

Lab 13

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```
library(BiocManager)
```

Bioconductor version '3.18' is out-of-date; the current release version '3.19' is available with R version '4.4'; see <https://bioconductor.org/install>

```
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, sort,
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
```

Import countData and colData

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

head(counts)
```

	SRR1039508	SRR1039509	SRR1039512	SRR1039513	SRR1039516
ENSG000000000003	723	486	904	445	1170
ENSG000000000005	0	0	0	0	0
ENSG000000000419	467	523	616	371	582
ENSG000000000457	347	258	364	237	318
ENSG000000000460	96	81	73	66	118
ENSG000000000938	0	0	1	0	2

	SRR1039517	SRR1039520	SRR1039521
ENSG000000000003	1097	806	604
ENSG000000000005	0	0	0
ENSG000000000419	781	417	509
ENSG000000000457	447	330	324
ENSG000000000460	94	102	74
ENSG000000000938	0	0	0

```
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?

There are 38694 genes in this data set.

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

There are 4 control cell lines.

Extract and summarize the control samples

```
control <- metadata[metadata$dex == "control",]
control.counts <- counts[,control$id]
control.mean <- rowMeans(control.counts)
head(control.mean)
```

ENSG000000000003	ENSG000000000005	ENSG000000000419	ENSG000000000457	ENSG000000000460
900.75	0.00	520.50	339.75	97.25
ENSG000000000938				
0.75				

Extract and summarize the treated samples

```
treated <- metadata[metadata$dex == "treated",]  
treated.counts <- counts[,treated$id]  
treated.mean <- rowMeans(treated.counts)  
head(treated.mean)
```

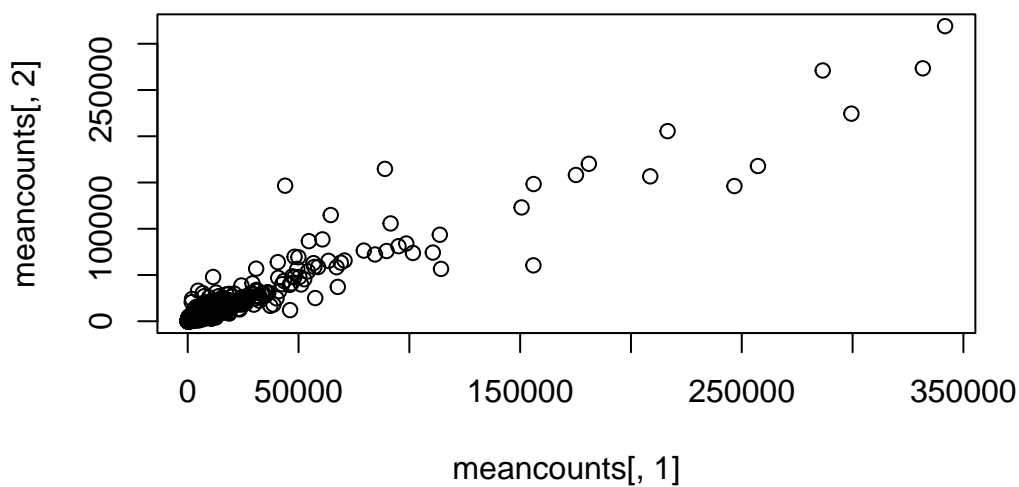
```
ENSG000000000003 ENSG000000000005 ENSG000000000419 ENSG000000000457 ENSG000000000460  
        658.00          0.00        546.00        316.50         78.75  
ENSG0000000000938  
        0.00
```

Store these results together in a dataframe called mean counts.

```
meancounts <- data.frame(control.mean, treated.mean)
```

Lets make a plot to explore the results a little.

```
plot(meancounts[,1], meancounts[,2])
```

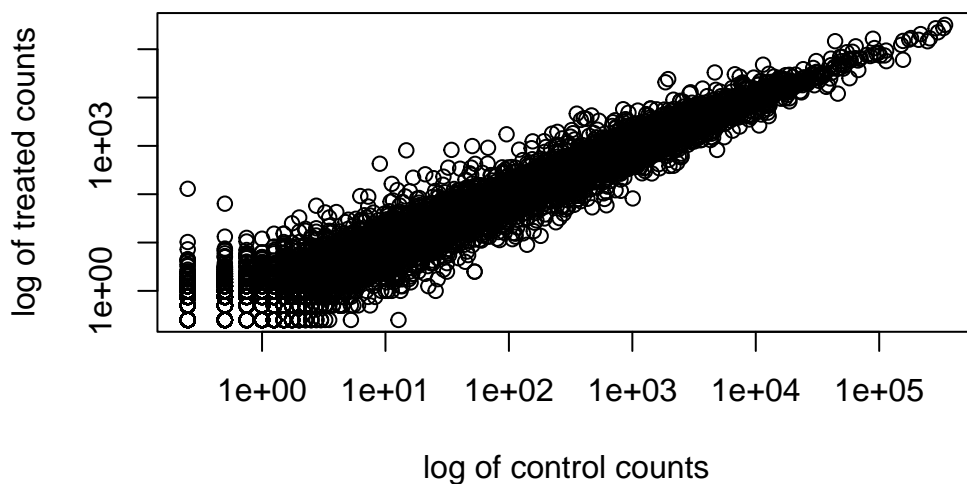


Make log-log plot to draw out this skewed data and see what is going on.

```
plot(meancounts[,1], meancounts[,2], log="xy", xlab = "log of control counts",
      ylab = "log of treated counts")
```

Warning in xy.coords(x, y, xlabel, ylabel, log): 15032 x values <= 0 omitted from logarithmic plot

Warning in xy.coords(x, y, xlabel, ylabel, log): 15281 y values <= 0 omitted from logarithmic plot



Log2 transformation has a nice property, where no change will make the log2 value zero, doubling will lead log2 to be 1 and halving will lead it to be -1.

Add log2 fold change column to our results so far.

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

# To get rid of NaN:

# says where the count is 0
zero.vals <- which(meancounts[,1:2]==0, arr.ind=TRUE)
```

```
to.rm <- unique(zero.vals[,1])
# removes genes with 0 counts
mycounts <- meancounts[-to.rm,]
head(mycounts)
```

	control.mean	treated.mean	log2fc
ENSG000000000003	900.75	658.00	-0.45303916
ENSG000000000419	520.50	546.00	0.06900279
ENSG000000000457	339.75	316.50	-0.10226805
ENSG000000000460	97.25	78.75	-0.30441833
ENSG000000000971	5219.00	6687.50	0.35769358
ENSG00000001036	2327.00	1785.75	-0.38194109

How many genes are remaining?

There are 21817 genes remaining.

Use fold change to see up and down regulated genes.

```
up.ind <- mycounts$log2fc > 2
down.ind <- mycounts$log2fc < (-2)
```

DESeq2 analysis

```
#load up DESeq2
library(DESeq2)

dds <- DESeqDataSetFromMatrix(countData=counts,
                              colData=metadata,
                              design=~dex) # design - which col to look at
```

converting counts to integer mode

Warning in DESeqDataSet(se, design = design, ignoreRank): some variables in design formula are characters, converting to factors

```
dds
```

```
class: DESeqDataSet
dim: 38694 8
metadata(1): version
assays(1): counts
rownames(38694): ENSG000000000003 ENSG000000000005 ... ENSG00000283120
               ENSG00000283123
rowData names(0):
colnames(8): SRR1039508 SRR1039509 ... SRR1039520 SRR1039521
colData names(4): id dex celltype geo_id
```

```
dds <- DESeq(dds)
```

estimating size factors

estimating dispersions

gene-wise dispersion estimates

mean-dispersion relationship

final dispersion estimates

fitting model and testing

```
res <- results(dds)
```

```
res
```

log2 fold change (MLE): dex treated vs control

Wald test p-value: dex treated vs control

DataFrame with 38694 rows and 6 columns

	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG000000000003	747.1942	-0.3507030	0.168246	-2.084470	0.0371175
ENSG000000000005	0.0000	NA	NA	NA	NA


```

ENSG00000000419  520.1342      0.2061078  0.101059  2.039475  0.0414026
ENSG00000000457  322.6648      0.0245269  0.145145  0.168982  0.8658106
ENSG00000000460   87.6826     -0.1471420  0.257007 -0.572521  0.5669691
...
ENSG00000283115  0.000000      NA          NA          NA          NA
ENSG00000283116  0.000000      NA          NA          NA          NA
ENSG00000283119  0.000000      NA          NA          NA          NA
ENSG00000283120  0.974916     -0.668258  1.69456  -0.394354  0.693319
ENSG00000283123  0.000000      NA          NA          NA          NA
      padj
      <numeric>
ENSG00000000003  0.163035
ENSG00000000005      NA
ENSG00000000419  0.176032
ENSG00000000457  0.961694
ENSG00000000460  0.815849
...
ENSG00000283115      NA
ENSG00000283116      NA
ENSG00000283119      NA
ENSG00000283120      NA
ENSG00000283123      NA

```

We can get some basic summary tallies using the `summary()` function.

```
summary(res, alpha = 0.05)
```

```

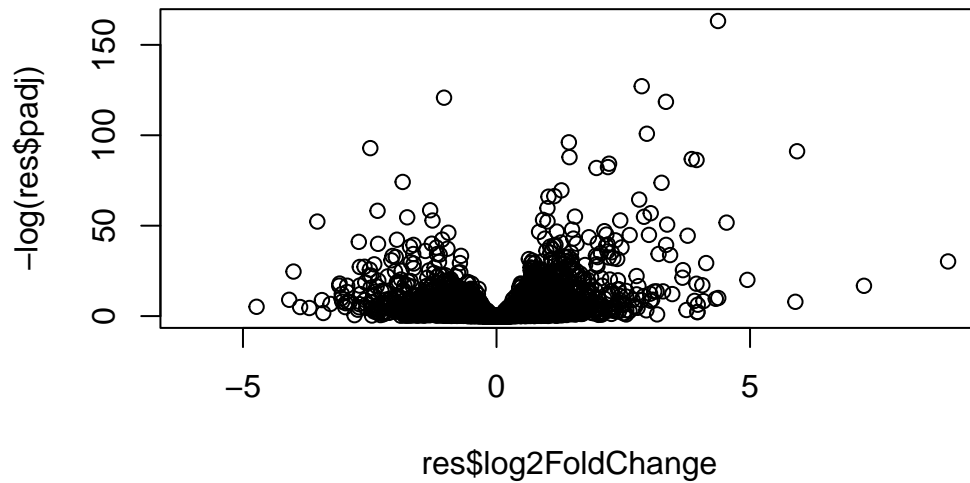
out of 25258 with nonzero total read count
adjusted p-value < 0.05
LFC > 0 (up)      : 1242, 4.9%
LFC < 0 (down)    : 939, 3.7%
outliers [1]      : 142, 0.56%
low counts [2]    : 9971, 39%
(mean count < 10)
[1] see 'cooksCutoff' argument of ?results
[2] see 'independentFiltering' argument of ?results

```

Volcano Plot

Let's make a summary plot of our results.

```
plot(res$log2FoldChange, -log(res$padj))
```

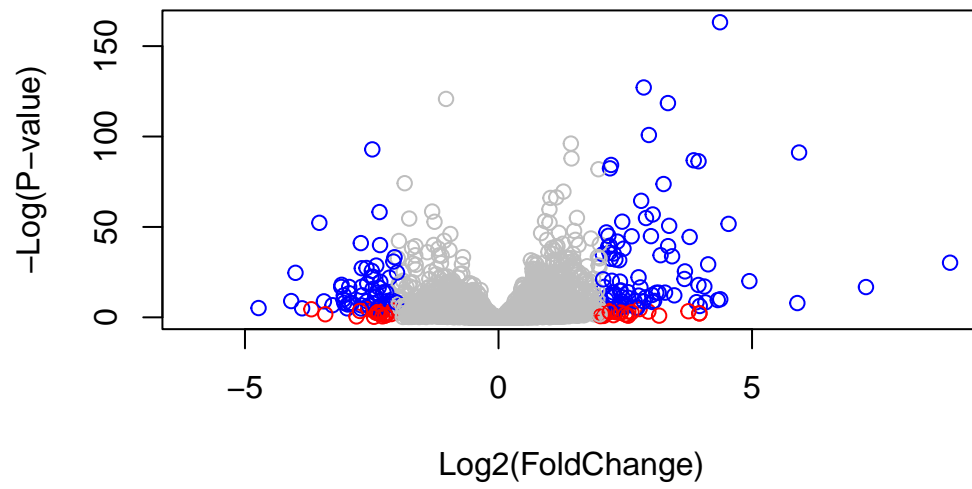


Let's add colors:

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

inds <- (res$padj < 0.01) & (abs(res$log2FoldChange) > 2 )
mycols[ inds ] <- "blue"

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



Finish for today by saving our results.

```
# write.csv(res, file = "DESeq2_results.csv")
```