

Class 14 Lab

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Background

The data for for hands-on session comes from GEO entry: GSE37704, which is associated with the following publication:

Trapnell C, Hendrickson DG, Sauvageau M, Goff L et al. “Differential analysis of gene regulation at transcript resolution with RNA-seq”. Nat Biotechnol 2013 Jan;31(1):46-53. PMID:

23222703 The authors report on differential analysis of lung fibroblasts in response to loss of the developmental transcription factor HOXA1. Their results and others indicate that HOXA1 is required for lung fibroblast and HeLa cell cycle progression. In particular their analysis show that “loss of HOXA1 results in significant expression level changes in thousands of individual transcripts, along with isoform switching events in key regulators of the cell cycle”. For our session we have used their Sailfish gene-level estimated counts and hence are restricted to protein-coding genes only.

Data Import

Reading in the counts and the metadata

```
counts <- read.csv("GSE37704_featurecounts.csv", row.names = 1)
metadata <- read.csv("GSE37704_metadata.csv")
```

Tidy and verify data

Q. How many genes are in this dataset?

```
head(metadata)
```

	id	condition
1	SRR493366	control_sirna
2	SRR493367	control_sirna
3	SRR493368	control_sirna
4	SRR493369	hoxa1_kd
5	SRR493370	hoxa1_kd
6	SRR493371	hoxa1_kd

```
nrow(counts)
```

```
[1] 19808
```

```
#19808
```

Q. How many control and knowndown experiemetns are there?

```
head(metadata)
```

```
      id      condition
1 SRR493366 control_sirna
2 SRR493367 control_sirna
3 SRR493368 control_sirna
4 SRR493369      hoxa1_kd
5 SRR493370      hoxa1_kd
6 SRR493371      hoxa1_kd
```

```
#3 control and 3 knockdown
```

Q. Does the metadata match the countdata

```
head(counts)
```

	length	SRR493366	SRR493367	SRR493368	SRR493369	SRR493370
ENSG00000186092	918	0	0	0	0	0
ENSG00000279928	718	0	0	0	0	0
ENSG00000279457	1982	23	28	29	29	28
ENSG00000278566	939	0	0	0	0	0
ENSG00000273547	939	0	0	0	0	0
ENSG00000187634	3214	124	123	205	207	212

	SRR493371
ENSG00000186092	0
ENSG00000279928	0
ENSG00000279457	46
ENSG00000278566	0
ENSG00000273547	0
ENSG00000187634	258

```
#Need to get rid of length since we only need 3 ctrl and 3 kd
```

```
metadata$id
```

```
[1] "SRR493366" "SRR493367" "SRR493368" "SRR493369" "SRR493370" "SRR493371"
```

```
newcounts <- counts[,-1]
dim(newcounts)
```

```
[1] 19808      6
```

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

```
[1] TRUE TRUE TRUE TRUE TRUE TRUE
```

Remove zero count genes

```
head(newcounts)
```

	SRR493366	SRR493367	SRR493368	SRR493369	SRR493370	SRR493371
ENSG00000186092	0	0	0	0	0	0
ENSG00000279928	0	0	0	0	0	0
ENSG00000279457	23	28	29	29	28	46
ENSG00000278566	0	0	0	0	0	0
ENSG00000273547	0	0	0	0	0	0
ENSG00000187634	124	123	205	207	212	258

```
#newcounts==0
```

```
##Removing all zero from data file  
to.keep <- rowSums(newcounts) !=0  
countData <- newcounts[to.keep,]
```

PCA quality control

We can use `prcomp()` function.

```
pc <- prcomp(t(countData), scale = T)  
summary(pc)
```

Importance of components:

	PC1	PC2	PC3	PC4	PC5	PC6
Standard deviation	87.7211	73.3196	32.89604	31.15094	29.18417	7.373e-13
Proportion of Variance	0.4817	0.3365	0.06774	0.06074	0.05332	0.000e+00
Cumulative Proportion	0.4817	0.8182	0.88594	0.94668	1.00000	1.000e+00

Color by “control” (blue) and kd (red)

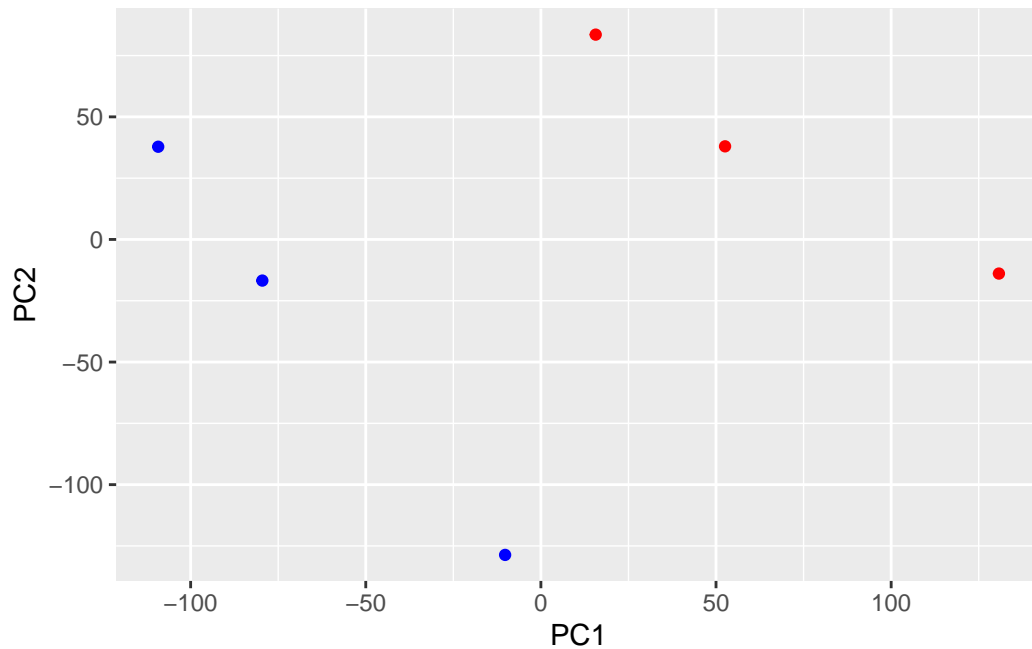
```
metadata$condition
```

```
[1] "control_sirna" "control_sirna" "control_sirna" "hoxa1_kd"  
[5] "hoxa1_kd"      "hoxa1_kd"
```

```
mycols <- c(rep("blue", 3), rep("red", 3))  
mycols
```

```
[1] "blue" "blue" "blue" "red"  "red"  "red"
```

```
library(ggplot2)  
ggplot(pc$x) +  
  aes(PC1, PC2) +  
  geom_point(col=mycols)
```



Q. How many genes do we have left after filtering?

```
nrow(countData)
```

```
[1] 15975
```

DESeq analysis

```
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

Attaching package: 'IRanges'

The following object is masked from 'package:grDevices':

windows

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
```

```
head(countData)
```

	SRR493366	SRR493367	SRR493368	SRR493369	SRR493370	SRR493371
ENSG00000279457	23	28	29	29	28	46
ENSG00000187634	124	123	205	207	212	258
ENSG00000188976	1637	1831	2383	1226	1326	1504
ENSG00000187961	120	153	180	236	255	357
ENSG00000187583	24	48	65	44	48	64
ENSG00000187642	4	9	16	14	16	16

Setup the DESeq input objects

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


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

Run DESeq

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

estimating size factors

estimating dispersions

gene-wise dispersion estimates

mean-dispersion relationship

final dispersion estimates

fitting model and testing

Extract results

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

```
head(res)
```

log2 fold change (MLE): condition hoxa1 kd vs control sirna

Wald test p-value: condition hoxa1 kd vs control sirna

DataFrame with 6 rows and 6 columns

	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG00000279457	29.9136	0.1792571	0.3248216	0.551863	5.81042e-01
ENSG00000187634	183.2296	0.4264571	0.1402658	3.040350	2.36304e-03
ENSG00000188976	1651.1881	-0.6927205	0.0548465	-12.630158	1.43990e-36
ENSG00000187961	209.6379	0.7297556	0.1318599	5.534326	3.12428e-08
ENSG00000187583	47.2551	0.0405765	0.2718928	0.149237	8.81366e-01
ENSG00000187642	11.9798	0.5428105	0.5215598	1.040744	2.97994e-01

```

                padj
                <numeric>
ENSG00000279457 6.86555e-01
ENSG00000187634 5.15718e-03
ENSG00000188976 1.76549e-35
ENSG00000187961 1.13413e-07
ENSG00000187583 9.19031e-01
ENSG00000187642 4.03379e-01

```

Volcano plot

A plot of log2 fold change vs -log of adjusted p-value with custom colors

```

mycols <- rep("grey", nrow(res))
mycols[res$log2FoldChange >= +2] <- "blue"
mycols[res$log2FoldChange <=-2] <- "blue"
mycols[res$padj >= 0.005] <- "grey"

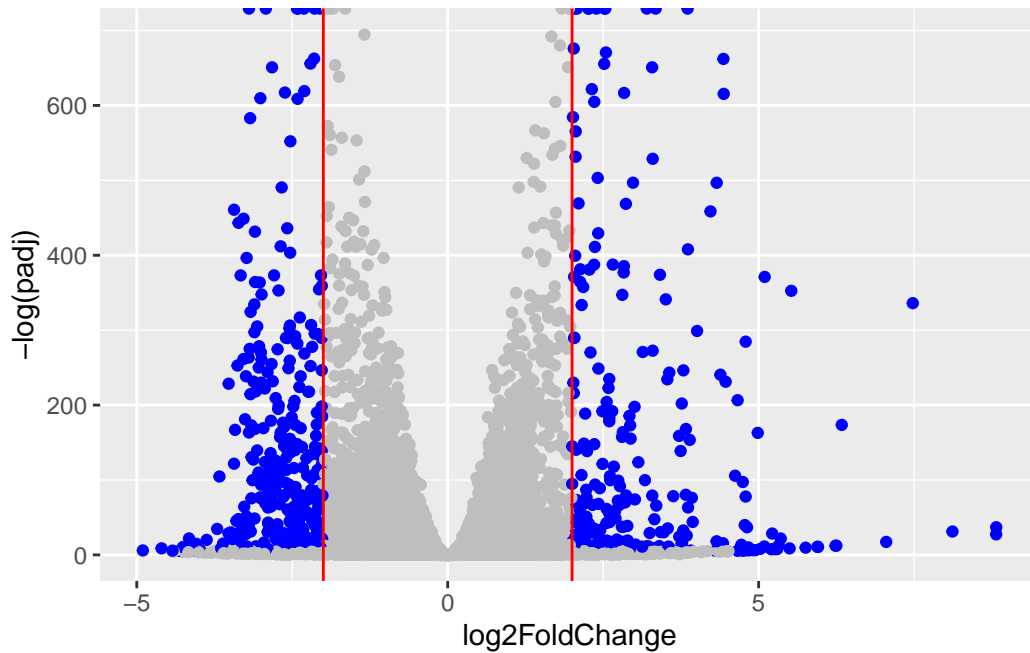
```

```

ggplot(res) +
  aes(log2FoldChange, -log(padj)) +
  geom_point(col=mycols) +
  geom_vline(xintercept = c(-2,2), col="red")

```

Warning: Removed 1237 rows containing missing values or values outside the scale range (`geom_point()`).



```
geom_hline(yintercept = -log(0.005), col="blue")
```

```
mapping: yintercept = ~yintercept
geom_hline: na.rm = FALSE
stat_identity: na.rm = FALSE
position_identity
```

Add gene annotations

We want to add gene SYMBOL and ENTREZID values to our results object.

```
library(AnnotationDbi)
library(org.Hs.eg.db)
```

```
columns(org.Hs.eg.db)
```

```

[1] "ACCNUM"      "ALIAS"      "ENSEMBL"    "ENSEMBLPROT" "ENSEMBLTRANS"
[6] "ENTREZID"    "ENZYME"     "EVIDENCE"    "EVIDENCEALL"  "GENENAME"
[11] "GENETYPE"    "GO"         "GOALL"      "IPI"          "MAP"
[16] "OMIM"        "ONTOLOGY"   "ONTOLOGYALL" "PATH"         "PFAM"
[21] "PMID"        "PROSITE"    "REFSEQ"     "SYMBOL"       "UCSCKG"
[26] "UNIPROT"

```

```

res$symbol <- mapIds(org.Hs.eg.db,
  keys=rownames(res),
  keytype = "ENSEMBL",
  column = "SYMBOL")

```

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

```

res$entrez <- mapIds(org.Hs.eg.db,
  keys=rownames(res),
  keytype = "ENSEMBL",
  column = "ENTREZID")

```

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

```
head(res)
```

log2 fold change (MLE): condition hoxa1 kd vs control sirna

Wald test p-value: condition hoxa1 kd vs control sirna

DataFrame with 6 rows and 8 columns

	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG00000279457	29.9136	0.1792571	0.3248216	0.551863	5.81042e-01
ENSG00000187634	183.2296	0.4264571	0.1402658	3.040350	2.36304e-03
ENSG00000188976	1651.1881	-0.6927205	0.0548465	-12.630158	1.43990e-36
ENSG00000187961	209.6379	0.7297556	0.1318599	5.534326	3.12428e-08
ENSG00000187583	47.2551	0.0405765	0.2718928	0.149237	8.81366e-01
ENSG00000187642	11.9798	0.5428105	0.5215598	1.040744	2.97994e-01
	padj	symbol	entrez		
	<numeric>	<character>	<character>		
ENSG00000279457	6.86555e-01	NA	NA		
ENSG00000187634	5.15718e-03	SAMD11	148398		
ENSG00000188976	1.76549e-35	NOC2L	26155		
ENSG00000187961	1.13413e-07	KLHL17	339451		
ENSG00000187583	9.19031e-01	PLEKHN1	84069		
ENSG00000187642	4.03379e-01	PERM1	84808		

Save results

```
write.csv(res, file="myresultsclass14.csv")
```

Pathway analysis

```
#|message: false  
library(gage)
```

```
library(gageData)  
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>).

```
#####
```

KEGG

```
data("kegg.sets.hs")
```

```
head(kegg.sets.hs, 1)
```

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

Make an input vector for `gage()` called `foldchanges` that has `names()` attribute set to ENTREZIDs

```
foldchanges <- res$log2FoldChange
names(foldchanges) <- res$entrez
```

```
keggres <- gage(foldchanges, gsets = kegg.sets.hs)
```

```
attributes(keggres)
```

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

```
head(keggres$less)
```

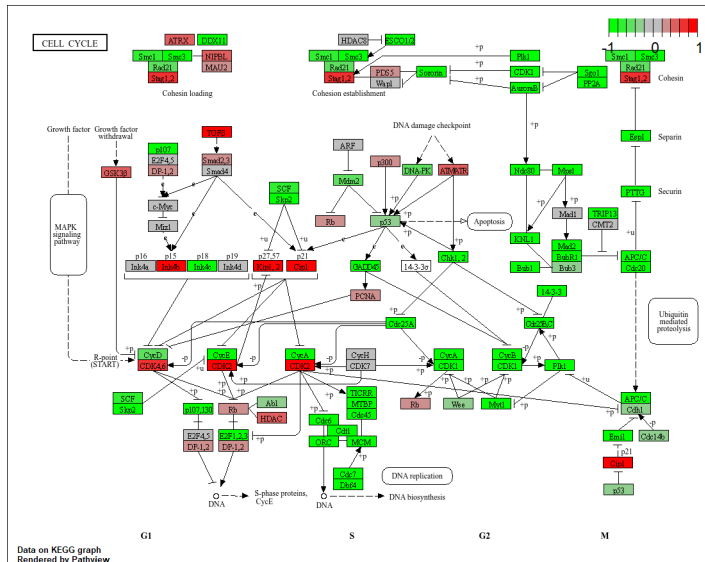
	p.geomean	stat.mean
hsa04110 Cell cycle	8.995727e-06	-4.378644
hsa03030 DNA replication	9.424076e-05	-3.951803
hsa05130 Pathogenic Escherichia coli infection	1.405864e-04	-3.765330
hsa03013 RNA transport	1.246882e-03	-3.059466
hsa03440 Homologous recombination	3.066756e-03	-2.852899
hsa04114 Oocyte meiosis	3.784520e-03	-2.698128
	p.val	q.val
hsa04110 Cell cycle	8.995727e-06	0.001889103
hsa03030 DNA replication	9.424076e-05	0.009841047
hsa05130 Pathogenic Escherichia coli infection	1.405864e-04	0.009841047
hsa03013 RNA transport	1.246882e-03	0.065461279
hsa03440 Homologous recombination	3.066756e-03	0.128803765
hsa04114 Oocyte meiosis	3.784520e-03	0.132458191
	set.size	exp1
hsa04110 Cell cycle	121	8.995727e-06
hsa03030 DNA replication	36	9.424076e-05
hsa05130 Pathogenic Escherichia coli infection	53	1.405864e-04
hsa03013 RNA transport	144	1.246882e-03
hsa03440 Homologous recombination	28	3.066756e-03
hsa04114 Oocyte meiosis	102	3.784520e-03

```
pathview(foldchanges, pathway.id = "hsa04110")
```

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

Info: Working in directory C:/Users/leonn/OneDrive - San Diego State University (SDSU.EDU)/U

Info: Writing image file hsa04110.pathview.png

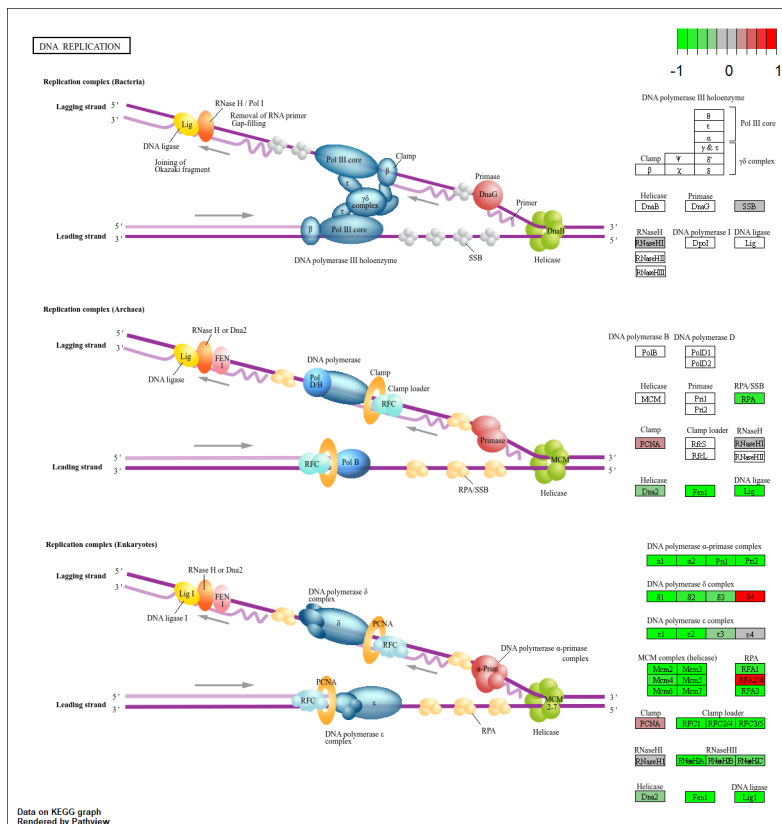


```
pathview(foldchanges, pathway.id = "hsa03030")
```

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

Info: Working in directory C:/Users/leonn/OneDrive - San Diego State University (SDSU.EDU)/U

Info: Writing image file hsa03030.pathview.png



GO & Gene Ontology

```
data(go.sets.hs)
data(go.subs.hs)

#Focus just on GO Biological Process (BP)

gobpsets = go.sets.hs[go.subs.hs$BP]
gobpres = gage(foldchanges, gsets=gobpsets, same.dir=TRUE)
```

```
head(gobpres$less)
```

	p.geomean	stat.mean	p.val
GO:0048285 organelle fission	1.536227e-15	-8.063910	1.536227e-15
GO:0000280 nuclear division	4.286961e-15	-7.939217	4.286961e-15
GO:0007067 mitosis	4.286961e-15	-7.939217	4.286961e-15

G0:0000087	M phase of mitotic cell cycle	1.169934e-14	-7.797496	1.169934e-14
G0:0007059	chromosome segregation	2.028624e-11	-6.878340	2.028624e-11
G0:0000236	mitotic prometaphase	1.729553e-10	-6.695966	1.729553e-10
		q.val	set.size	exp1
G0:0048285	organelle fission	5.841698e-12	376	1.536227e-15
G0:0000280	nuclear division	5.841698e-12	352	4.286961e-15
G0:0007067	mitosis	5.841698e-12	352	4.286961e-15
G0:0000087	M phase of mitotic cell cycle	1.195672e-11	362	1.169934e-14
G0:0007059	chromosome segregation	1.658603e-08	142	2.028624e-11
G0:0000236	mitotic prometaphase	1.178402e-07	84	1.729553e-10

##Reactome

We can use reactome via R or via their fancy new website interface. The web interface wants a set of ENTREZ ID values for your genes of interest. Let's generate that.

```
inds <- abs(res$log2FoldChange) >=2 & res$padj <=0.05
top.genes <- res$entrez[inds]
```

```
write.table(top.genes, file="significant_genes.txt", row.names=FALSE, col.names=FALSE, quote=
```