Lab 12

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library(BiocManager) 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, saveRDS, 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, rowWeightedMedians, rowWeightedMedians, rowWeightedMedians, 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

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

[1] 38694

head(counts)

nrow(counts)

	SRR1039508	SRR1039509	SRR1039512	SRR1039513	SRR1039516
ENSG00000000003	723	486	904	445	1170
ENSG00000000005	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		
ENSG00000000003	1097	806	604		
ENSG0000000005	0	0	0		
ENSG00000000419	781	417	509		
ENSG00000000457	447	330	324		
ENSG00000000460	94	102	74		
ENSG00000000938	0	0	0		

View(metadata)

Q1. How many genes are in this dataset?

There are 38694 genes

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

[1] 4

```
table(metadata$dex)
```

```
control treated 4 4
```

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

There are 4 control cell lines

Toy differential expression analysis

Calculate the mean per gene count values for all "control" samples (i.e columns in counts) and do the same for "treated" and then compare them.

1. Find all "control values/columns in counts

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

```
FALSE TRUE 4 4
```

Q3. How would you make the above code in either approach more robust? Is there a function that could help here?

```
control.inds <- metadata$dex == "control"
control.counts <- counts[,control.inds]</pre>
```

2. Find the mean per gene across all control columns.

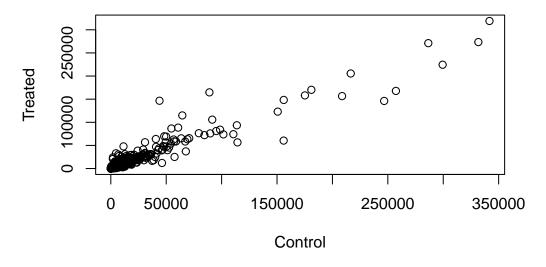
```
control.mean <- apply(control.counts, 1, mean)</pre>
```

Q4. Follow the same procedure for the treated samples (i.e. calculate the mean per gene across drug treated samples and assign to a labeled vector called treated mean)

```
treated.inds <- metadata$dex == "treated"
treated.counts <- counts[,treated.inds]
treated.mean <- apply(treated.counts, 1, mean)</pre>
```

Q5 (a). Create a scatter plot showing the mean of the treated samples against the mean of the control samples. Your plot should look something like the following.

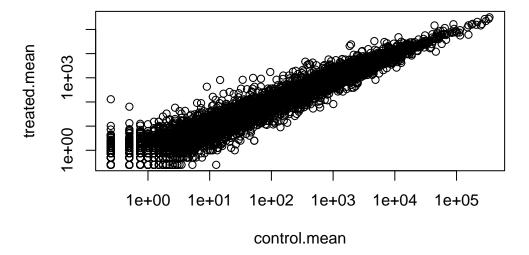
```
meancounts <- data.frame(control.mean, treated.mean)
plot(meancounts[,1],meancounts[,2], xlab="Control", ylab="Treated")</pre>
```



```
plot(meancounts, log='xy')
```

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



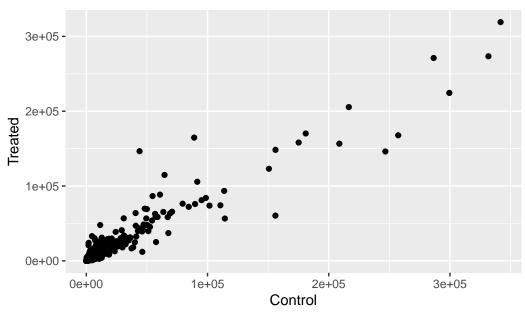
Q5 (b). You could also use the ggplot2 package to make this figure producing the plot below. What geom_?() function would you use for this plot?

point

```
library(ggplot2)

ggplot(meancounts, aes(x = control.mean, y = treated.mean)) +
    geom_point() +
    xlab("Control") +
    ylab("Treated") +
    ggtitle("Control vs Treated Mean Counts")
```

Control vs Treated Mean Counts



Q6. Try plotting both axes on a log scale. What is the argument to plot() that allows you to do this?

plot(meancounts, log='xy')

We most frequently use log2 transformations for this type of data

log2(10/10)

[1] 0

log2(20/10)

[1] 1

log2(30/10)

[1] 1.584963

log2(40/10)

[1] 2

log2(10/20)

[1] -1

These $\log 2$ values make the interpretation of "fold-change" a little easier and a rule-of-thumb in the field is a $\log 2$ fold-change of +2 or -2 is where we start to pay attention

```
log2(40/10)
```

[1] 2

Lets calculate the log2 (fold-change) and add it to our meancounts data.frame

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

	${\tt control.mean}$	${\tt treated.mean}$	log2fc
ENSG00000000003	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

Q7. What is the purpose of the arr.ind argument in the which() function call above? Why would we then take the first column of the output and need to call the unique() function?

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

Q. how many genes do I have left after this zero count filtering

nrow(mycounts)

[1] 21817

Q. How many genes are "up" regulated upon drug treatment with a threshold of +2 log2-fold-change?

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

[1] 250

250 up-regulated genes

Q. How many genes are "down" regulated upon drug treatment with a threshold of -2 $\log 2$ -fold-change?

```
down.ind <- mycounts$log2fc < (-2)
count(down.ind)</pre>
```

[1] 367

367 down regulated genes