

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 HOX gene.

Data Import

Reading the `counts` and `metadata` CSV files

```
metaFile <- "GSE37704_metadata.csv"  
countFile <- "GSE37704_featurecounts.csv"  
colData = read.csv(metaFile, row.names=1)  
countData = read.csv(countFile, row.names=1)
```

Check on data structure

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

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

```
countData <- read.csv(countFile, row.names = 1)
countData[] <- lapply(countData, function(x) as.numeric(as.character(x)))
cleancounts <- as.matrix(countData[, -1])
nonzero_counts <- cleancounts[rowSums(cleancounts) > 0, ]
head(nonzero_counts)
```

	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

DESeq Analysis

Load the package

```
library(DESeq2)
```

Setup DESeq

```
dds = DESeqDataSetFromMatrix(countData=nonzero_counts,
                             colData=colData,
                             design=~condition)
```

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

Run DESeq

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

Get results

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

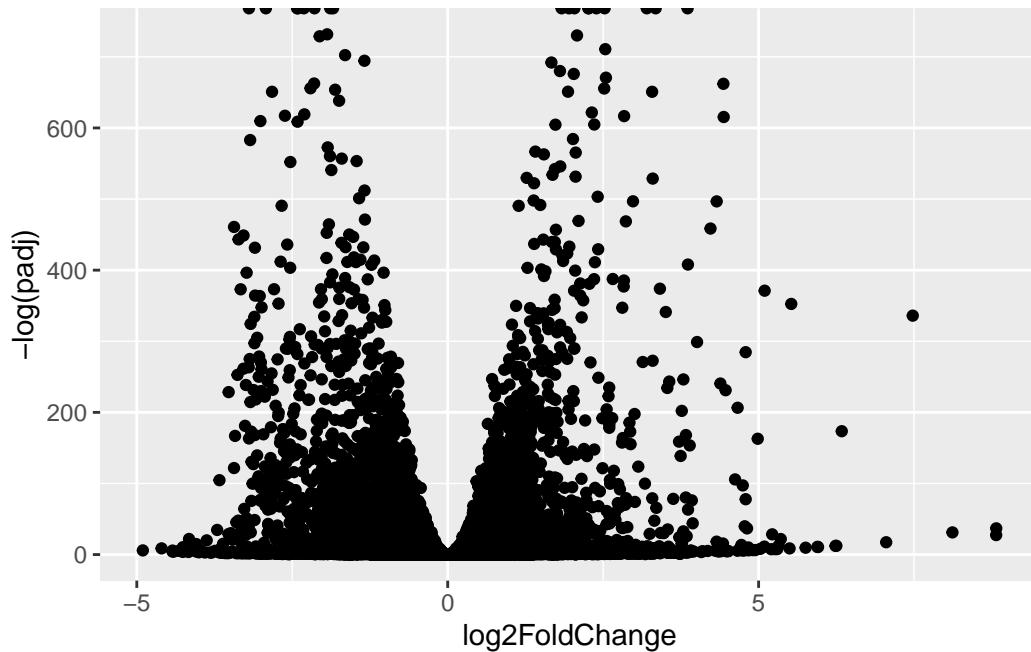
Data Visualization

Volcano Plot

```
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.

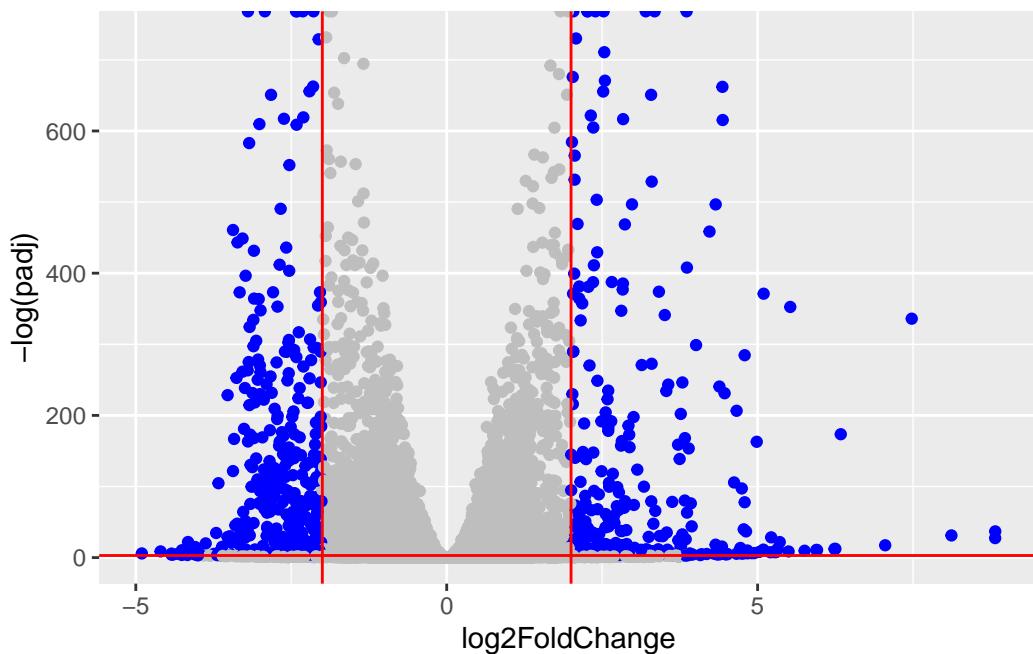
```

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

ggplot(res) +
  aes(log2FoldChange, -log(padj)) +
  geom_point(col = mycols) +
  geom_vline(xintercept =c(-2,2), col="red") +
  geom_hline(yintercept = -log(0.05), col ="red")

```

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(org.Hs.eg.db,
  keys = row.names(res),
  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(res),
  keytype = "ENSEMBL",
  column = "ENTREZID",
  multiVals = "first")

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

res$name = mapIds(org.Hs.eg.db,
  keys = row.names(res),
  keytype = "ENSEMBL",
  column = "GENENAME",
  multiVals = "first")

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

head(res, 10)

log2 fold change (MLE): condition hoxa1 kd vs control sirna
Wald test p-value: condition hoxa1 kd vs control sirna
DataFrame with 10 rows and 9 columns
  baseMean log2FoldChange      lfcSE      stat      pvalue
  <numeric>      <numeric> <numeric> <numeric>      <numeric>
ENSG00000279457    29.913579     0.1792571  0.3248216   0.551863 5.81042e-01
ENSG00000187634   183.229650     0.4264571  0.1402658   3.040350 2.36304e-03
ENSG00000188976  1651.188076    -0.6927205  0.0548465 -12.630158 1.43990e-36
ENSG00000187961   209.637938     0.7297556  0.1318599   5.534326 3.12428e-08
ENSG00000187583    47.255123     0.0405765  0.2718928   0.149237 8.81366e-01
ENSG00000187642   11.979750     0.5428105  0.5215598   1.040744 2.97994e-01
ENSG00000188290   108.922128    2.0570638  0.1969053  10.446970 1.51282e-25

```

ENSG00000187608	350.716868	0.2573837	0.1027266	2.505522	1.22271e-02
ENSG00000188157	9128.439422	0.3899088	0.0467163	8.346304	7.04321e-17
ENSG00000237330	0.158192	0.7859552	4.0804729	0.192614	8.47261e-01
	padj	symbol	entrez		name
	<numeric>	<character>	<character>		<character>
ENSG00000279457	6.86555e-01	NA	NA		NA
ENSG00000187634	5.15718e-03	SAMD11	148398	sterile alpha motif ..	
ENSG00000188976	1.76549e-35	NOC2L	26155	NOC2 like nucleolar ..	
ENSG00000187961	1.13413e-07	KLHL17	339451	kelch like family me..	
ENSG00000187583	9.19031e-01	PLEKHN1	84069	pleckstrin homology ..	
ENSG00000187642	4.03379e-01	PERM1	84808	PPARGC1 and ESRR ind..	
ENSG00000188290	1.30538e-24	HES4	57801	hes family bHLH tran..	
ENSG00000187608	2.37452e-02	ISG15	9636	ISG15 ubiquitin like..	
ENSG00000188157	4.21963e-16	AGRN	375790		agrin
ENSG00000237330	NA	RNF223	401934	ring finger protein ..	

Let's Reorder results by adjusted p-value and save as CSV file:

```
res <- res[order(res$padj), ]  
  
write.csv(res, file = "deseq_results.csv")
```

Pathway Analysis

KEGG pathways

- run this in console: `BiocManager::install(c("pathview", "gage", "gageData")) *`

Run gage analysis w/ KEGG

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

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

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

```
1266      54855      1465      2034      2150      6659  
-2.422719  3.201955 -2.313738 -1.888019  3.344508  2.392288
```

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

```
head(keggres$less, 2)
```

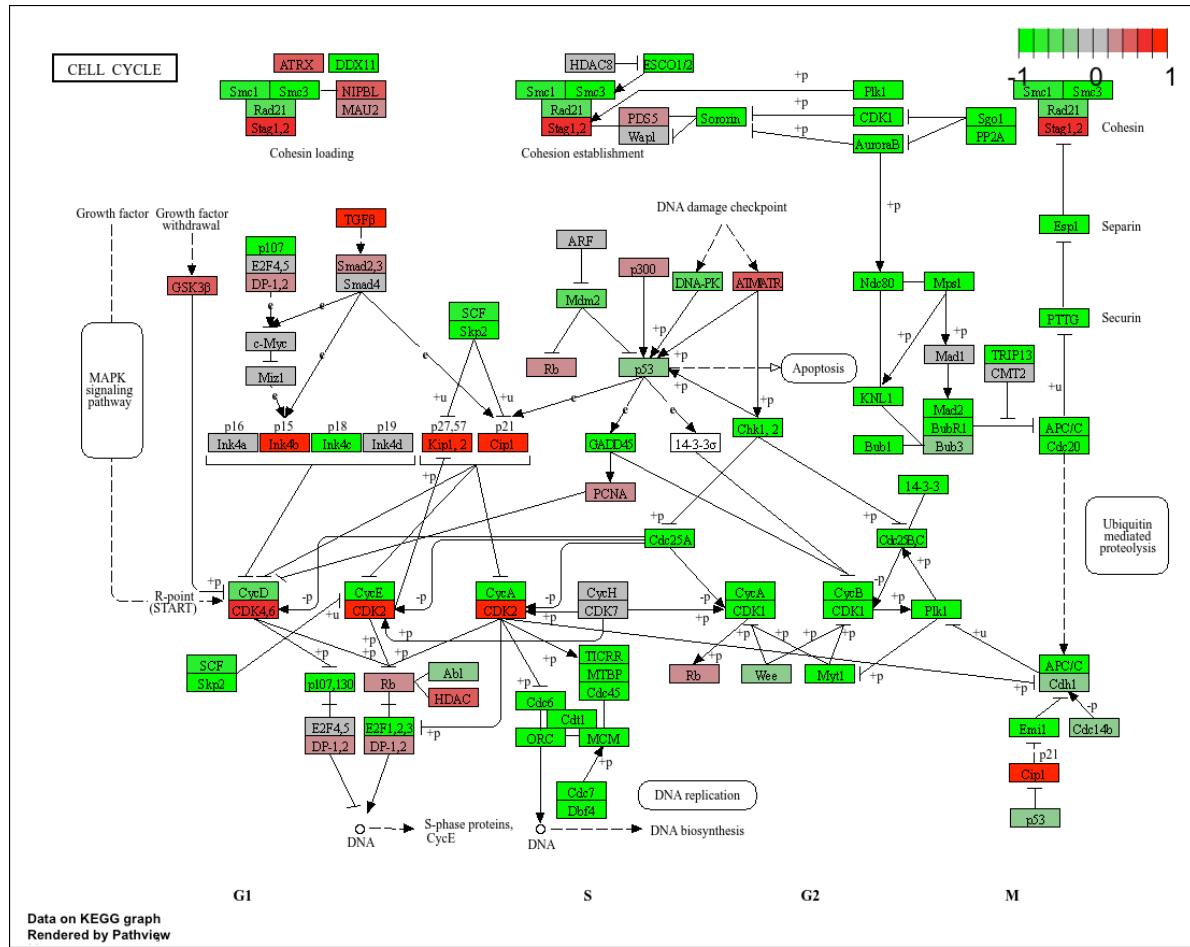
	p.geomean	stat.mean	p.val	q.val
hsa04110 Cell cycle	8.995727e-06	-4.378644	8.995727e-06	0.001889103
hsa03030 DNA replication	9.424076e-05	-3.951803	9.424076e-05	0.009841047
	set.size	exp1		
hsa04110 Cell cycle	121	8.995727e-06		
hsa03030 DNA replication	36	9.424076e-05		

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

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

```
Info: Working in directory /Users/lourdyakoob/Desktop/Class 14
```

```
Info: Writing image file hsa04110.pathview.png
```

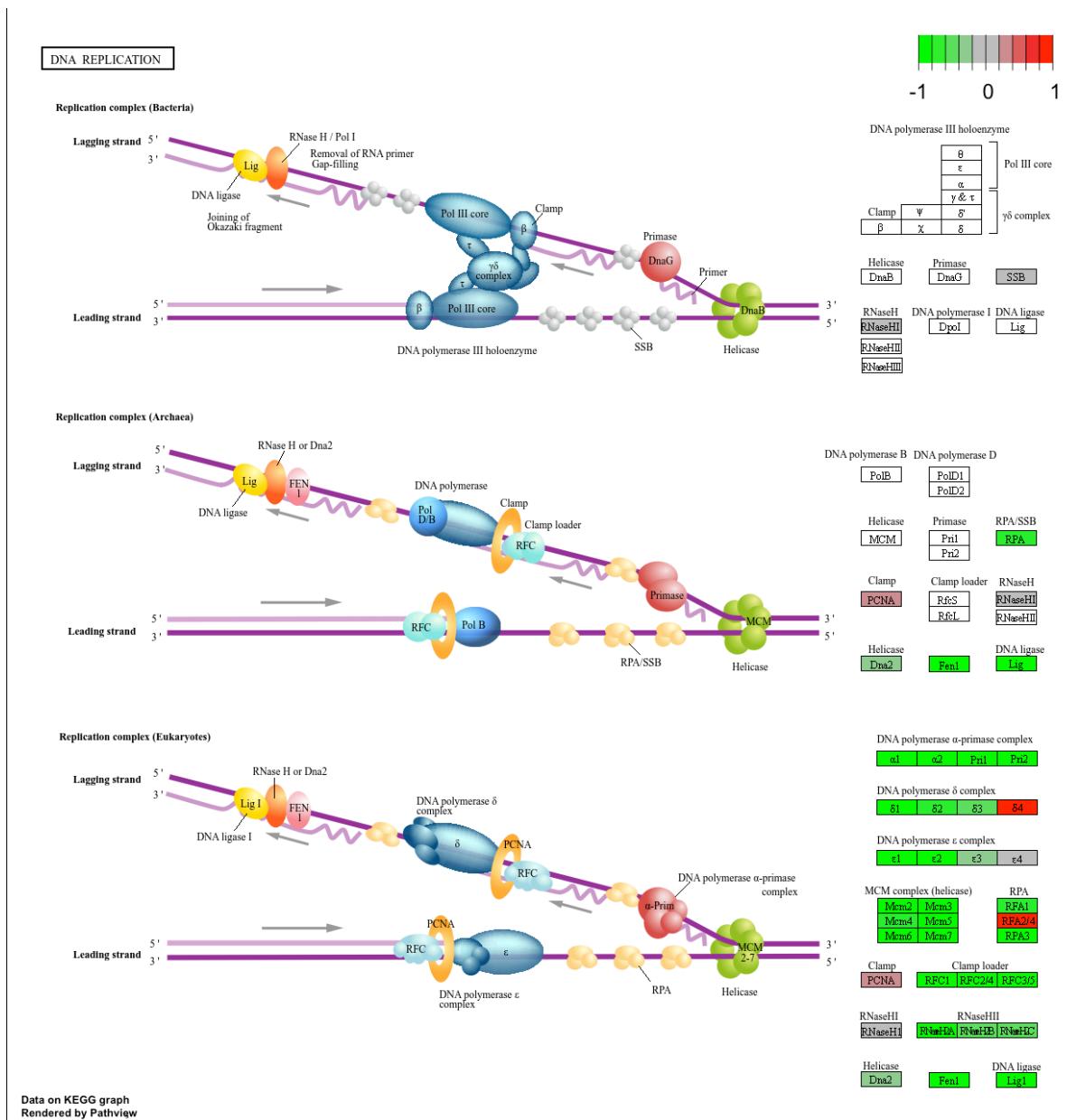


```
pathview(pathway.id = "hsa03030", gene.data = foldchanges)
```

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

```
Info: Working in directory /Users/lourdyakoob/Desktop/Class 14
```

```
Info: Writing image file hsa03030.pathview.png
```



GO terms

Same analysis but using GO genesets rather than KEGG.

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

```

gobpsets = go.sets.hs[go.subs.hs$BP]

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

lapply(gobpres, head)

```

\$greater

	p.geomean	stat.mean	p.val
GO:0007156 homophilic cell adhesion	8.519724e-05	3.824205	8.519724e-05
GO:0002009 morphogenesis of an epithelium	1.396681e-04	3.653886	1.396681e-04
GO:0048729 tissue morphogenesis	1.432451e-04	3.643242	1.432451e-04
GO:0007610 behavior	1.925222e-04	3.565432	1.925222e-04
GO:0060562 epithelial tube morphogenesis	5.932837e-04	3.261376	5.932837e-04
GO:0035295 tube development	5.953254e-04	3.253665	5.953254e-04
	q.val	set.size	exp1
GO:0007156 homophilic cell adhesion	0.1951953	113	8.519724e-05
GO:0002009 morphogenesis of an epithelium	0.1951953	339	1.396681e-04
GO:0048729 tissue morphogenesis	0.1951953	424	1.432451e-04
GO:0007610 behavior	0.1967577	426	1.925222e-04
GO:0060562 epithelial tube morphogenesis	0.3565320	257	5.932837e-04
GO:0035295 tube development	0.3565320	391	5.953254e-04

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

\$stats

	stat.mean	exp1
GO:0007156 homophilic cell adhesion	3.824205	3.824205

```

GO:0002009 morphogenesis of an epithelium 3.653886 3.653886
GO:0048729 tissue morphogenesis          3.643242 3.643242
GO:0007610 behavior                   3.565432 3.565432
GO:0060562 epithelial tube morphogenesis 3.261376 3.261376
GO:0035295 tube development           3.253665 3.253665

```

```
head(gobpres$less, 4)
```

	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
GO:0000087 M phase of mitotic cell cycle	1.169934e-14	-7.797496	1.169934e-14
	q.val	set.size	exp1
GO:0048285 organelle fission	5.841698e-12	376	1.536227e-15
GO:0000280 nuclear division	5.841698e-12	352	4.286961e-15
GO:0007067 mitosis	5.841698e-12	352	4.286961e-15
GO:0000087 M phase of mitotic cell cycle	1.195672e-11	362	1.169934e-14

Reactome

Lots of folks like the reactome web interface. You can also run this as an R function but lets look at the website first <https://reactome.org/>

The website wants a text file with one gene symbol per line of the genes you want to map to the pathways.

```
sig_genes <- res[res$padj <= 0.05 & !is.na(res$padj), "symbol"]
print(paste("Total number of significant genes:", length(sig_genes)))
```

```
[1] "Total number of significant genes: 8147"
```

and write out to a file:

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

Save Our Results

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