Class 14: RNA-Seq analysis mini-project

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Background

The data 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

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

Inspect and Tidy Data

Does the counts columns match the colData rows?

```
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
	SRR4933	371				
ENSG00000186092		0				
ENSG00000279928		0				
ENSG00000279457		46				
ENSG00000278566		0				
ENSG00000273547		0				
ENSG00000187634	2	258				

colData\$id

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

colnames(counts)

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

Q. Complete the code below to remove the troublesome first column from count-Data

The fix here looks to be removing the first "length" column from 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

Check for matching countData and colData:

```
colnames(countData) == colData$id
```

- [1] TRUE TRUE TRUE TRUE TRUE TRUE
- Q. How many genes in total?

```
nrow(countData)
```

- [1] 19808
- Q. Filter to remove zero count genes (rows where there are zero counts in all columns). How many genes are left and Q. Complete the code below to filter countData to exclude genes (i.e. rows) where we have 0 read count across all samples (i.e. columns).

```
to.keep.inds <- rowSums(countData) > 0
new.counts <- countData[to.keep.inds,]
nrow(new.counts)</pre>
```

[1] 15975

Setup for DESeq

```
library(DESeq2)
```

Setup input object for DESeq:

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</pre>
```

final dispersion estimates

fitting model and testing

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

head(res)

 $\log 2$ 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>
                  29.9136
ENSG00000279457
                               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
                               0.0405765 0.2718928
ENSG00000187583
                  47.2551
                                                     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
```

Q. Call the summary() function on your results to get a sense of how many genes are up or down-regulated at the default 0.1 p-value cutoff.

```
summary(res)
```

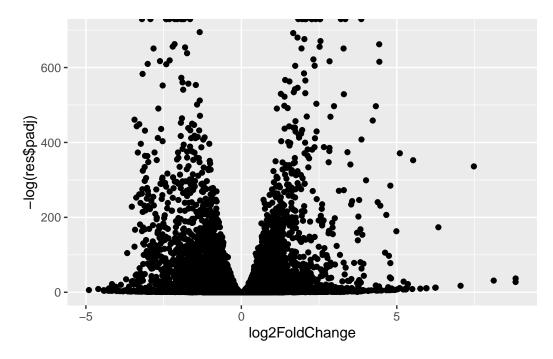
```
out of 15975 with nonzero total read count
adjusted p-value < 0.1
LFC > 0 (up) : 4349, 27%
LFC < 0 (down) : 4396, 28%
outliers [1] : 0, 0%
low counts [2] : 1237, 7.7%
(mean count < 0)
[1] see 'cooksCutoff' argument of ?results
[2] see 'independentFiltering' argument of ?results</pre>
```

Volcano Plot of Results

```
library(ggplot2)

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

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

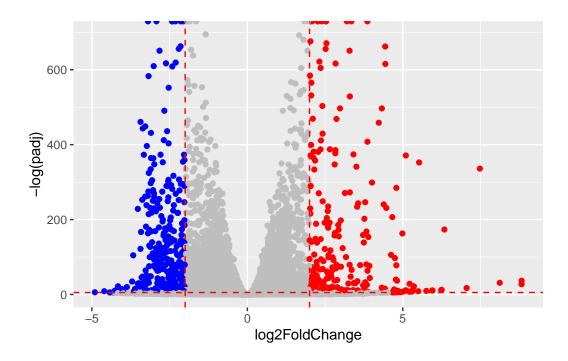


Q. Improve this plot by completing the below code, which adds color and axis labels.

```
mycols <- rep("grey", nrow(res))
mycols[res$log2FoldChange >= 2] <- "red"
mycols[res$log2FoldChange <= -2] <- "blue"
mycols[res$padj > 0.005] <- "gray"</pre>
```

```
ggplot(res) +
  aes(log2FoldChange, -log(padj)) +
  geom_point(col=mycols) +
  geom_vline(xintercept = c(-2,2), col="red", linetype = "dashed") +
  geom_hline(yintercept = -log(0.005), col = "red", linetype = "dashed")
```

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



Gene Annotation

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

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

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

Q. Use the mapIDs() function multiple times to add SYMBOL, ENTREZID and GENENAME annotation to our results by completing the code below.

Add gene SYMBOL and ENTREZID

```
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"
                                   "REFSEQ"
[21] "PMID"
                                                                  "UCSCKG"
                    "PROSITE"
                                                   "SYMBOL"
[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

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

```
head(res, 10)
```

 $\log 2$ 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

Data Tame with 10 10ws and 5 columns					
pvalue	stat	lfcSE	log2FoldChange	baseMean	
<numeric></numeric>	<numeric></numeric>	<numeric></numeric>	<numeric></numeric>	<numeric></numeric>	
5.81042e-01	0.551863	0.3248216	0.1792571	29.913579	ENSG00000279457
2.36304e-03	3.040350	0.1402658	0.4264571	183.229650	ENSG00000187634
1.43989e-36	-12.630158	0.0548465	-0.6927205	1651.188076	ENSG00000188976
3.12428e-08	5.534326	0.1318599	0.7297556	209.637938	ENSG00000187961
8.81366e-01	0.149237	0.2718928	0.0405765	47.255123	ENSG00000187583
2.97994e-01	1.040744	0.5215599	0.5428105	11.979750	ENSG00000187642
1.51282e-25	10.446970	0.1969053	2.0570638	108.922128	ENSG00000188290
1.22271e-02	2.505522	0.1027266	0.2573837	350.716868	ENSG00000187608
7.04321e-17	8.346304	0.0467163	0.3899088	9128.439422	ENSG00000188157
8.47261e-01	0.192614	4.0804729	0.7859552	0.158192	ENSG00000237330
	name	entrez	symbol	padj	
	<pre><character></character></pre>	naracter>	<character> <cl< td=""><td><numeric></numeric></td><td></td></cl<></character>	<numeric></numeric>	
	NA	NA	NA	6.86555e-01	ENSG00000279457
	148398	148398	SAMD11	5.15718e-03	ENSG00000187634
	26155	26155	NOC2L	1.76549e-35	ENSG00000188976
	339451	339451	KLHL17	1.13413e-07	ENSG00000187961
	84069	84069	PLEKHN1	9.19031e-01	ENSG00000187583
	84808	84808	PERM1	4.03379e-01	ENSG00000187642
	57801	57801	HES4	1.30538e-24	ENSG00000188290
	9636	9636	ISG15	2.37452e-02	ENSG00000187608
	375790	375790	AGRN	4.21963e-16	ENSG00000188157
	401934	401934	RNF223	NA	ENSG00000237330

Q. Finally for this section let's reorder these results by adjusted p-value and save them to a CSV file in your current project directory.

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

```
head(read.csv("deseq_results.csv"))
```

X baseMean log2FoldChange lfcSE stat pvalue padj

```
1 ENSG00000117519 4483.627
                              -2.422719 0.06000162 -40.37756
                                                                      0
2 ENSG00000183508 2053.881
                               3.201955 0.07241720 44.21540
                                                                      0
3 ENSG00000159176 5692.463
                              -2.313738 0.05755337 -40.20160
                                                                 0
                                                                      0
4 ENSG00000150938 7442.986
                              -2.059631 0.05384491 -38.25118
                                                                 0
                                                                      0
5 ENSG00000116016 4423.947
                              -1.888019 0.04316799 -43.73656
                                                                 0
                                                                      0
6 ENSG00000136068 3796.127
                              -1.649792 0.04393544 -37.55037
                                                                      0
  symbol entrez name
   CNN3
         1266 1266
2 TENT5C 54855 54855
3 CSRP1 1465 1465
4 CRIM1 51232 51232
5 EPAS1 2034 2034
6 FLNB
          2317 2317
```

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.

Input vector for gage():

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

Load up the KEGG gene-sets:

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

Run pathway analysis:

```
keggres = gage(foldchanges, 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
                                               1.375901e-03 -3.028500
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
                                               1.375901e-03 0.072234819
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 1.375901e-03
hsa03440 Homologous recombination
                                                     28 3.066756e-03
hsa04114 Oocyte meiosis
                                                    102 3.784520e-03
```

Cell cycle figure:

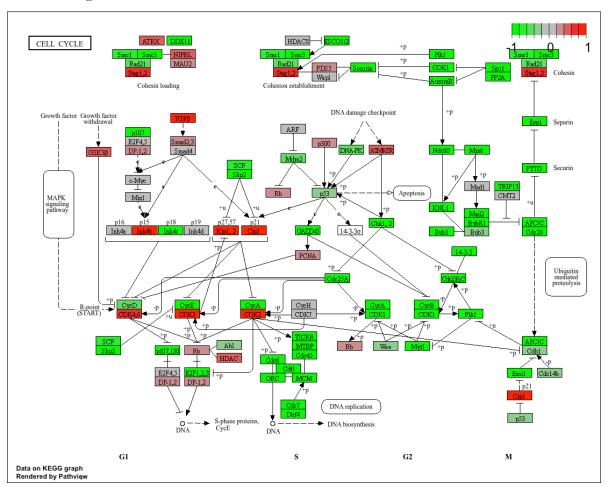
```
pathview(foldchanges, pathway.id = "hsa04110")
```

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa04110.pathview.png

Insert this figure:



Change the display in various ways including generating a PDF graph:

```
# A different PDF based output of the same data pathview(gene.data=foldchanges, pathway.id="hsa04110", kegg.native=FALSE)
```

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

Warning: reconcile groups sharing member nodes!

```
[,1] [,2]
[1,] "9" "300"
[2,] "9" "306"
```

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa04110.pathview.pdf

```
## Focus on top 5 upregulated pathways here for demo purposes only
keggrespathways <- rownames(keggres$greater)[1:5]

# Extract the 8 character long IDs part of each string
keggresids = substr(keggrespathways, start=1, stop=8)
keggresids</pre>
```

[1] "hsa04060" "hsa05323" "hsa05146" "hsa05332" "hsa04640"

```
pathview(gene.data=foldchanges, pathway.id=keggresids, species="hsa")
```

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa04060.pathview.png

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa05323.pathview.png

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa05146.pathview.png

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

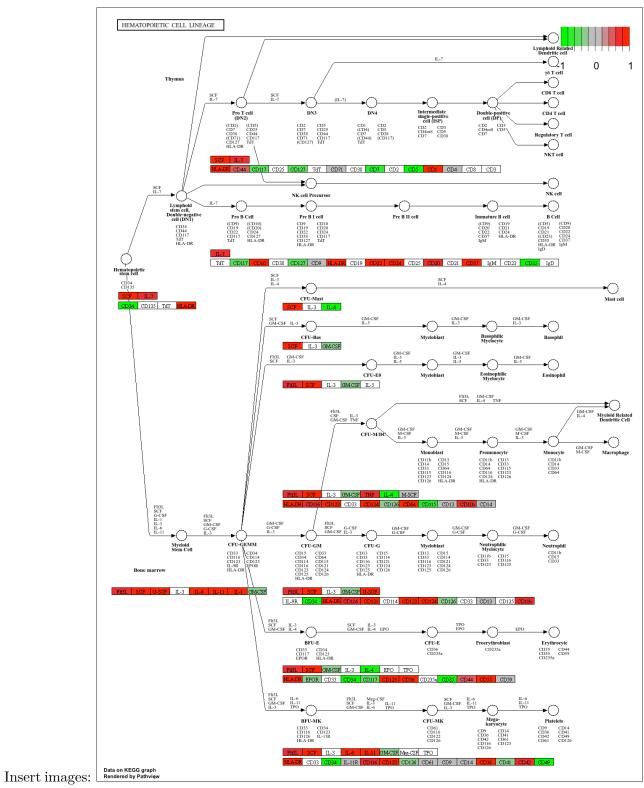
Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

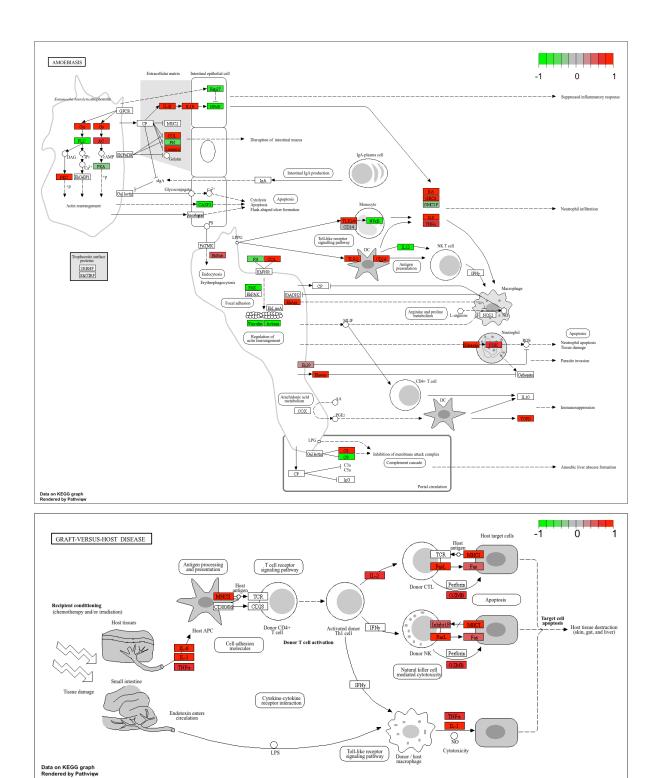
Info: Writing image file hsa05332.pathview.png

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa04640.pathview.png

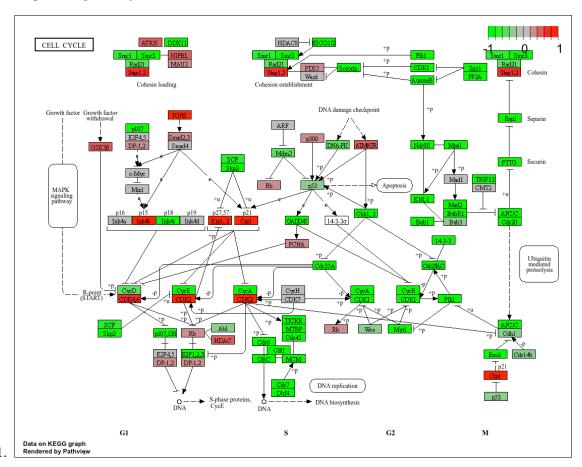




Q. Can you do the same procedure as above to plot the pathview figures for the

top 5 down-regulated pathways?

Down-regulated pathways:



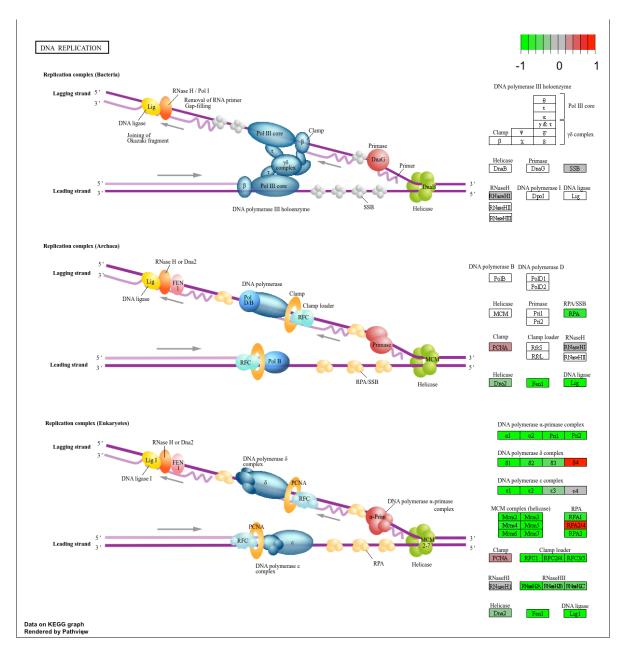
2.

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

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa03030.pathview.png



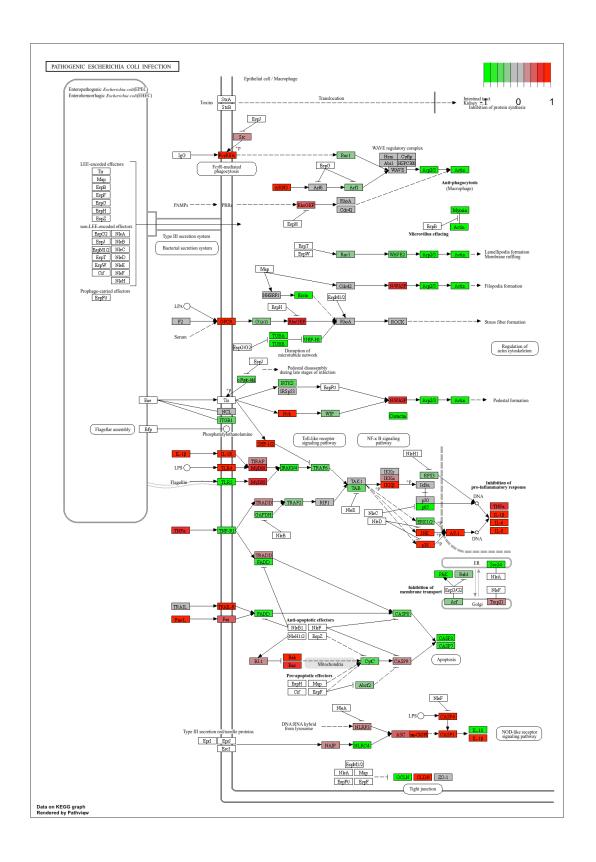
3.

pathview(foldchanges, pathway.id = "hsa05130")

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa05130.pathview.png



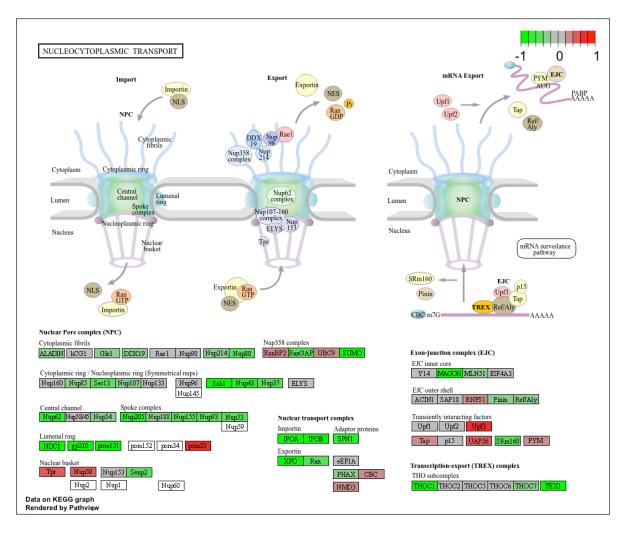
4.

pathview(foldchanges, pathway.id = "hsa03013")

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa03013.pathview.png

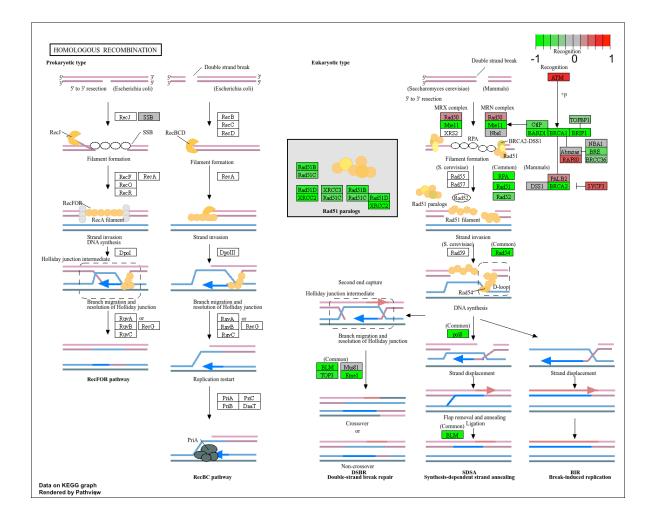


5.

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

Info: Working in directory /Users/ruthbarnes/Desktop/School/bimm143/Class 14

Info: Writing image file hsa03440.pathview.png



Gene Ontology Analysis

Run pathway analysis with GO

```
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, 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
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
                                                           84 1.729553e-10
                                        1.178402e-07
```