

Class 14

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Data Import

```
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

Warning: package 'GenomeInfoDb' was built under R version 4.3.2

Loading required package: SummarizedExperiment

Warning: package 'SummarizedExperiment' was built under R version 4.3.2

Loading required package: MatrixGenerics

Loading required package: matrixStats

Attaching package: 'MatrixGenerics'

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

```
colAlls, colAnyNAs, colAnys, colAvgPerRowSet, 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, rowAvgPerColSet,
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
```

```
metaFile <- "GSE37704_metadata.csv"
countFile <- "GSE37704_featurecounts.csv"

# Import metadata and take a peak
metadata = read.csv(metaFile, row.names=1)
head(metadata)
```

```
      condition
SRR493366 control_sirna
SRR493367 control_sirna
SRR493368 control_sirna
SRR493369      hoxa1_kd
SRR493370      hoxa1_kd
SRR493371      hoxa1_kd
```

Have a wee peak at these objects

```
# Import countdata
counts = read.csv(countFile, row.names=1)
head(counts)
```

```
      length SRR493366 SRR493367 SRR493368 SRR493369 SRR493370
ENSG00000186092    918         0         0         0         0
ENSG00000279928    718         0         0         0         0
ENSG00000279457   1982        23        28        29        28
ENSG00000278566    939         0         0         0         0
ENSG00000273547    939         0         0         0         0
ENSG00000187634   3214       124       123       205       212
      SRR493371
ENSG00000186092         0
ENSG00000279928         0
ENSG00000279457        46
ENSG00000278566         0
ENSG00000273547         0
ENSG00000187634       258
```

Data Tidying

I need to get rid of the first `length` column in `counts`:

```
counts <- counts[,-1]
```

Remove any genes with zero counts in all samples/columns

Q. How many genes do we have to start with?

```
nrow(counts)
```

```
[1] 19808
```

- Find the rowSums() this will be zero for any genes with no data
- Find the zero sum genes
- Remove them before doing our DESeq

```
to.rm.ind <- rowSums(counts) == 0  
counts <- counts[!to.rm.ind,]  
nrow(counts)
```

```
[1] 15975
```

DESeq setup and analysis

```
library(DESeq2)
```

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

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

Now run our main DESeq analysis

```
#1 message: false  
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)
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	1.43989e-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	0.5215599	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				

Side-note: QC with PCA

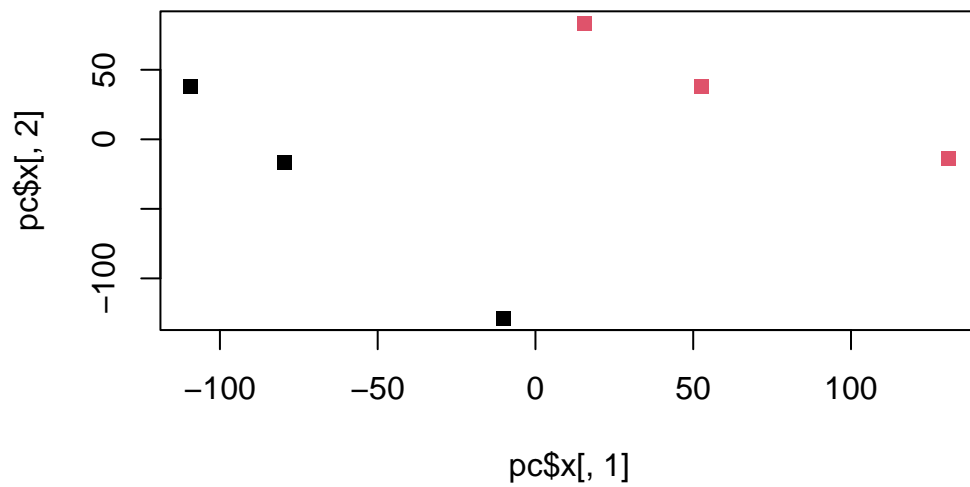
```
pc <- prcomp(t(counts),scale=T)
```

```
summary(pc)
```

Importance of components:

	PC1	PC2	PC3	PC4	PC5	PC6
Standard deviation	87.7211	73.3196	32.89604	31.15094	29.18417	7.387e-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

```
plot(pc$x[,1],pc$x[,2], col=as.factor(metadata$condition), pch=15)
```

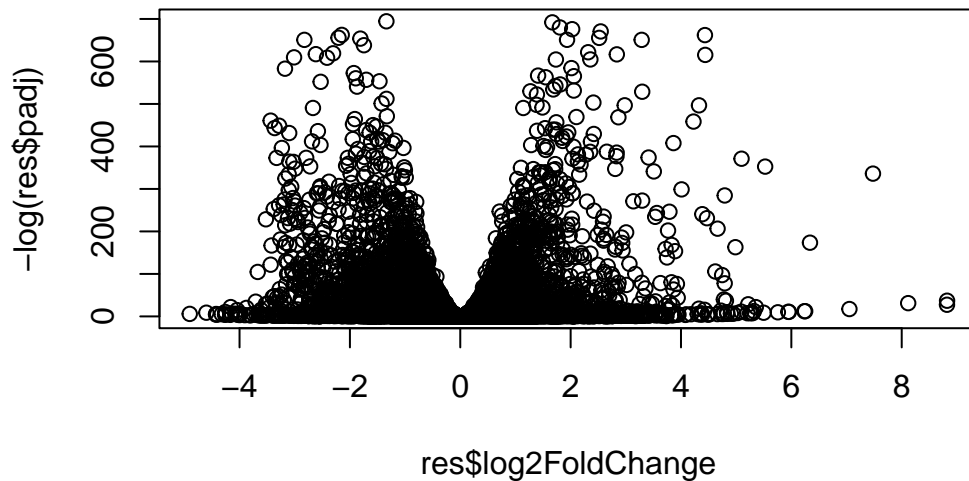


Add annotation data

Save my results

Visualization

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



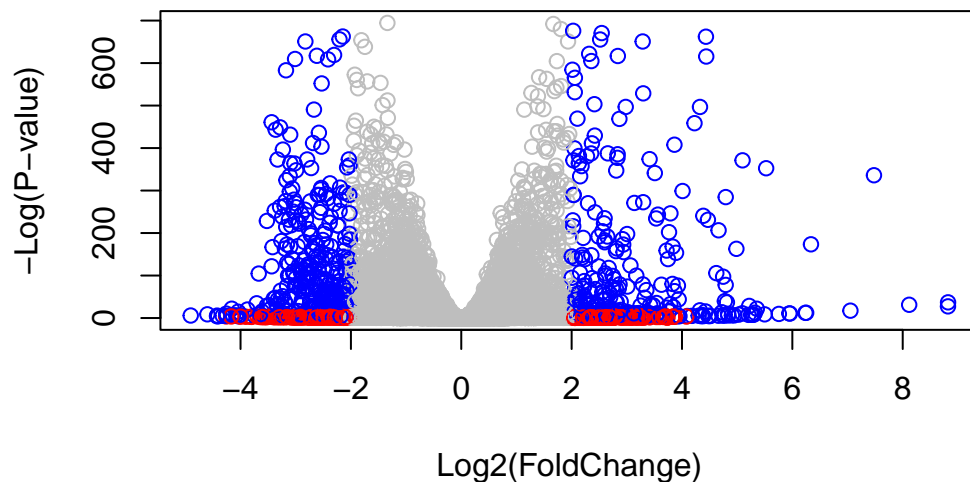
```
# Make a color vector for all genes
mycols <- rep("gray", nrow(res) )

# Color red the genes with absolute fold change above 2
mycols[ abs(res$log2FoldChange) > 2 ] <- "red"

# Color blue those with adjusted p-value less than 0.01
# and absolute fold change more than 2
inds <- (res$pvalue<0.01) & (abs(res$log2FoldChange) > 2 )
mycols[ inds ] <- "blue"
```



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



Let's add some color and annotation data to this plot.

```
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"
```

```
head(row.names(counts))
```

```
[1] "ENSG00000279457" "ENSG00000187634" "ENSG00000188976" "ENSG00000187961"
[5] "ENSG00000187583" "ENSG00000187642"
```

```
res$symbol <- mapIds(org.Hs.eg.db,
                     keys=row.names(counts),
                     keytype = "ENSEMBL",
                     column = "SYMBOL",
                     multiVals="first")
```

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

```
res$entrez <- mapIds(org.Hs.eg.db,
                    keys=row.names(counts),
                    keytype = "ENSEMBL",
                    column = "ENTREZID",
                    multiVals="first")
```

'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	1.43989e-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	0.5215599	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

Geneset enrichmen/Pathway analysis

```
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>).

```
#####
```

The gage() function wants a “vector of importance” in our case here it will be fold-change values with associated entrez gene names.

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

```
data(kegg.sets.hs)
# Get the results
keggres = gage(foldchange, gsets=kegg.sets.hs)
```

```
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

hsa04110 cell cycle

```
pathview(gene.data=foldchange, pathway.id="hsa04110")
```

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

Info: Working in directory /Users/sophialu1999/Desktop/UCSD Biological Sciences Ph.D./BGGN21

Info: Writing image file hsa04110.pathview.png

Have a look at my figure (Figure 1)

Gene Ontology

```
data(go.sets.hs)
data(go.subs.hs)
# Focus on Biological Process subset of GO
gobpsets = go.sets.hs[go.subs.hs$BP]

gobpres = gage(foldchange, gsets=gobpsets, same.dir=TRUE)

head(gobpres$less)
```


	p.geomean	stat.mean	p.val
G0:0048285 organelle fission	1.536227e-15	-8.063910	1.536227e-15
G0:0000280 nuclear division	4.286961e-15	-7.939217	4.286961e-15
G0:0007067 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	5.843127e-12	376	1.536227e-15
G0:0000280 nuclear division	5.843127e-12	352	4.286961e-15
G0:0007067 mitosis	5.843127e-12	352	4.286961e-15
G0:0000087 M phase of mitotic cell cycle	1.195965e-11	362	1.169934e-14
G0:0007059 chromosome segregation	1.659009e-08	142	2.028624e-11
G0:0000236 mitotic prometaphase	1.178690e-07	84	1.729553e-10

Reactome

We will use the online version of Reactome. It wants a list of your genes. We will write this out from R here:

```
sig_genes <- res[res$padj <= 0.05 & !is.na(res$padj), "symbol"]

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

Reactome website: <https://reactome.org/PathwayBrowser/#TOOL=AT>

