txt'  
output_prefix <- 'Bcell_background_vs_Tcell'  
group_of_interest <- 'Tcell'  
background_group <- 'Bcell'
```

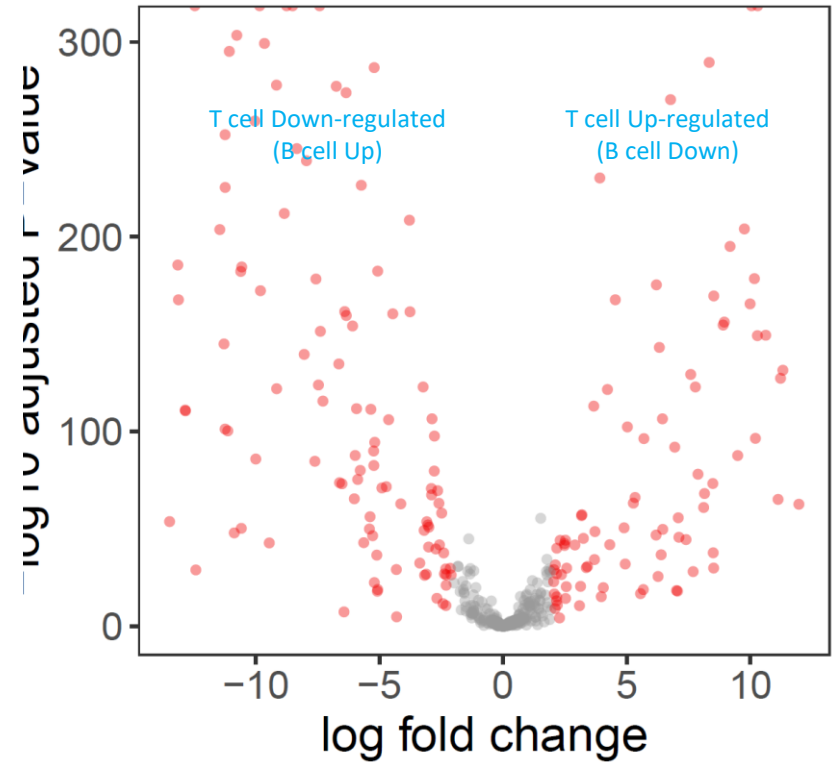
```
DESeq2(read_count, label, output_prefix, group_of_interest, background_group)
```

Output plots for DESeq2

Tcell vs Bcell



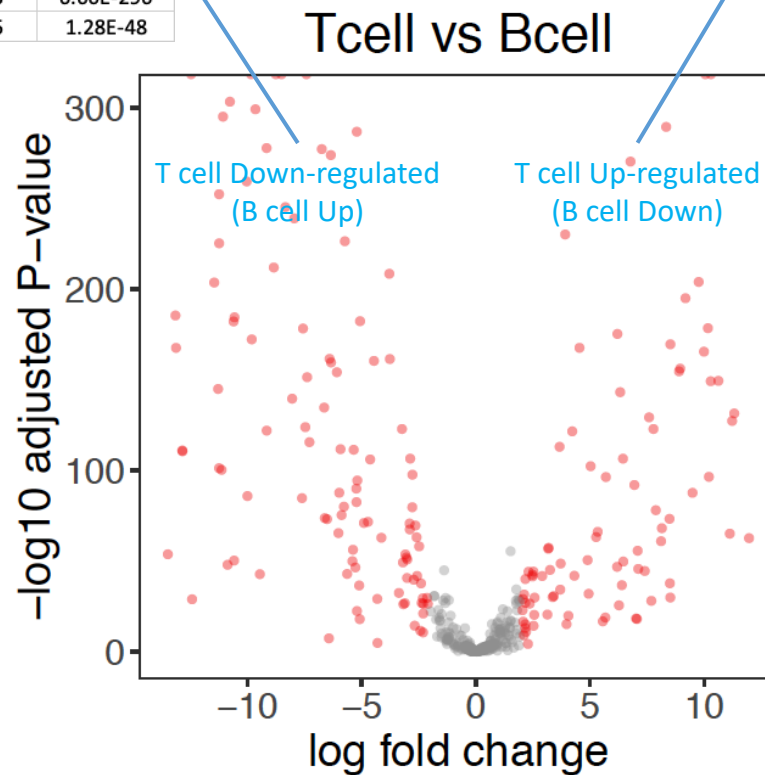
Tcell vs Bcell



Output files for DESeq2

gene	baseMean	log2FoldChange	padj
BLNK	2918.700756	-13.48608045	1.98E-54
IGJ	22518.04467	-13.15386721	3.49E-186
PAX5	10692.86908	-13.12825157	2.51E-168
FCRLA	4976.028019	-12.85521385	1.04E-111
TCL1A	6164.88851	-12.84404967	2.79E-111
CD22	24352.29797	-12.4651748	0
TNFRSF17	649.0131117	-12.42537173	1.34E-29
FCER2	6978.040165	-11.45555395	2.20E-204
SPIB	3614.323174	-11.28601261	1.17E-145
CD180	2702.923748	-11.24383308	6.01E-102
BANK1	17125.89172	-11.24194382	4.12E-253
ADAM28	9602.185366	-11.23896948	4.92E-226
CLEC17A	2095.91957	-11.13254712	5.20E-101
CD19	7848.139949	-11.07137013	6.60E-296
CD1C	1106.907196	-10.87108855	1.28E-48

gene	baseMean	log2FoldChange	padj
CD8B	2234.852487	11.96167336	2.51E-63
CD3D	3356.313502	11.31389779	3.84E-132
CD28	3174.132576	11.22003263	5.19E-128
TRAT1	1454.112144	11.11966502	8.11E-66
CD8A	5733.489469	10.62088863	3.94E-150
ITK	8005.9051	10.29467119	0
CD3G	2435.870636	10.2838521	6.27E-150
CD2	4216.003782	10.20424873	3.64E-97
GIMAP5	4753.338392	10.16280914	3.40E-179
CD3E	10366.87008	10.04740398	0



Python script to extract significant DEGs

```
$ python getSigGenes.py  
Bcell_background_vs_Tcellall.txt  
readcount.txt
```

```
#!/usr/bin/env python  
import sys  
deg_result= sys.argv[1]  
  
deg_open= open(deg_result, "r")  
deg_lineL= deg_open.readlines()  
deg_open.close()  
  
sig_up= list()  
sig_down= list()  
  
for i_line in deg_lineL[1:]: ## First line is header so skip it  
    infoL= i_line.strip().split("\t")  
    geneid= infoL[0]  
    try:  
        l2fc= float(infoL[2])  
        padj= float(infoL[6])  
    except :  
        continue  
  
    if padj > 0.05: ## padj cutoff is 0.05  
        continue  
    if l2fc > 2.0:  
        sig_up.append(geneid)  
    elif l2fc < -2.0:  
        sig_down.append(geneid)  
    else:  
        pass  
  
print("Number of genes highly expressed in T-cell :", len(sig_up))  
print("Number of genes highly expressed in B-cell :", len(sig_down))
```

```
Number of genes highly expressed in T-cell : 85  
Number of genes highly expressed in B-cell : 106
```

Python script to extract significant DEGs

```
countfile= sys.argv[2]

count_open= open(countfile, "r")
count_lineL= count_open.readlines()
count_open.close()

gene_idsymbolD= dict()
for i_line in count_lineL[1:]: ## first line is header
    infoL= i_line.strip().split("\t")
    geneid= infoL[0]
    genesymbol= infoL[1]
    gene_idsymbolD[geneid]= genesymbol

## Function to write output gene list (gene symbol)
def writeOutput(outputname, genelist, gene_idsymbolD):
    fileopen= open(outputname, "w")
    for i_gene in genelist:
        #
        geneid= i_gene[0]
        genesymbol= gene_idsymbolD[i_gene]
        outputline= genesymbol+ "\n"
        fileopen.write(outputline)
    fileopen.close()

## T-cell specific genes
tcell_output= "tcell_specific_gene_symbols.txt"
writeOutput(tcell_output, sig_up, gene_idsymbolD)

## B-cell specific genes
bcell_output= "bcell_specific_gene_symbols.txt"
writeOutput(bcell_output, sig_down, gene_idsymbolD)
```

List of genes (symbols) that are significantly, differentially expressed

```
-rw-rw-r-- 1 biguser biguser 678 Dec 8 16:46 bcell_specific_gene_symbols.txt  
-rw-r--r-- 1 biguser biguser 1517 Dec 8 16:47 getSigGenes.py  
drwxr-xr-x 2 biguser biguser 116 Dec 8 16:39 inputTables  
-rw-rw-r-- 1 biguser biguser 479 Dec 8 16:46 tcell_specific_gene_symbols.txt
```

```
CD74  
HLA-DRA  
HLA-DRB1  
CD79A  
IGJ  
MS4A1  
CD79B  
FCRL1  
HLA-DPB1  
HLA-DRB5  
CD37  
CD22  
HLA-DMA  
HLA-DPA1  
FAM129C  
FCER2  
CD19  
BANK1  
TCL1A  
IL4R  
HLA-DQA2  
HLA-DMB  
BLK  
IRF8  
FCRL2  
VPREB3  
HLA-DQB1  
CTSZ  
TNFRSF13C  
RALGPS2  
SYK  
ALOX5  
HLA-DQA  
CYBB  
MZB1  
P2RX5  
FCRLA  
STAP1  
HLA-DQA1  
RNASE6  
HSH2D  
BTK  
BIRC3  
HLA-DOB  
bcell_specific_gene_symbols.txt
```

```
IL7R  
CCL5  
NKG7  
S100A4  
CD3E  
CD3D  
GNLY  
CD2  
SPOCK2  
IL32  
LCK  
HCST  
GZMH  
SELPG  
ZAP70  
IFITM1  
PRF1  
GZMA  
CST7  
GIMAP4  
GIMAP7  
CD8A  
CTSW  
TCF7  
KLRB1  
GZMM  
ANXA1  
PIM1  
ARL4C  
CD8B  
CD5  
LEF1  
CD6  
GZMB  
GIMAP6  
SAMHD1  
TNFRSF25  
S100A11  
MAL  
CD247  
FGFBP2  
PLCG1  
CD7  
ATP8B2  
tcell_specific_gene_symbols.txt
```

Biological signatures associated with list of genes

The screenshot shows the g:Profiler web interface. At the top, there is a navigation bar with the logo and links for News, Archives, Beta, API, R client, FAQ, Docs, Contact, Cite g:Profiler, Services using g:P, and List of organisms. A blue notification banner states: "g:Profiler has been updated with new data from Ensembl." Below this are four main tool buttons: g:GOST (Functional profiling), g:Convert (Gene ID conversion), g:Orth (Orthology search), and g:SNPense (SNP id to gene name). The main interface is divided into two sections: "Query" and "Options".

Query Section:

- Buttons: Query, Upload query, Upload bed file
- Text: Input is whitespace-separated list of genes
- Text area containing a list of genes: CD74, HLA-DRA, HLA-DRB1, CD79A, IGJ, MS4A1, CD79B, FCRL1, HLA-DPB1, HLA-DRB5, CD37, CD22, HLA-DMA, HLA-DPA1, FAM129C
- Buttons: Run query, random example, mixed query example

Options Section:

- Organism: Homo sapiens (Human)
- Ordered query
- Run as multiquery
- Advanced options (expanded):
 - All results
 - Measure underrepresentation
 - No evidence codes
- Statistical domain scope: Only annotated genes
- Significance threshold: Benjamini-Hochberg FDR
- User threshold: 0.05
- Numeric IDs treated as: ENTREZGENE_ACC
- Data sources
- Bring your data (Custom GMT)

At the bottom, there are tabs for Results, Detailed Results, and Query info (1).