

Analysis and Exploration of Proteomics Data

R Markdown

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Proteomics Data Analysis

Below are shown the steps typically used to perform Differential Gene Expression analysis. The analysis steps have been applied over the *EGF-driven protein synthesis* transcriptomics data from D.A. Rothenberg et al. A Proteomics Approach to Profiling the Temporal Translational Response to Stress and Growth. iScience. 2018; 9:367-381.

Loading of R-Packages

We start by loading the R-packages we need to use for our analysis:

```
library(edgeR)
library(lattice)
library(biomaRt)
library(gplots)
library(gridExtra)
library(openxlsx)
library(tools)
library(ggplot2)
library(dplyr)
library(Glimma)
library(openxlsx)
```

Loading of the data

We load the generated **subread_counts** object, define the experimental conditions for each available sample and prepare the data in the *DGEObj* format which represents a flexible container to manage and annotate Differential Gene Expression (DGE) analysis results.

```
load("course.Rdata")

groups <- factor(c("PBS", "EGF", "PBS", "EGF"))

label <- c("PBS2", "EGF1", "PBS1", "EGF2")

print(colnames(subread_counts$counts))

## [1] "SRR7451179.Rep.2.MP60.2.mRNA.Seq.PBS.bam"
## [2] "SRR7451182.Rep.1.ME60.1.mRNA.Seq.EGF.bam"
## [3] "SRR7451187.Rep.1.MP60.1.mRNA.Seq.PBS.bam"
## [4] "SRR7451201.Rep.2.ME60.2.mRNA.Seq.EGF.bam"

print(colnames(subread_counts$counts))
```

```
## [1] "SRR7451179.Rep.2.MP60.2.mRNA.Seq.PBS.bam"
## [2] "SRR7451182.Rep.1.ME60.1.mRNA.Seq.EGF.bam"
## [3] "SRR7451187.Rep.1.MP60.1.mRNA.Seq.PBS.bam"
## [4] "SRR7451201.Rep.2.ME60.2.mRNA.Seq.EGF.bam"

DGEObj <- DGEList(group=groups, counts=subread_counts$counts, genes=subread_counts$annotation[,c("GeneID", "Length")])

print(DGEObj)

## An object of class "DGEList"
## $counts
##               SRR7451179.Rep.2.MP60.2.mRNA.Seq.PBS.bam
## ENSG00000223972                                0
## ENSG00000227232                                14
## ENSG00000278267                                0
## ENSG00000243485                                0
## ENSG00000284332                                0
##               SRR7451182.Rep.1.ME60.1.mRNA.Seq.EGF.bam
## ENSG00000223972                                0
## ENSG00000227232                                24
## ENSG00000278267                                0
## ENSG00000243485                                0
## ENSG00000284332                                0
##               SRR7451187.Rep.1.MP60.1.mRNA.Seq.PBS.bam
## ENSG00000223972                                0
## ENSG00000227232                                10
## ENSG00000278267                                0
## ENSG00000243485                                0
## ENSG00000284332                                0
##               SRR7451201.Rep.2.ME60.2.mRNA.Seq.EGF.bam
## ENSG00000223972                                0
## ENSG00000227232                                22
## ENSG00000278267                                0
## ENSG00000243485                                0
## ENSG00000284332                                0
## 58879 more rows ...
##
## $samples
##               group lib.size norm.factors
## SRR7451179.Rep.2.MP60.2.mRNA.Seq.PBS.bam  PBS 26653631          1
## SRR7451182.Rep.1.ME60.1.mRNA.Seq.EGF.bam  EGF 30763088          1
## SRR7451187.Rep.1.MP60.1.mRNA.Seq.PBS.bam  PBS 28377976          1
## SRR7451201.Rep.2.ME60.2.mRNA.Seq.EGF.bam  EGF 31739139          1
##
## $genes
##               GeneID Length
## ENSG00000223972 ENSG00000223972 1735
## ENSG00000227232 ENSG00000227232 1351
## ENSG00000278267 ENSG00000278267  68
## ENSG00000243485 ENSG00000243485 1021
## ENSG00000284332 ENSG00000284332  138
## 58879 more rows ...
```

Experimental Design

We make an experimental design matrix where we indicate the condition and replicate ID's for each sample. We then define the testing contrasts of our experimental design (*EGF_vs_PBS*). Our targeted *EGF_vs_PBS* contrast tests whether the average expression across all *EGF* groups is equal to the average expression across all *PBS* groups using *makeContrasts*.

```
design <- model.matrix(~0+groups)
colnames(design) <- levels(groups)
print(design)
```

```
##      EGF PBS
## 1    0    1
## 2    1    0
## 3    0    1
## 4    1    0
## attr("assign")
## [1] 1 1
## attr("contrasts")
## attr("contrasts")$groups
## [1] "contr.treatment"
```

```
contrasts <- makeContrasts(
  EGF_vs_PBS = EGF - PBS,
  levels=design
)
counts <- DGEObj$counts
```

DGEObj Processing

We simplify a bit the names of the samples.

```
head(DGEObj$samples)
```

```
##                                     group lib.size norm.factors
## SRR7451179.Rep.2.MP60.2.mRNA.Seq.PBS.bam   PBS 26653631          1
## SRR7451182.Rep.1.ME60.1.mRNA.Seq.EGF.bam   EGF 30763088          1
## SRR7451187.Rep.1.MP60.1.mRNA.Seq.PBS.bam   PBS 28377976          1
## SRR7451201.Rep.2.ME60.2.mRNA.Seq.EGF.bam   EGF 31739139          1

rownames(DGEObj$samples) <- gsub("\\.", "_", rownames(DGEObj$samples))
rownames(DGEObj$samples) <- gsub("KD_", "", rownames(DGEObj$samples))
rownames(DGEObj$samples) <- gsub("_circSLC8A1", "", rownames(DGEObj$samples))
rownames(DGEObj$samples) <- gsub("_bam", "", rownames(DGEObj$samples))
rownames(DGEObj$counts) <- gsub("_1_mRNA_Seq_PBS", "", rownames(DGEObj$counts))
rownames(DGEObj$counts) <- gsub("_1_mRNA_Seq_EGF", "", rownames(DGEObj$counts))
rownames(DGEObj$counts) <- gsub("_2_mRNA_Seq_PBS", "", rownames(DGEObj$counts))
rownames(DGEObj$counts) <- gsub("_2_mRNA_Seq_EGF", "", rownames(DGEObj$counts))

colnames(DGEObj$counts) <- gsub("\\.", "_", colnames(DGEObj$counts))
colnames(DGEObj$counts) <- gsub("KD_", "", colnames(DGEObj$counts))
colnames(DGEObj$counts) <- gsub("_1_mRNA_Seq_PBS", "", colnames(DGEObj$counts))
colnames(DGEObj$counts) <- gsub("_1_mRNA_Seq_EGF", "", colnames(DGEObj$counts))
colnames(DGEObj$counts) <- gsub("_2_mRNA_Seq_PBS", "", colnames(DGEObj$counts))
colnames(DGEObj$counts) <- gsub("_2_mRNA_Seq_EGF", "", colnames(DGEObj$counts))
colnames(DGEObj$counts) <- gsub("_bam", "", colnames(DGEObj$counts))
```

```
DGEObj$samples
```

```
##                               group lib.size norm.factors
## SRR7451179_Rep_2_MP60_2_mRNA_Seq_PBS   PBS 26653631         1
## SRR7451182_Rep_1_ME60_1_mRNA_Seq_EGF   EGF 30763088         1
## SRR7451187_Rep_1_MP60_1_mRNA_Seq_PBS   PBS 28377976         1
## SRR7451201_Rep_2_ME60_2_mRNA_Seq_EGF   EGF 31739139         1
```

```
head(DGEObj$counts)
```

```
##                SRR7451179_Rep_2_MP60 SRR7451182_Rep_1_ME60
## ENSG00000223972                     0                     0
## ENSG00000227232                    14                    24
## ENSG00000278267                     0                     0
## ENSG00000243485                     0                     0
## ENSG00000284332                     0                     0
## ENSG00000237613                     0                     0
##                SRR7451187_Rep_1_MP60 SRR7451201_Rep_2_ME60
## ENSG00000223972                     0                     0
## ENSG00000227232                    10                    22
## ENSG00000278267                     0                     0
## ENSG00000243485                     0                     0
## ENSG00000284332                     0                     0
## ENSG00000237613                     0                     0
```

```
head(DGEObj$genes)
```

```
##                GeneID Length
## ENSG00000223972 ENSG00000223972 1735
## ENSG00000227232 ENSG00000227232 1351
## ENSG00000278267 ENSG00000278267  68
## ENSG00000243485 ENSG00000243485 1021
## ENSG00000284332 ENSG00000284332  138
## ENSG00000237613 ENSG00000237613 1219
```

```
DGEObj
```

```
## An object of class "DGEList"
```

```
## $counts
```

```
##                SRR7451179_Rep_2_MP60 SRR7451182_Rep_1_ME60
## ENSG00000223972                     0                     0
## ENSG00000227232                    14                    24
## ENSG00000278267                     0                     0
## ENSG00000243485                     0                     0
## ENSG00000284332                     0                     0
##                SRR7451187_Rep_1_MP60 SRR7451201_Rep_2_ME60
## ENSG00000223972                     0                     0
## ENSG00000227232                    10                    22
## ENSG00000278267                     0                     0
## ENSG00000243485                     0                     0
## ENSG00000284332                     0                     0
```

```
## 58879 more rows ...
```

```
##
```

```
## $samples
```

```
##                               group lib.size norm.factors
```

```
## SRR7451179_Rep_2_MP60_2_mRNA_Seq_PBS    PBS 26653631      1
## SRR7451182_Rep_1_ME60_1_mRNA_Seq_EGF    EGF 30763088      1
## SRR7451187_Rep_1_MP60_1_mRNA_Seq_PBS    PBS 28377976      1
## SRR7451201_Rep_2_ME60_2_mRNA_Seq_EGF    EGF 31739139      1
##
## $genes
##                               GeneID Length
## ENSG00000223972 ENSG00000223972    1735
## ENSG00000227232 ENSG00000227232    1351
## ENSG00000278267 ENSG00000278267     68
## ENSG00000243485 ENSG00000243485    1021
## ENSG00000284332 ENSG00000284332     138
## 58879 more rows ...
```

Data Filtering

We filter the data by keeping only candidates with > 2 counts per million mapped reads AND in 3 libs.

```
keep <- rowSums(cpm(DGEObj) > 1) >= 3

# select sub set from above
DGEObj <- DGEObj[keep, keep.lib.sizes = FALSE]

dim(DGEObj)

## [1] 14691      4
```

edgeR Analysis

We calculate the normalization factor, estimate the dispersion, fit the linear model, and then summarize the results in an edgeR table format.

```
# The filtered raw counts are then normalized with calcNormFactors according to
# the weighted trimmed mean of M-values (TMM) to eliminate composition biases
# between libraries.
DGEObj <- calcNormFactors(DGEObj) # recalc norm factors - TMM

# With the normalized gene counts and design matrix we can now generate the
# negative binomial (NB) dispersion estimates using the estimateDisp function.
# The NB dispersion estimates reflect the overall biological variability under
# the Quasi-Likelihood framework in edgeR.
DGEObj <- estimateDisp(DGEObj, design) # estimate dispersion
DGEObj$common.dispersion

## [1] 0.001686058

# A generalized linear model (GLM) of the Negative Binomial family is used to
# fit the data.
fit <- glmFit(DGEObj, design) # fit generalized linear model
lrt <- glmLRT(fit, contrast = contrasts[, "EGF_vs_PBS"])

# main data table of edgeR results
edgeRTable <- topTags(lrt, n = nrow(DGEObj))$table
```

Mapping

We do mapping of the Ensembl gene ID's.

```

ensembl <- useMart("ensembl", dataset = "hsapiens_gene_ensembl", host="ensembl.org")

## Warning: Ensembl will soon enforce the use of https.
## Ensure the 'host' argument includes "https://"

genemap <- getBM(attributes = c("external_gene_name", "ensembl_gene_id", "description"),
                 values = edgeTable$GeneID, mart = ensembl)
genemap2 <- getBM(attributes = c("ensembl_gene_id", "go_id"), filter = "ensembl_gene_id",
                  values = edgeTable$GeneID, mart = ensembl)

edgeTable.idx <- match(edgeTable$GeneID, genemap$ensembl_gene_id)
edgeTable.idx2 <- match(edgeTable$GeneID, genemap2$ensembl_gene_id)

# add biomaRt data to results tables TRUE
edgeTable$external_gene_name <- genemap$external_gene_name[edgeTable.idx]
# edgeTable$entrezgene <- genemap$entrezgene[edgeTable.idx]
edgeTable$description <- genemap$description[edgeTable.idx]
edgeTable$GO <- genemap2$go_id[edgeTable.idx2]
edgeTable$ensembl_gene_id <- genemap$ensembl_gene_id[edgeTable.idx]
lrt$table$gene <- genemap$external_gene_name[edgeTable.idx]
gene_names <- data.frame(lrt$table$gene)

edgeTable <- edgeTable[c("external_gene_name", "GeneID", "Length", "logFC",
                        "logCPM", "PValue", "FDR", "GO", "description")]

ttop_dge <- edgeTable
head(ttop_dge)

```

```

##           external_gene_name      GeneID Length  logFC  logCPM
## ENSG00000125740      FOSB ENSG00000125740   5553 4.962212 6.902998
## ENSG00000138166      DUSP5 ENSG00000138166   2535 3.983957 6.291272
## ENSG00000119508      NR4A3 ENSG00000119508   6314 3.302585 8.107344
## ENSG00000175592      FOSL1 ENSG00000175592   1887 3.307155 5.967628
## ENSG00000171223      JUNB ENSG00000171223   1830 3.283124 7.112127
## ENSG00000162772      ATF3 ENSG00000162772   4040 3.119576 7.369775
##           PValue      FDR      GO
## ENSG00000125740 0.000000e+00 0.000000e+00 GO:0003677
## ENSG00000138166 0.000000e+00 0.000000e+00 GO:0016791
## ENSG00000119508 0.000000e+00 0.000000e+00 GO:0003677
## ENSG00000175592 5.605306e-296 2.058689e-292 GO:0003677
## ENSG00000171223 1.089103e-281 3.200002e-278 GO:0003677
## ENSG00000162772 9.076298e-280 2.222331e-276 GO:0003677
##
## ENSG00000125740 FosB proto-oncogene, AP-1 transcription factor subunit [Source:HGNC Symbol;Acc:HGNC:
## ENSG00000138166      dual specificity phosphatase 5 [Source:HGNC Symbol;Acc:HGNC:
## ENSG00000119508      nuclear receptor subfamily 4 group A member 3 [Source:HGNC Symbol;Acc:HGNC:
## ENSG00000175592      FOS like 1, AP-1 transcription factor subunit [Source:HGNC Symbol;Acc:HGNC:1
## ENSG00000171223 JunB proto-oncogene, AP-1 transcription factor subunit [Source:HGNC Symbol;Acc:HGNC:
## ENSG00000162772      activating transcription factor 3 [Source:HGNC Symbol;Acc:HGNC:

```