# Class 13: RNASeq Analysis

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The data for today's lab comes from a published RNA-seq experiment where airway smooth muscle cells were treated with dexamethasone, a synthetic glucocorticoid steroid with anti-inflammatory effects.

#### **Import Data**

We need two things for this analysis: counts and metadata these are called "countData" and "colData" in the DESeq2 world.

```
counts <- read.csv("airway_scaledcounts.csv", row.names=1)
metadata <- read.csv("airway_metadata.csv")</pre>
```

#### head(counts)

|                 | SRR1039508 | SRR1039509 | SRR1039512 | SRR1039513 | SRR1039516 |
|-----------------|------------|------------|------------|------------|------------|
| ENSG0000000003  | 723        | 486        | 904        | 445        | 1170       |
| ENSG0000000005  | 0          | 0          | 0          | 0          | 0          |
| ENSG00000000419 | 467        | 523        | 616        | 371        | 582        |
| ENSG00000000457 | 347        | 258        | 364        | 237        | 318        |
| ENSG00000000460 | 96         | 81         | 73         | 66         | 118        |

| ENSG00000000938 | 0          | 0          | 1          | 0 | 2 |
|-----------------|------------|------------|------------|---|---|
|                 | SRR1039517 | SRR1039520 | SRR1039521 |   |   |
| ENSG0000000003  | 1097       | 806        | 604        |   |   |
| ENSG0000000005  | 0          | 0          | 0          |   |   |
| ENSG00000000419 | 781        | 417        | 509        |   |   |
| ENSG00000000457 | 447        | 330        | 324        |   |   |
| ENSG00000000460 | 94         | 102        | 74         |   |   |
| ENSG00000000938 | 0          | 0          | 0          |   |   |

The counts are organized with a gene per row and experiment per column.

#### head(metadata)

```
id dex celltype geo_id
1 SRR1039508 control N61311 GSM1275862
2 SRR1039509 treated N61311 GSM1275863
3 SRR1039512 control N052611 GSM1275866
4 SRR1039513 treated N052611 GSM1275867
5 SRR1039516 control N080611 GSM1275870
6 SRR1039517 treated N080611 GSM1275871
```

Q1. How many genes are in this dataset?

#### nrow(counts)

#### [1] 38694

Q2. How many 'control' cell lines do we have?

```
sum(metadata$dex == "control")
```

#### [1] 4

#### table(metadata\$dex)

```
control treated 4 4
```

#### Check on match of metaData and colData

#### colnames(counts)

- [1] "SRR1039508" "SRR1039509" "SRR1039512" "SRR1039513" "SRR1039516"
- [6] "SRR1039517" "SRR1039520" "SRR1039521"

#### metadata\$id

- [1] "SRR1039508" "SRR1039509" "SRR1039512" "SRR1039513" "SRR1039516"
- [6] "SRR1039517" "SRR1039520" "SRR1039521"

```
colnames(counts) == metadata$id
```

[1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE

If you want to know that all the elements of a vector are TRUE we can sue the all() function.

```
all( c(T, T, T) )
```

[1] TRUE

```
all( colnames(counts) == metadata$id )
```

[1] TRUE

#### **Examine Data**

#### **Analysis**

I want to start by comparing "control" and "treated" columns. To do this I will find the average for each gene (row) in all "control" columns. Then I will find the average in the "treated" columns. Then I will compare them.

Let's extract all "control" columns first.

```
control.inds <- metadata$dex == "control"</pre>
```

```
control.counts <- counts[,control.inds]</pre>
```

Now find the mean count value per gene using the apply() function.

```
control.mean <- apply(control.counts, 1, mean)</pre>
```

Now do the same for the "treated" columns. i.e. find treated.mean values.

```
treated.inds <- metadata$dex == "treated"</pre>
```

```
treated.counts <- counts[,treated.inds]</pre>
```

```
treated.mean <- apply(treated.counts, 1, mean)</pre>
```

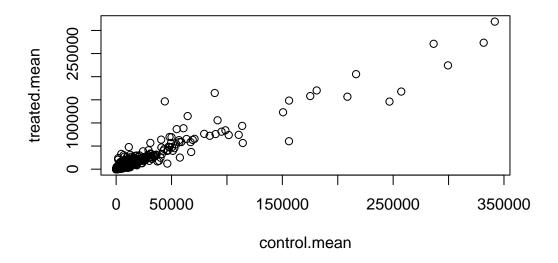
Put these two mean vector together for ease of book-keeping.

```
meancounts <- data.frame(control.mean, treated.mean)
head(meancounts)</pre>
```

|           |        | control.mean | treated.mean |
|-----------|--------|--------------|--------------|
| ENSG00000 | 000003 | 900.75       | 658.00       |
| ENSG00000 | 000005 | 0.00         | 0.00         |
| ENSG00000 | 000419 | 520.50       | 546.00       |
| ENSG00000 | 000457 | 339.75       | 316.50       |
| ENSG00000 | 000460 | 97.25        | 78.75        |
| ENSG00000 | 000938 | 0.75         | 0.00         |

Let's have a wee look with a quick plot.

```
plot(meancounts)
```



log(10, base=2)

[1] 3.321928

log2(10/10)

[1] 0

log2(20/10)

[1] 1

log2(10/20)

[1] -1

log2(40/10)

[1] 2

We most often work in log2 units because they have a more simple interpertation.

Here we calculate the log2 Fold-change of treated/control values and add it to our wee data frame of results.

```
meancounts$log2fc <- log2(meancounts$treated.mean / meancounts$control.mean)
head(meancounts)</pre>
```

|                 | ${\tt control.mean}$ | ${\tt treated.mean}$ | log2fc      |
|-----------------|----------------------|----------------------|-------------|
| ENSG0000000003  | 900.75               | 658.00               | -0.45303916 |
| ENSG0000000005  | 0.00                 | 0.00                 | NaN         |
| ENSG00000000419 | 520.50               | 546.00               | 0.06900279  |
| ENSG00000000457 | 339.75               | 316.50               | -0.10226805 |
| ENSG00000000460 | 97.25                | 78.75                | -0.30441833 |
| ENSG00000000938 | 0.75                 | 0.00                 | -Inf        |

There are some funky answers in there like NaN (Not a number) and -Inf (minus infinity) that all come because I have zero count genes in my dataset.

It is common practice to filter these zero count genes out before we go too deep.

```
to.keep.inds <- (rowSums(meancounts[,1:2] == 0) == 0)
mycounts <- meancounts[to.keep.inds, ]
head(mycounts)</pre>
```

|                 | ${\tt control.mean}$ | ${\tt treated.mean}$ | log2fc      |
|-----------------|----------------------|----------------------|-------------|
| ENSG0000000003  | 900.75               | 658.00               | -0.45303916 |
| ENSG00000000419 | 520.50               | 546.00               | 0.06900279  |
| ENSG00000000457 | 339.75               | 316.50               | -0.10226805 |
| ENSG00000000460 | 97.25                | 78.75                | -0.30441833 |
| ENSG00000000971 | 5219.00              | 6687.50              | 0.35769358  |
| ENSG00000001036 | 2327.00              | 1785.75              | -0.38194109 |

Q. How many genes do we have left after zero count filterting?

```
nrow(mycounts)
```

#### [1] 21817

A common threshold for calling a gene "up" or "down" is a  $\log 2$  fold change of +2 or -2.

Q. How many "up" regulated genes do we have?

```
sum(mycounts log2fc >= +2)
```

[1] 314

#### **DESeq** analysis

We need to do this analysis properly with our inner stats person keep happy.

```
#/ message: false
library(DESeq2)
```

Loading required package: S4Vectors

Loading required package: stats4

Loading required package: BiocGenerics

Attaching package: 'BiocGenerics'

The following objects are masked from 'package:stats':

IQR, mad, sd, var, xtabs

The following objects are masked from 'package:base':

anyDuplicated, aperm, append, as.data.frame, basename, cbind, colnames, dirname, do.call, duplicated, eval, evalq, Filter, Find, get, grep, grepl, intersect, is.unsorted, lapply, Map, mapply, match, mget, order, paste, pmax, pmax.int, pmin, pmin.int, Position, rank, rbind, Reduce, rownames, sapply, setdiff, table, tapply, union, unique, unsplit, which.max, which.min

Attaching package: 'S4Vectors'

The following object is masked from 'package:utils':
findMatches

The following objects are masked from 'package:base':

expand.grid, I, unname

Loading required package: IRanges

Loading required package: GenomicRanges

Loading required package: GenomeInfoDb

Loading required package: SummarizedExperiment

Loading required package: MatrixGenerics

Loading required package: matrixStats

Attaching package: 'MatrixGenerics'

The following objects are masked from 'package:matrixStats':

colAlls, colAnyNAs, colAnys, colAvgsPerRowSet, colCollapse, colCounts, colCummaxs, colCummins, colCumprods, colCumsums, colDiffs, colIQRDiffs, colIQRs, colLogSumExps, colMadDiffs, colMads, colMaxs, colMeans2, colMedians, colMins, colOrderStats, colProds, colQuantiles, colRanges, colRanks, colSdDiffs, colSds, colSums2, colTabulates, colVarDiffs, colVars, colWeightedMads, colWeightedMeans, colWeightedMedians, colWeightedSds, colWeightedVars, rowAlls, rowAnyNAs, rowAnys, rowAvgsPerColSet, rowCollapse, rowCounts, rowCummaxs, rowCummins, rowCumprods, rowCumsums, rowDiffs, rowIQRDiffs, rowIQRs, rowLogSumExps, rowMadDiffs, rowMads, rowMaxs, rowMeans2, rowMedians, rowMins, rowOrderStats, rowProds, rowQuantiles, rowRanges, rowRanks, rowSdDiffs, rowSds, rowSums2, rowTabulates, rowVarDiffs, rowVars, rowWeightedMads, rowWeightedMeans, rowWeightedMedians, rowWeightedSds, rowWeightedVars

```
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 object is masked from 'package:MatrixGenerics':
    rowMedians
The following objects are masked from 'package:matrixStats':
    anyMissing, rowMedians
To use DESeq we need to get our input data in very particular format.
dds <- DESeqDataSetFromMatrix(countData = counts,</pre>
                       colData = metadata,
                       design = ~dex)
converting counts to integer mode
Warning in DESeqDataSet(se, design = design, ignoreRank): some variables in
design formula are characters, converting to factors
Run DESeq analysis
dds <- DESeq(dds)
estimating size factors
estimating dispersions
```

```
mean-dispersion relationship

final dispersion estimates

fitting model and testing

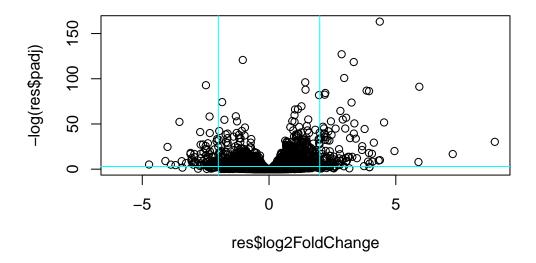
Get the results
```

```
res <- results(dds)
head(res)</pre>
```

```
log2 fold change (MLE): dex treated vs control
Wald test p-value: dex treated vs control
DataFrame with 6 rows and 6 columns
                 baseMean log2FoldChange
                                           lfcSE
                                                             pvalue
                                                      stat
                <numeric>
                              <numeric> <numeric> <numeric> <numeric>
ENSG00000000003 747.194195
                             -0.3507030 0.168246 -2.084470 0.0371175
ENSG00000000005
                 0.000000
                                              NA
                                                       NA
                                                                 NA
ENSG00000000419 520.134160
                              ENSG00000000457 322.664844
                              0.0245269 0.145145 0.168982 0.8658106
ENSG00000000460 87.682625
                             -0.1471420 0.257007 -0.572521 0.5669691
ENSG00000000938
                             -1.7322890 3.493601 -0.495846 0.6200029
                 0.319167
                   padj
               <numeric>
ENSG0000000000 0.163035
ENSG00000000005
ENSG00000000419 0.176032
ENSG00000000457 0.961694
ENSG00000000460 0.815849
ENSG00000000938
                     NA
```

I want to make a figure showing an overview of all my results to date. A plot of log2 fold change vs the p-value (adjusted p-value)

```
plot(res$log2FoldChange, -log(res$padj))
abline(v=-2, col="cyan")
abline(v=2, col="cyan")
abline(h=-log(0.05), col="cyan")
```



```
log(0.5)

[1] -0.6931472

log(0.000005)

[1] -12.20607

# Setup our custom point color vector
mycols <- rep("cyan", nrow(res))
mycols[ abs(res$log2FoldChange) > 2 ] <- "blue"

inds <- (res$padj < 0.01) & (abs(res$log2FoldChange) > 2 )
mycols[ inds ] <- "turquoise"</pre>
```

## Volcano plot with custom colors

```
plot( res$log2FoldChange, -log(res$padj),
  col=mycols, ylab="-Log(P-value)", xlab="Log2(FoldChange)" )

# Cut-off lines
abline(v=c(-2,2), col="gray", lty=2)
abline(h=-log(0.1), col="gray", lty=2)
```

