

# Class 14: RNASeq mini project

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### Background

Here we work through a complete RNASeq analysis project. The input data comes from a knock-down experiment of a HDX gene.

## Data Import  
Reading the `count()` and `metadata` CSV files

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

Check on data structure

```
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				

```
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
```

```
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
```

Some book-keeping is required as there looks to be a mis-match between metadata and counts columns

```
ncol(counts)
```

```
[1] 7
```

```
nrow(metadata)
```

```
[1] 6
```

Looks like we need to get rid of the first “length” column of our `counts` object.

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

```
colnames(cleancounts)
```

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

```
metadata$id
```

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

```
all( colnames(cleancounts) == metadata$id)
```

```
[1] TRUE
```

### Remove zero count genes

There are lots of genes with zero counts. We can remove these from further analysis

```
head(cleancounts)
```

	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

```
to.keep inds <- rowSums(cleancounts) > 0  
nonzero_counts <- cleancounts[to.keep inds,]
```

### DESeq analysis

Load the package

```
library(DESeq2)
```

Setup DESeq object

```
dds <- DESeqDataSetFromMatrix(countData = nonzero_counts,  
                                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  
  
get results
```

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

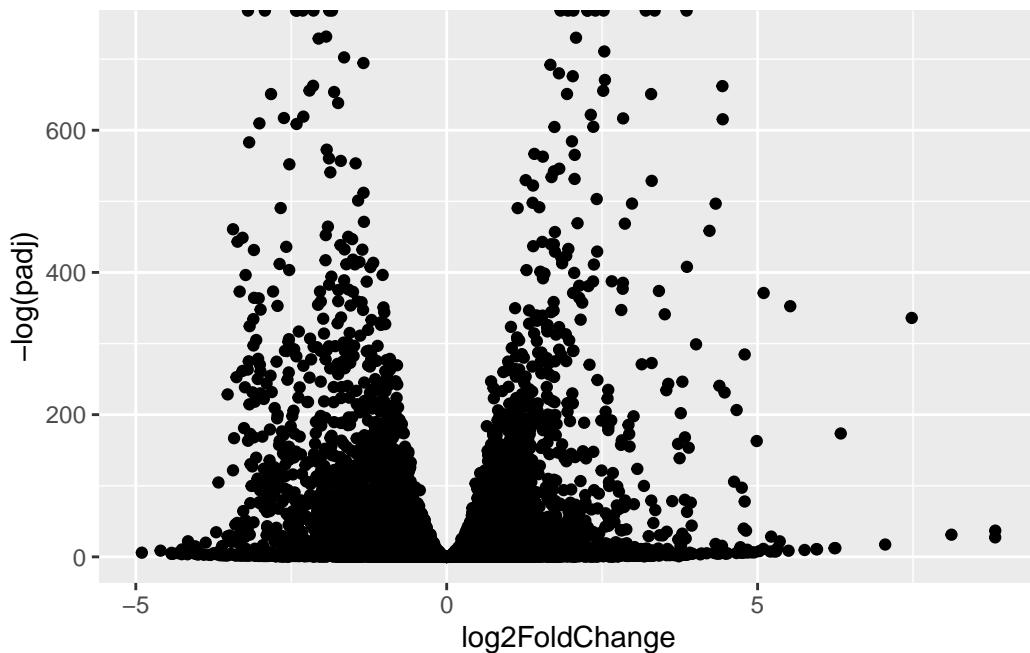
```
log2 fold change (MLE): condition hoxa1 kd vs control sirna  
Wald test p-value: condition hoxa1 kd vs control sirna  
DataFrame with 6 rows and 6 columns  
  baseMean log2FoldChange      lfcSE       stat      pvalue  
  <numeric>      <numeric> <numeric> <numeric> <numeric>  
ENSG00000279457    29.9136     0.1792571  0.3248215  0.551863 5.81042e-01  
ENSG00000187634   183.2296     0.4264571  0.1402658  3.040350 2.36304e-03  
ENSG00000188976  1651.1881    -0.6927205  0.0548465 -12.630156 1.43993e-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 1.76553e-35  
ENSG00000187961 1.13413e-07  
ENSG00000187583 9.19031e-01  
ENSG00000187642 4.03379e-01
```

## Data Visualization

```
library(ggplot2)

ggplot(res) +
  aes(log2FoldChange, -log(padj) ) +
  geom_point()
```

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



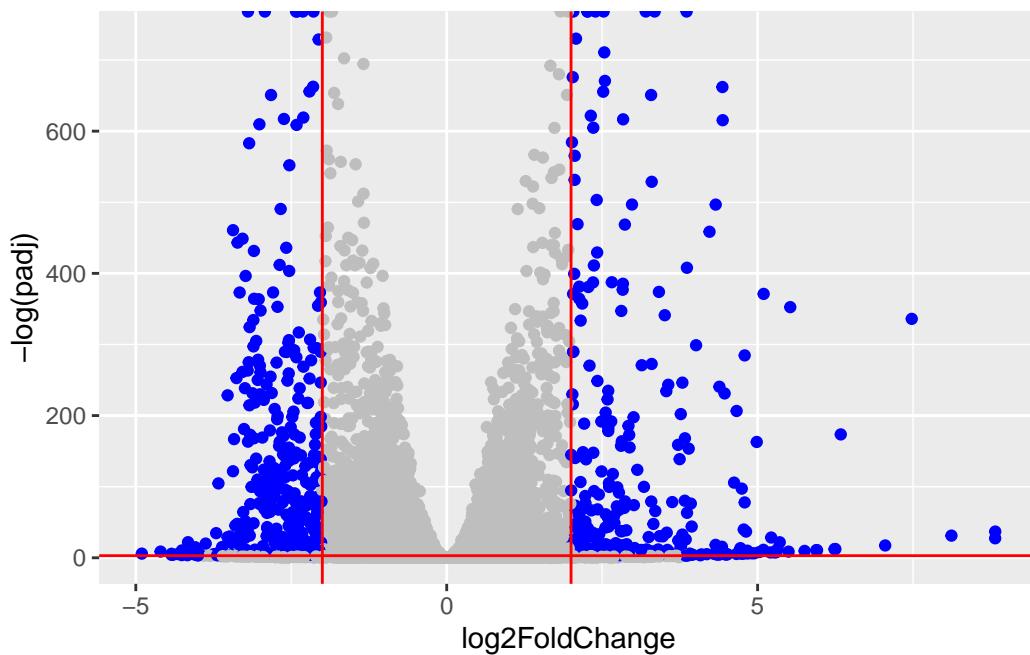
Add threshold lines for fold-change and P-value and color our subset of genes that make these threshold cut-offs in the plot

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

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

```
geom_vline(xintercept = c(-2, 2), color = "red") +  
geom_hline(yintercept = -log(0.05), color = "red") +  
scale_color_identity()
```

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



## Add Annotation

Add gene symbols and entrez ids

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

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

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

```
res$symbol <- mapIds(x=org.Hs.eg.db,
                      keys=row.names(res),
                      keytype = "ENSEMBL",
                      column = "ENTREZID" )
```

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

## Pathway Analysis

Run gage analysis

```
library(gage)
library(gageData)
library(pathview)
```

We need a named vector of fold-change values as input for gage

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

```
[1] 0.17925708 0.42645712 -0.69272046 0.72975561 0.04057653 0.54281049
```

```
data(kegg.sets.hs)

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

head(keggres$less, 2)
```

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