

Problem 1 Preprocessing a data set

Install the "ArrayExpress" package from Bioconductor. Load the yeast microarray data using R commands:

```
library(ArrayExpress)
yeast.raw = ArrayExpress('E-MEXP-1551')
```

(a) Preprocess the raw data set into an expression data set using: the “mas” background correction method, the “quantiles” normalization method, “pmonly” pm correction method and “medianpolish” summary method. Give the R command here for doing this task.

(b) Print out the mean expression values for the first five genes across all samples.

(c) How many genes and how many samples are in the preprocessed expression data set?

A)

Rscript:

```
biocLite("ArrayExpress")
library(ArrayExpress)
yeast.raw = ArrayExpress('E-MEXP-1551')
eset <- expresso(yeast.raw,bgcorrect.method="mas",
normalize.method="quantiles",pmcorrect.method="pmonly",
summary.method="medianpolish")
```

Answer:

```
background correction: mas
normalization: quantiles
PM/MM correction : pmonly
expression values: medianpolish
background correcting...done.
normalizing...done.
10928 ids to be processed
|          |
|#####|
```

B)

Rscript:

```
expressionvalues<- exprs(eset)
expressionvalues[1:5,]
```

Answer:

Gre_MCA_2822 Gre_MCA_5014 Gre_MCA_3174 Gre_MCA_4108
Gre_MCA_3864

1769308_at 9.049451 8.968067 9.203207 8.862262
8.929268

1769309_at 5.582373 5.671969 5.504717 5.718791
5.833522

1769310_at 5.697383 5.536656 5.358618 5.573735
5.720136

1769311_at 11.427141 11.100304 11.550417 11.407650
11.152456

1769312_at 9.870687 9.760277 9.673015 9.888257
9.706339

Gre_MCA_9147 Gre_MCA_8454 Gre_MCA_4493 Gre_MCA_9211
Gre_MCA_3824

1769308_at 8.858289 9.074716 8.453985 8.883578
8.806324

1769309_at 5.700246 5.479024 5.831566 5.516686
5.940651

1769310_at 5.598420 5.828015 5.537205 5.545576
5.626586

1769311_at 11.465841 11.391676 11.395314 11.234055
11.377907

1769312_at 9.822943 9.764995 9.626873 9.882233
9.854645

Gre_MCA_8298 Gre_MCA_6510 Gre_MCA_7031 Gre_MCA_6671
Gre_MCA_2172

1769308_at 8.990849 8.915155 8.836336 9.060353
8.860005

1769309_at 5.734243 5.956699 5.707443 5.942985
5.782136

1769310_at 5.793489 5.760122 5.700734 5.688692
5.687390

1769311_at 11.269368 11.320259 11.483310 11.078343
11.452286

1769312_at 9.718603 9.716625 9.769313 9.539288
9.767874

Gre_MCA_8016	Gre_MCA_7817	Gre_MCA_6886	Gre_MCA_6857
Gre_MCA_9353			
1769308_at	9.235485	8.852164	8.853089
8.801891			
9.198097			
1769309_at	5.413773	5.852047	5.730331
5.740296			
5.203052			
1769310_at	5.531625	5.865332	5.815449
5.571610			
5.559456			
1769311_at	11.604265	11.452660	11.296400
11.466942			
11.623890			
1769312_at	9.629088	9.771854	9.793377
9.764324			
9.688466			
Gre_MCA_1726 Gre_MCA_0356 Gre_MCA_8052 Gre_MCA_9301			
Gre_MCA_5948			
1769308_at	9.134587	8.872302	8.859122
8.807088			
8.982770			
1769309_at	5.741257	5.764816	5.723271
5.554010			
5.573772			
1769310_at	5.543069	5.927136	5.792484
5.569385			
5.688268			
1769311_at	11.500176	11.372680	11.414907
11.305940			
11.320536			
1769312_at	9.734023	9.883580	9.815190
9.779764			
9.910259			
Gre_MCA_7528 Gre_MCA_1103 Gre_MCA_6052 Gre_MCA_2282			
Gre_MCA_3378			
1769308_at	8.886055	9.068463	8.745493
8.997327			
9.038056			
1769309_at	5.707336	5.379089	5.673917
5.419490			
5.601698			
1769310_at	5.470141	5.629583	5.853489
5.389162			
5.655055			
1769311_at	11.462510	11.449835	11.393147
11.463031			
11.195200			
1769312_at	9.807047	9.730797	9.764790
9.579366			
9.560505			

C)

Rscript:

```
dim(exprs(eset))
```

Answer:

```
10928 30
```

10928 genes and 30 samples

Problem 2 Searching Annotations

(a) What is the annotation package for the yeast data set in question 1?

Install the annotation package from Bioconductor.

(b) Search the 1769308_at gene GO numbers related to Molecular Function (MF). How many GO numbers do you get?

(c) Find the GO parents of the GO IDs in part (b). How many GO parents are there?

(d) Find the GO children of the GO IDs in part (b). How many GO children are there?

A)

Rscript:

```
annotation(yeast.raw)
```

Answer:

"yeast2" is the annotation package for the yeast data set

B)

Rscript:

```
biocLite("yeast2.db")
library("yeast2.db")
go<-get("1769308_at", env = yeast2GO)
gomf<-getOntology(go,"MF")
gomf
```

Answer:

```
[1] "GO:0016491" "GO:0003824" "GO:0016616" "GO:0016829"
"GO:0016853" "GO:0004300"
[7] "GO:0003857"
I get 7 GO numbers.
```

C)

Rscript:

```
library("GO.db")
go<-get("1769308_at", env = yeast2GO)
gonr <- getOntology(go, "MF")
gP <- getGOParents(gonr)
```

```
pa <- sapply(gP,function(x) x$Parents)
pa
length(unlist(pa))
```

Answer:

```
GO:0016491.is_a GO:0003824.is_a GO:0016616.is_a GO:0016829.is_a
GO:0016853.is_a
"GO:0003824" "GO:0003674" "GO:0016614" "GO:0003824"
"GO:0003824"
GO:0004300.is_a GO:0003857.is_a
"GO:0016836" "GO:0016616"
```

There are 7 GO parents

```
D)
Rscript:
library("GO.db")
go<-get("1769308_at", env = yeast2GO)
gonr <- getOntology(go, "MF")
gC <- getGOChildren(gonr)
ch <- sapply(gC,function(x) x$Children)
ch
length(unlist(ch))
```

Answer:

423 GO children.

Problem 3 Gene filtering on B-cell ALL patients

We work with the patients in stages "B2","B3".

- (a) We look for genes expressed differently in stages B2 and B3. Use `genefilter` to program the Wilcoxon test and the Welch t-test separately for each gene. For each test, we select the genes with $p\text{-value} < 0.001$. To save computational time, we set `exact=F` in the Wilcoxon test function.
- (b) Compute a Venn diagram for the Wilcoxon test and the t-test, and plot it.
- (c) How many pass the Wilcoxon filter? How many passes both filters?
- (d) What is the annotation package for the ALL data set? Find the GO numbers for “oncogene”.
- (e) How many genes passing the filters in (a) are oncogenes?

A)

Rscript:

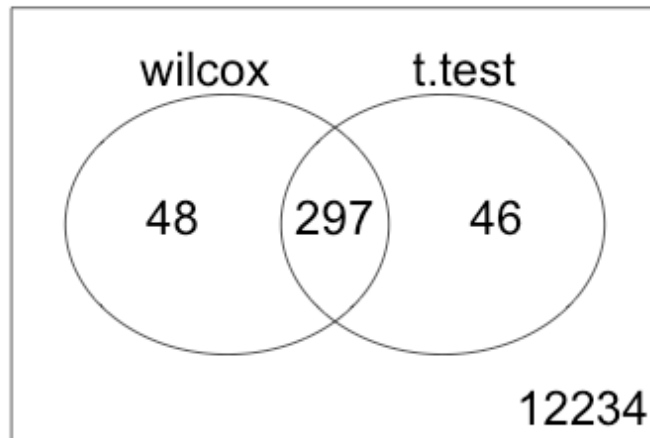
```
library("genefilter")
library("ALL"); data(ALL, package = "ALL");
patientB <- exprs(ALL)[,(ALL$BT %in% c("B2","B3"))]
factor <- droplevels(ALL$BT[ALL$BT %in% c("B2","B3")])
f1 <- function(x) (wilcox.test(x ~ factor, exact = F)$p.value < 0.001)
f2 <- function(x) (t.test(x ~ factor)$p.value < 0.001)
wilcox <- genefilter(patientB, filterfun(f1))
t.test <- genefilter(patientB, filterfun(f2))
```

B)

Rscript:

```
library(limma)
x <- apply(cbind(wilcox,t.test), 2, as.integer)
vc <- vennCounts(x, include="both")
vennDiagram(vc)
```

Answer:



C)

Answer:

From the vein diagram:

How many pass the Wilcoxon filter – $48 + 297 = 345$

How many passes both filters – 297

D)

Rscript:

```
annotation(ALL)
```

```
library("GO.db"); library("annotate"); library("hgu95av2.db")
```

```
GOTerm2Tag <- function(term) {  
  GTL <- eapply(GOTERM, function(x) {grep(term, x@Term, value=TRUE)})  
  Gl <- sapply(GTL, length)  
  names(GTL[Gl>0])  
}
```

```
GOTerm2Tag("oncogene")
```


Answer:

annotation package for the ALL data set - hgu95av2

the GO numbers for "oncogene" - "GO:0090402"

E)

Rscript:

```
selected<- wilcox & t.test
```

```
ALLs<-ALL[selected,]
```

```
tran <- hgu95av2GO2ALLPROBES$"GO:0090402"
```

```
inboth <- tran %in% row.names(exprs(ALLs))
```

```
print(sum(inboth))
```

Answer:

0 genes passing the filters in (a) are oncogenes

Problem 4

Stages of B-cell ALL in the ALL data. Use the limma package to answer the questions below.

(a) Select the persons with B-cell leukemia which are in stage B1, B2, and B3.

(b) Use the linear model to test the hypothesis of all zero group means. Use “topTable()” to report the **top five** genes with nonzero means in **B3 group**.

(c) Use two contrasts to perform analysis of variance to test the null hypothesis of equal group means. Do this with a false discovery rate of 0.01. **How many** differentially expressed genes are found? Use “topTable()” to report the top five genes that express differently among the three groups.

A)

Rscript:

```
library("limma"); library("ALL"); data(ALL, package = "ALL");
allB <- ALL[,which(ALL$BT %in% c("B1","B2","B3"))] #Patients in 3 stages
exprs(allB)
```

B)

Rscript:

```
library("limma"); library("ALL"); data(ALL, package = "ALL");
allB <- ALL[,which(ALL$BT %in% c("B1","B2","B3"))]
design.ma <- model.matrix(~ 0 + factor(allB$BT))
colnames(design.ma) <- c("B1","B2","B3")
fit <- lmFit(allB,design.ma)
fit <- eBayes(fit)

sum(topTable(fit, number=Inf,adjust.method="fdr")$adj.P.Val<0.05)

print( topTable(fit, coef=3, number=5, adjust.method="fdr"), digits=4)
```

Answer:

	logFC	AveExpr	t	P.Value	adj.P.Val	B
AFFX-hum_alu_at	13.61	13.53	355.6	5.059e-127	6.387e-123	270.8
32466_at	12.71	12.71	316.7	4.247e-123	2.681e-119	263.9
31962_at	13.05	13.09	307.1	4.695e-122	1.976e-118	262.0
32748_at	12.15	12.12	302.8	1.407e-121	4.406e-118	261.2
35278_at	12.52	12.48	302.0	1.745e-121	4.406e-118	261.0

Above we see all 12625 genes have p value less than 0.05 hence we can reject the null hypothesis off all zero group means and conclude that all the genes express differently.

Top five genes with nonzero means in B3 group are:

AFFX-hum_alu_at

32466_at

31962_at

32748_at

35278_at

C)

Rscript:

```
cont.ma <- makeContrasts(B1-B2,B2-B3, levels=factor(allB$BT))
```

```
fit1 <- contrasts.fit(fit, cont.ma)
```

```
fit1 <- eBayes(fit1)
```

```
sum(topTable(fit1,number=Inf,adjust.method="fdr")$adj.P.Val<0.01)
```

```
print(topTable(fit1, number=5,adjust.method="fdr"), digits=4)
```

Answer:

	B1...B2	B2...B3	AveExpr	F	P.Value	adj.P.Val
1389_at	-1.7852	-0.74038	9.678	49.15	1.532e-14	1.934e-10
1914_at	2.0976	0.35648	4.693	42.20	3.785e-13	2.389e-09
33358_at	1.4890	-0.20733	5.214	29.52	2.837e-10	1.194e-06
38555_at	0.8058	0.62321	6.124	25.93	2.322e-09	7.329e-06
40763_at	1.5921	-0.01192	3.220	23.08	1.337e-08	2.758e-05

With false discovery rate of 0.01, 314 genes are found to be differentially expressed

The top five genes that express differently among the three groups are:

1389_at

1914_at

33358_at

38555_at

40763_at