Class 14: RNASeq mini-project

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Here we will perform a complete RNASeq analysis from counts to pathways and biological interpretation.

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.

Differential Expression Analysis

Required packages

```
library(DESeq2)
library(AnnotationDbi)
library(org.Hs.eg.db)
library(pathview)
library(gage)
library(gageData)
```

Data import

```
colData <- read.csv("GSE37704_metadata.csv", row.names=1)
countData <- read.csv("GSE37704_featurecounts.csv", row.name=1)</pre>
```

head(colData)

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

head(countData)

	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
	SRR4933	371				

ENSG00000186092 0 ENSG00000279928 0

ENSG00000279457	46
ENSG00000278566	0
ENSG00000273547	0
ENSG00000187634	258

Check the corespondance of colData row and countData columns.

rownames(colData)

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

colnames(countData)

- [1] "length" "SRR493366" "SRR493367" "SRR493368" "SRR493369" "SRR493370"
- [7] "SRR493371"

Remove the troublesome first column so we match the metadata.

counts <- countData[,-1]</pre>

head(counts)

	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

all (rownames(colData) == colnames(counts))

[1] TRUE

To remove the first column \dots

all(c(T,T,T))

[1] TRUE

Remove zero count genes

We will have rows in **counts** for genes that we can not say anything about because they have zero expression in the particular tissue we are looking at.

head(counts)

	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

If the rowSums() is zero then a give gene (i.e. row) has no count data and we should exclude these genes from further consideration.

```
to.keep <- rowSums(counts) != 0
cleancounts <- counts[to.keep,]</pre>
```

Q. How many genes do we have left?

```
nrow(cleancounts)
```

[1] 15975

Setup DeSeq object for analysis

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

Run DESeq analysis

dds <- DESeq(dds)

```
estimating size factors

estimating dispersions

gene-wise dispersion estimates

mean-dispersion relationship

final dispersion estimates

fitting model and testing
```

Extract the results

```
res <- results(dds)
head(res)</pre>
```

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></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
	pao	dj			
	<numerio< td=""><td>c></td><td></td><td></td><td></td></numerio<>	c>			

ENSG00000279457 6.86555e-01

ENSG00000187634 5.15718e-03

ENSG0000188976 1.76549e-35

ENSG00000187961 1.13413e-07

ENSG00000187583 9.19031e-01

ENSG00000187642 4.03379e-01

Add Gene annotation

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 0.0405765 0.2718928 0.149237 8.81366e-01 ENSG00000187583 47.2551 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 head(rownames(res)) [1] "ENSG00000279457" "ENSG00000187634" "ENSG00000188976" "ENSG00000187961" [5] "ENSG00000187583" "ENSG00000187642" library(AnnotationDbi) library(org.Hs.eg.db) columns(org.Hs.eg.db) [1] "ACCNUM" "ALIAS" "ENSEMBL" "ENSEMBLPROT" "ENSEMBLTRANS" [6] "ENTREZID" "EVIDENCEALL" "ENZYME" "EVIDENCE" "GENENAME"

"GOALL"

"REFSEQ"

"ONTOLOGYALL"

"IPI"

"PATH"

"SYMBOL"

"MAP"

"PFAM"

"UCSCKG"

"GO"

"ONTOLOGY"

"PROSITE"

[11] "GENETYPE"

[16] "OMIM"

[21] "PMID"

[26] "UNIPROT"

```
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"
                                                                 "MAP"
                    "GO"
                                                  "IPI"
                                   "GOALL"
[16] "OMIM"
                    "ONTOLOGY"
                                   "ONTOLOGYALL" "PATH"
                                                                 "PFAM"
[21] "PMID"
                    "PROSITE"
                                   "REFSEQ"
                                                  "SYMBOL"
                                                                 "UCSCKG"
[26] "UNIPROT"
```

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

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

'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	lfcSH	E stat pvalue
	<numeric></numeric>	<numeric></numeric>	<numeric></numeric>	<pre>> <numeric> <numeric></numeric></numeric></pre>
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	5 -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	3 1.040744 2.97994e-01
ENSG00000188290	108.922128	2.0570638	0.1969053	3 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></numeric>	<character> <c< td=""><td>haracter></td><td><character></character></td></c<></character>	haracter>	<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

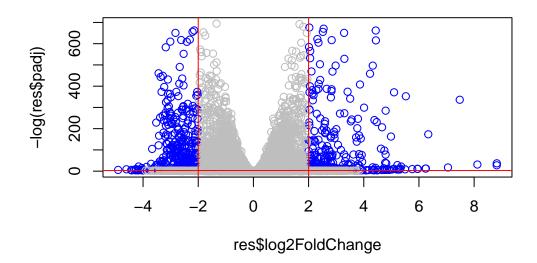
Save my results to a CSV file

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

Result visualization

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

```
#mycols
plot(res$log2FoldChange, -log(res$padj), col=mycols)
abline(v=-2, col="red")
abline(v=+2, col="red")
abline(h=-log(0.05), col="red")
```



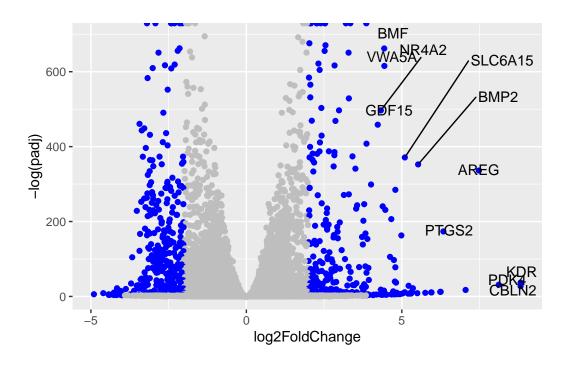
```
library(ggplot2)
library(ggrepel)

ggplot(res) +
  aes(log2FoldChange, -log(padj), label=res$symbol) +
  geom_point(col=mycols) +
  geom_text_repel(max.overlaps=8)
```

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

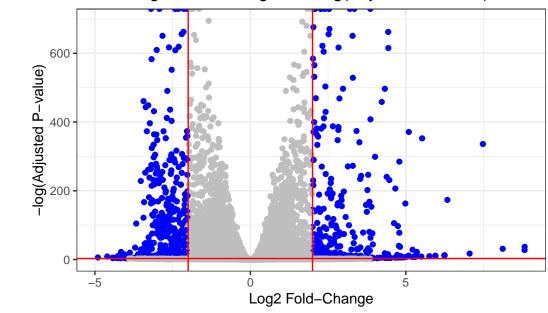
Warning: Removed 1409 rows containing missing values or values outside the scale range (`geom_text_repel()`).

Warning: ggrepel: 14555 unlabeled data points (too many overlaps). Consider increasing max.overlaps



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

Plot of Log2 Fold-Change vs. -log(Adjusted P-value)



Pathway analysis

KEGG genesets/paathways

```
data(kegg.sets.hs)
data(sigmet.idx.hs)
# Focus on signaling and metabolic pathways only
kegg.sets.hs = kegg.sets.hs[sigmet.idx.hs]
# Examine the first 3 pathways
head(kegg.sets.hs, 3)
$`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"
                                                                     "54576"
                                                  "54490"
                                                            "54575"
```

```
[25] "54577"
                                  "54600"
               "54578"
                        "54579"
                                            "54657"
                                                      "54658"
                                                               "54659"
                                                                         "54963"
[33] "574537" "64816"
                         "7083"
                                  "7084"
                                            "7172"
                                                      "7363"
                                                               "7364"
                                                                         "7365"
[41] "7366"
                                            "7378"
                                                      "7498"
                                                               "79799"
                                                                         "83549"
               "7367"
                         "7371"
                                  "7372"
[49] "8824"
               "8833"
                         "9"
                                  "978"
$`hsa00230 Purine metabolism`
  [1] "100"
                "10201"
                         "10606"
                                   "10621"
                                             "10622"
                                                       "10623"
                                                                "107"
                                                                          "10714"
  [9] "108"
                "10846"
                         "109"
                                   "111"
                                             "11128"
                                                       "11164"
                                                                "112"
                                                                          "113"
 [17] "114"
                "115"
                         "122481" "122622" "124583" "132"
                                                                "158"
                                                                          "159"
                                                       "204"
 [25] "1633"
                "171568" "1716"
                                   "196883" "203"
                                                                "205"
                                                                          "221823"
 [33] "2272"
                "22978"
                         "23649"
                                   "246721"
                                             "25885"
                                                       "2618"
                                                                "26289"
                                                                          "270"
 [41] "271"
                "27115"
                         "272"
                                   "2766"
                                             "2977"
                                                       "2982"
                                                                "2983"
                                                                          "2984"
                "2987"
                                                                "318"
                                                                          "3251"
 [49] "2986"
                         "29922"
                                   "3000"
                                             "30833"
                                                       "30834"
 [57] "353"
                "3614"
                         "3615"
                                   "3704"
                                             "377841" "471"
                                                                "4830"
                                                                          "4831"
 [65] "4832"
                "4833"
                         "4860"
                                   "4881"
                                             "4882"
                                                       "4907"
                                                                "50484"
                                                                          "50940"
                                                                "5139"
 [73] "51082"
                "51251"
                         "51292"
                                   "5136"
                                             "5137"
                                                       "5138"
                                                                          "5140"
 [81] "5141"
                "5142"
                         "5143"
                                   "5144"
                                             "5145"
                                                       "5146"
                                                                "5147"
                                                                          "5148"
 [89] "5149"
                "5150"
                         "5151"
                                   "5152"
                                             "5153"
                                                       "5158"
                                                                "5167"
                                                                          "5169"
 [97] "51728"
                "5198"
                         "5236"
                                   "5313"
                                             "5315"
                                                       "53343"
                                                                "54107"
                                                                          "5422"
                "5425"
                         "5426"
                                   "5427"
                                                       "5431"
                                                                "5432"
[105] "5424"
                                             "5430"
                                                                          "5433"
[113] "5434"
                                   "5437"
                                             "5438"
                                                                "5440"
                "5435"
                         "5436"
                                                       "5439"
                                                                          "5441"
[121] "5471"
                "548644" "55276"
                                   "5557"
                                             "5558"
                                                       "55703"
                                                                "55811"
                                                                          "55821"
                "5634"
[129] "5631"
                         "56655"
                                   "56953"
                                             "56985"
                                                       "57804"
                                                                "58497"
                                                                          "6240"
[137] "6241"
                "64425"
                         "646625" "654364"
                                             "661"
                                                       "7498"
                                                                "8382"
                                                                          "84172"
                                   "8622"
[145] "84265"
                "84284"
                         "84618"
                                             "8654"
                                                       "87178"
                                                                "8833"
                                                                          "9060"
[153] "9061"
                "93034"
                         "953"
                                   "9533"
                                             "954"
                                                       "955"
                                                                "956"
                                                                          "957"
                "9615"
[161] "9583"
foldchanges = res$log2FoldChange
names(foldchanges) = res$entrez
head(foldchanges)
```

```
1266 54855 1465 51232 2034 2317 -2.422719 3.201955 -2.313738 -2.059631 -1.888019 -1.649792
```

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

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

p.geomean stat.mean p.val hsa04110 Cell cycle 8.995727e-06 -4.378644 8.995727e-06

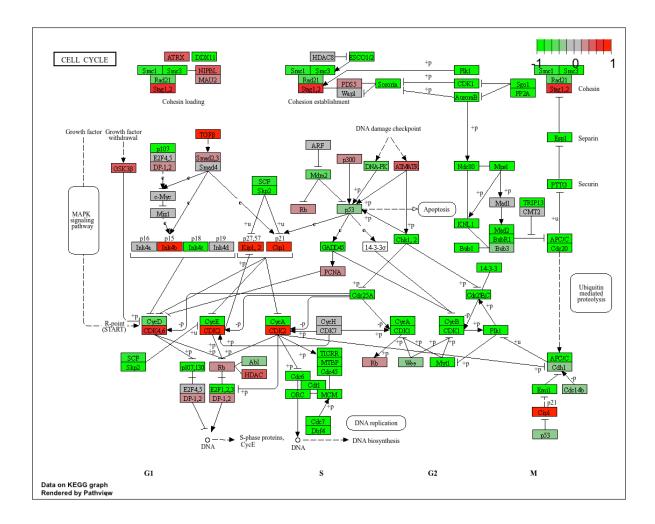
```
hsa03013 RNA transport
                          1.375901e-03 -3.028500 1.375901e-03
hsa03440 Homologous recombination 3.066756e-03 -2.852899 3.066756e-03
                                q.val set.size
hsa04110 Cell cycle
                           0.001448312
                                       121 8.995727e-06
hsa03030 DNA replication
                           0.007586381
                                         36 9.424076e-05
hsa03013 RNA transport
                           0.073840037
                                         144 1.375901e-03
hsa03440 Homologous recombination 0.121861535
                                         28 3.066756e-03
```

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

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

Info: Working in directory /Users/yoonjinlim/Desktop/BIMM 143/Class014

Info: Writing image file hsa04110.pathview.png



Gene Ontology (GO) genesets

head(gobpres\$less, 5)

```
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(foldchanges, gsets=gobpsets)
```

p.val

p.geomean stat.mean

```
GO:0048285 organelle fission
                                         1.536227e-15 -8.063910 1.536227e-15
GO:0000280 nuclear division
                                         4.286961e-15 -7.939217 4.286961e-15
GD: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
                                                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
                                                           352 4.286961e-15
                                         5.841698e-12
GO:0000087 M phase of mitotic cell cycle 1.195672e-11
                                                           362 1.169934e-14
GO:0007059 chromosome segregation
                                                           142 2.028624e-11
                                         1.658603e-08
```

Reactome analysis online

We need to make a little file of our significant genes that we can upload to the reactome webpage:

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

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

```
sig_genes[6]
```

ENSG00000136068 "FLNB"

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

Then, to perform pathway analysis online go to the Reactome website (https://reactome.org/PathwayBrowser/# Select "choose file" to upload your significant gene list. Then, select the parameters "Project to Humans", then click "Analyze".

