

Class12

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Background

Today we will analyze some RNASeq data from Himes et al. on the effects of a common steroid (dexmethylasone also called “dex”) on airway smooth muscle cells (ASMs).

For this analysis we need two main inputs

- `countData`: a table of **counts** per gene (in rows) across experiments (in columns).
- `-colData`: **metadata** about the design of the experiments. The rows match the columns in `countData`

Data Import

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

Let's have a peek at our `counts` data.

```
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		

and the metadata

```
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 experiments (i.e. columns in counts or rows in metadata) do we have?

```
ncol(counts)
```

```
[1] 8
```

Q3. How many ‘control’ experiments do we have?

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

```
[1] 4
```

1. Extract the “control” columns from `counts`
2. Find the mean value for each gene in these “control” columns 3-4. Do the same for the “treated” columns
3. Compare these values for each gene

```
control inds <- metadata$dex=="control"  
control counts <- counts[,control inds]
```

Step2.

```
control mean <- rowMeans(control counts)
```

Step3-4

```
treated inds <- metadata$dex=="treated"  
treated counts <- counts[,treated inds]
```

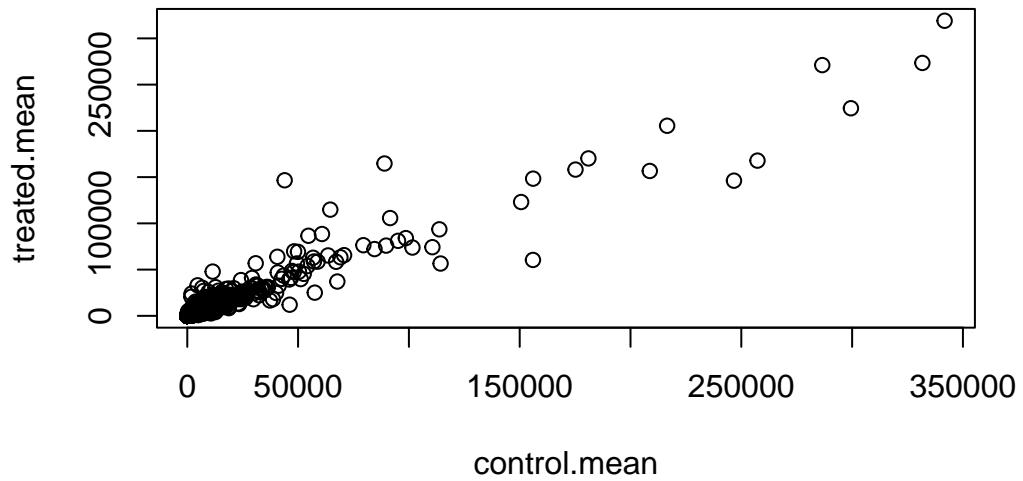
```
treated mean <- rowMeans(treated counts)
```

For ease of book-keeping we can store these together in one data frame called `meancounts`

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

	control mean	treated mean
ENSG000000000003	900.75	658.00
ENSG000000000005	0.00	0.00
ENSG000000000419	520.50	546.00
ENSG000000000457	339.75	316.50
ENSG000000000460	97.25	78.75
ENSG000000000938	0.75	0.00

```
plot(meancounts)
```

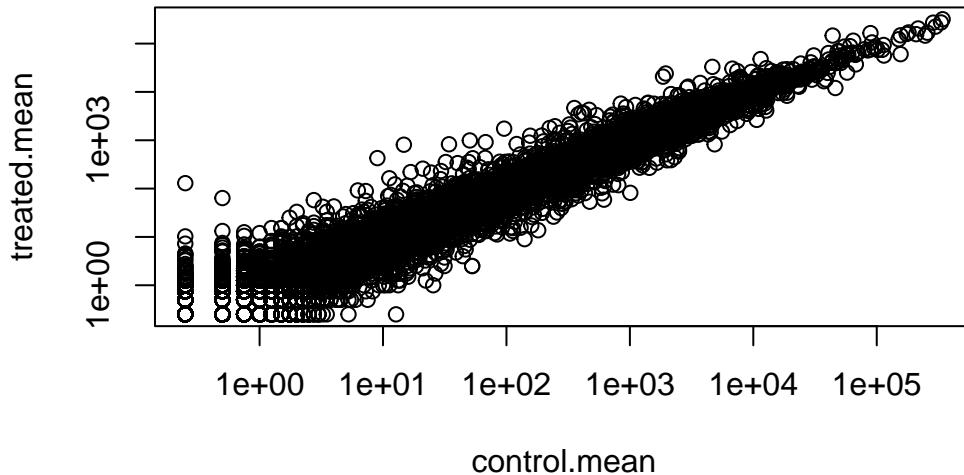


This is screaming at me to log transform this data!!

```
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



We use log2 “fold-change” as a way to compare

```
#treated/control
log2(10/10) #no change
```

```
[1] 0
```

```
log2(20/10)
```

```
[1] 1
```

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

	control.mean	treated.mean	log2fc
ENSG000000000003	900.75	658.00	-0.45303916
ENSG000000000005	0.00	0.00	NaN
ENSG000000000419	520.50	546.00	0.06900279
ENSG000000000457	339.75	316.50	-0.10226805
ENSG000000000460	97.25	78.75	-0.30441833
ENSG000000000938	0.75	0.00	-Inf

2 folds means 4 times higher gene number A common “rule-of-thumb” threshold for calling something “up” regulated is a log2-fold-change of +2 or greater. For “down” regulated is -2 or less.

```
zero.ind <- which(meancounts[,1:2]==0, arr.ind=T)

to.rm <- unique(zero.ind[,1])
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
ENSG000000001036	2327.00	1785.75	-0.38194109

```
# ALTERNATE METHOD FIND ZERO VALUES
```

Q How many genes are “up” regulated at the +2 log2FC threshold?

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

```
sum(up.ind)
```

[1] 250

Q How many genes are “down” regulated at the +2 log2FC threshold?

```
sum(down.ind)
```

[1] 367

PCA

```
library(DESeq2)
```

Loading required package: S4Vectors

Loading required package: stats4

Loading required package: BiocGenerics

Loading required package: generics

Attaching package: 'generics'

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

```
as.difftime, as.factor, as.ordered, intersect, is.element, setdiff,  
setequal, union
```

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, is.unsorted, lapply, Map, mapply, match, mget,  
order, paste, pmax, pmax.int, pmin, pmin.int, Position, rank,  
rbind, Reduce, rownames, sapply, saveRDS, table, tapply, 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: Seqinfo
```

```
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
```

```
# 1) Read data  
counts <- read.csv("airway_scaledcounts.csv", row.names = 1, check.names = FALSE)  
metadata <- read.csv("airway_metadata.csv", stringsAsFactors = FALSE)  
  
# 2) Clean whitespace/case everywhere  
colnames(counts) <- trimws(colnames(counts))  
names(metadata) <- trimws(names(metadata))  
metadata[] <- lapply(metadata, function(x) if (is.character(x)) trimws(x) else x)  
  
# 3) Set sample IDs as rownames for metadata (adjust 'run' if your id column is named differently)  
id_col <- if ("run" %in% names(metadata)) "run" else names(metadata)[1]  
rownames(metadata) <- metadata[[id_col]]  
  
# 4) Align samples (keep shared samples, consistent order)  
common <- intersect(colnames(counts), rownames(metadata))  
counts <- counts[, common, drop = FALSE]  
metadata <- metadata[common, , drop = FALSE]  
  
# 5) Normalize the 'dex' column values and inspect  
metadata$dex <- tolower(trimws(metadata$dex))  
table(metadata$dex, useNA = "ifany") # should show only 'control' and 'treated' and no <NA>
```

```

control treated
      4      4

# If you still see NA, show the offending rows:
which(is.na(metadata$dex))           # indices with NA

integer(0)

metadata[is.na(metadata$dex), , drop = FALSE]

[1] id      dex      celltype geo_id
<0 rows> (or 0-length row.names)

# 6) Drop samples with missing dex OR fix the values
keep <- !is.na(metadata$dex) & metadata$dex %in% c("control","treated")
counts  <- counts[, keep, drop = FALSE]
metadata <- metadata[keep, , drop = FALSE]

# 7) Make factor with correct levels (control as reference)
metadata$dex <- factor(metadata$dex, levels = c("control","treated"))

# 8) Build DESeq2 object (ensure integer counts)
counts_mat <- as.matrix(round(counts))
dds <- DESeqDataSetFromMatrix(countData = counts_mat,
                               colData   = metadata,
                               design    = ~ dex)

```

converting counts to integer mode

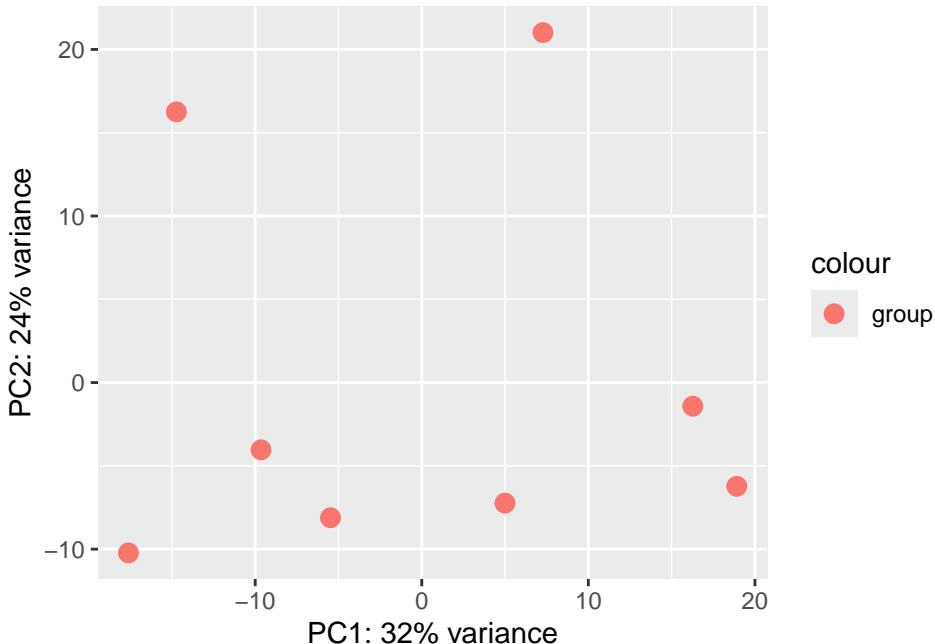
```

# Optional: filter zero-count genes
dds <- dds[rowSums(counts(dds)) > 0, ]

# 9) PCA prep and plot
vsd <- vst(dds, blind = FALSE)
plotPCA(vsd, intgroup = "dex")

```

using ntop=500 top features by variance



```
pcaData <- plotPCA(vsd, intgroup=c("dex"), returnData=TRUE)
```

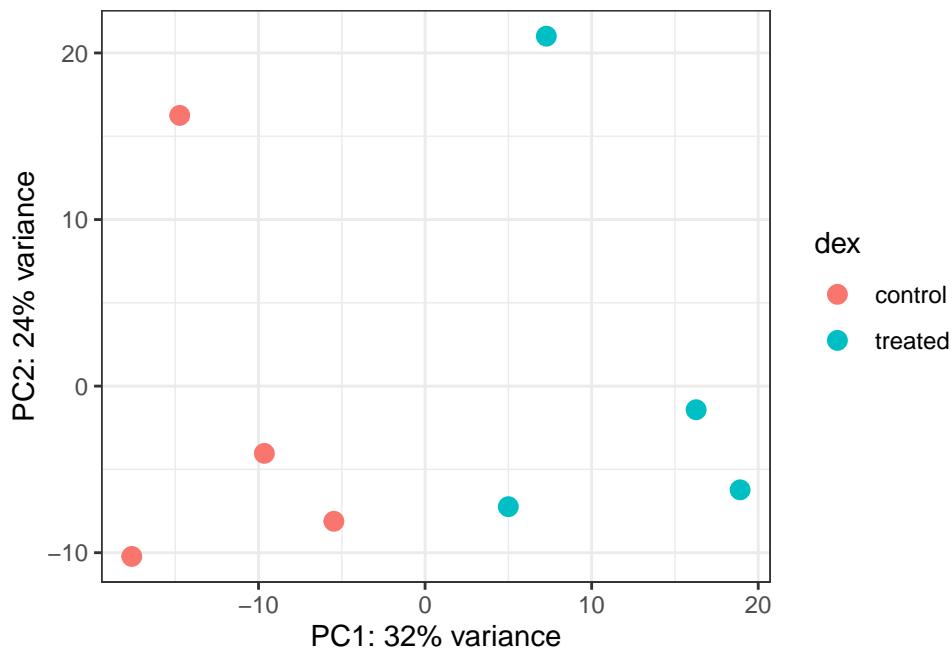
using ntop=500 top features by variance

```
head(pcaData)
```

	PC1	PC2	group	name	id	dex	celltype
SRR1039508	-17.607922	-10.225252	control	SRR1039508	SRR1039508	control	N61311
SRR1039509	4.996738	-7.238117	treated	SRR1039509	SRR1039509	treated	N61311
SRR1039512	-5.474456	-8.113993	control	SRR1039512	SRR1039512	control	N052611
SRR1039513	18.912974	-6.226041	treated	SRR1039513	SRR1039513	treated	N052611
SRR1039516	-14.729173	16.252000	control	SRR1039516	SRR1039516	control	N080611
SRR1039517	7.279863	21.008034	treated	SRR1039517	SRR1039517	treated	N080611
	geo_id	sizeFactor					
SRR1039508	GSM1275862	1.0193796					
SRR1039509	GSM1275863	0.9005653					
SRR1039512	GSM1275866	1.1784239					
SRR1039513	GSM1275867	0.6709854					
SRR1039516	GSM1275870	1.1731984					
SRR1039517	GSM1275871	1.3929361					

```
# Calculate percent variance per PC for the plot axis labels
percentVar <- round(100 * attr(pcaData, "percentVar"))
```

```
library(ggplot2)
ggplot(pcaData) +
  aes(x = PC1, y = PC2, color = dex) +
  geom_point(size = 3) +
  xlab(paste0("PC1: ", percentVar[1], "% variance")) +
  ylab(paste0("PC2: ", percentVar[2], "% variance")) +
  coord_fixed() +
  theme_bw()
```



DESeq analysis

Let's do this with DESeq2 and put some stats behind these numbers

```
library(DESeq2)
```

DESeq wants 3 things for analysis, countData, colData and desing.

```
dds <- DESeqDataSetFromMatrix(countData=counts, colData = metadata, design= ~dex)
```

converting counts to integer mode

The main function in the DESeq package to run analysis is called `DESeq()`.

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

estimating size factors

estimating dispersions

gene-wise dispersion estimates

mean-dispersion relationship

final dispersion estimates

fitting model and testing

get the results out of this DESeq object with the function `results()`

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

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	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG00000000003	747.194195	-0.350703	0.168242	-2.084514	0.0371134
ENSG00000000005	0.000000	NA	NA	NA	NA
ENSG00000000419	520.134160	0.206107	0.101042	2.039828	0.0413675
ENSG00000000457	322.664844	0.024527	0.145134	0.168996	0.8658000
ENSG00000000460	87.682625	-0.147143	0.256995	-0.572550	0.5669497
ENSG00000000938	0.319167	-1.732289	3.493601	-0.495846	0.6200029

padj
<numeric>

```

ENSG000000000003  0.163017
ENSG000000000005      NA
ENSG000000000419  0.175937
ENSG000000000457  0.961682
ENSG000000000460  0.815805
ENSG000000000938      NA

```

Volcano Plot

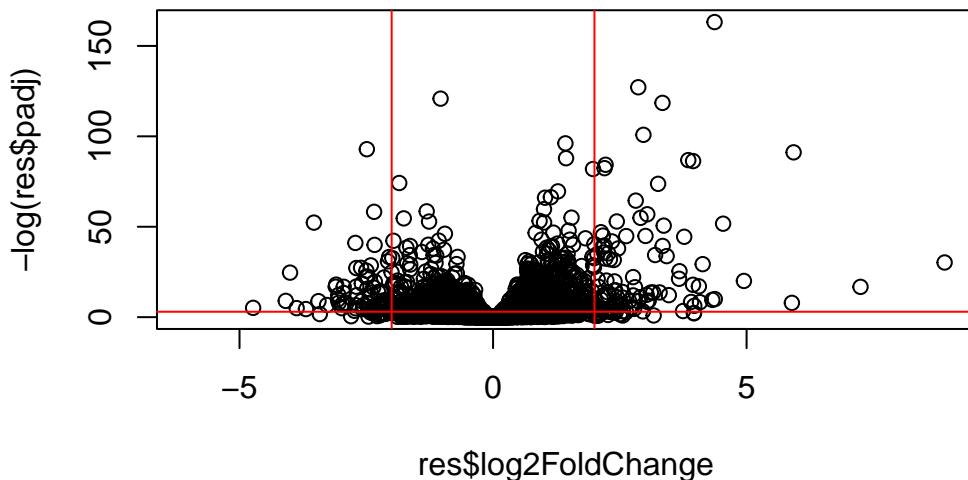
This is plot of log2FC vs adjusted p-value

```

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

abline(v=c(-2,2), col="red")
abline(h=-log(0.05), col="red")

```



A nicer ggplot volcano plot

```

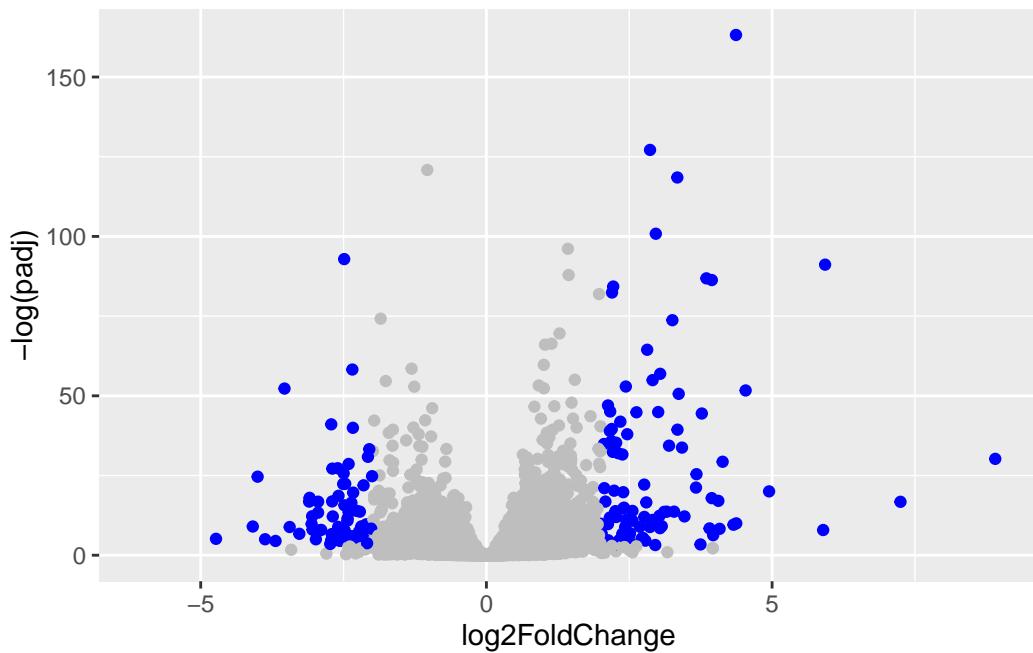
library(ggplot2)

mycols <- rep("gray", nrow(res))
mycols[abs(res$log2FoldChange)>2] <- "blue"
mycols[res$padj>=0.05] <- "gray"

ggplot(res)+ aes(log2FoldChange, -log(padj))+ geom_point(col=mycols)

```

Warning: Removed 23549 rows containing missing values or values outside the scale range
(`geom_point()`).



Save our results

```
write.csv(res,file="myresults.csv")
```

```
library(pathview)
```

```
#####
# Pathview is an open source software package distributed under GNU General
# Public License version 3 (GPLv3). Details of GPLv3 is available at
# http://www.gnu.org/licenses/gpl-3.0.html. Particullary, users are required to
# formally cite the original Pathview paper (not just mention it) in publications
# or products. For details, do citation("pathview") within R.
```

The pathview downloads and uses KEGG data. Non-academic uses may require a KEGG license agreement (details at <http://www.kegg.jp/kegg/legal.html>).

```
#####
```

```
library(gage)
```

```
library(gageData)

data(kegg.sets.hs)

# Examine the first 2 pathways in this kegg set for humans
head(kegg.sets.hs, 2)
```

```
$`hsa00232 Caffeine metabolism`
[1] "10"    "1544"  "1548"  "1549"  "1553"  "7498"  "9"

$`hsa00983 Drug metabolism - other enzymes`
[1] "10"    "1066"  "10720" "10941" "151531" "1548"  "1549"  "1551"
[9] "1553"  "1576"  "1577"  "1806"  "1807"   "1890"  "221223" "2990"
[17] "3251"  "3614"  "3615"  "3704"  "51733"  "54490" "54575"  "54576"
[25] "54577" "54578" "54579" "54600" "54657"  "54658" "54659"  "54963"
[33] "574537" "64816" "7083"  "7084"  "7172"   "7363"  "7364"   "7365"
[41] "7366"  "7367"  "7371"  "7372"  "7378"  "7498"  "79799" "83549"
[49] "8824"  "8833"  "9"     "978"
```

```
foldchanges = res$log2FoldChange
names(foldchanges) = res$entrez
head(foldchanges)
```

```
[1] -0.35070296      NA  0.20610728  0.02452701 -0.14714263 -1.73228897
```

```

# Get the results
keggres = gage(foldchanges, gsets=kegg.sets.hs)

attributes(keggres)

$names
[1] "greater" "less"      "stats"

# Look at the first three down (less) pathways
head(keggres$less, 3)

```

	p.geomean	stat.mean	p.val	q.val
hsa00232 Caffeine metabolism	NA	NaN	NA	NA
hsa00983 Drug metabolism - other enzymes	NA	NaN	NA	NA
hsa01100 Metabolic pathways	NA	NaN	NA	NA
	set.size	exp1		
hsa00232 Caffeine metabolism	0	NA		
hsa00983 Drug metabolism - other enzymes	0	NA		
hsa01100 Metabolic pathways	0	NA		

```
pathview(gene.data=foldchanges, pathway.id="hsa05310")
```

Warning: None of the genes or compounds mapped to the pathway!
 Argument gene.idtype or cpd.idtype may be wrong.

'select()' returned 1:1 mapping between keys and columns

Info: Working in directory /Users/ebruy/Desktop/UCSD/Bioinformatics 213/class12.

Info: Writing image file hsa05310.pathview.png

```
# A different PDF based output of the same data
pathview(gene.data=foldchanges, pathway.id="hsa05310", kegg.native=FALSE)
```

Warning: None of the genes or compounds mapped to the pathway!
 Argument gene.idtype or cpd.idtype may be wrong.

'select()' returned 1:1 mapping between keys and columns

Info: Working in directory /Users/ebruy/Desktop/UCSD/Bioinformatics 213/class12.

Info: Writing image file hsa05310.pathview.pdf

