Class14: RNA-Seq analysis mini-project

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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

Loading required package: SummarizedExperiment

Loading required package: MatrixGenerics

Loading required package: matrixStats

Attaching package: 'MatrixGenerics'

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

colAlls, colAnyNAs, colAnys, colAvgsPerRowSet, 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, rowAvgsPerColSet, 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, rowWeightedMedians, rowWeightedMedians, rowWeightedMedians, 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
Data import. For our session we have used the Sailfish gene-level estimated counts.
  metaFile <- "GSE37704_metadata.csv"</pre>
  countFile <- "GSE37704 featurecounts.csv"</pre>
  # Import metadata and take a peak
  colData = read.csv(metaFile, row.names = 1)
  head(colData)
              condition
SRR493366 control_sirna
SRR493367 control_sirna
SRR493368 control_sirna
SRR493369
               hoxa1 kd
SRR493370
             hoxa1_kd
SRR493371
               hoxa1_kd
  countData = read.csv(countFile, row.names = 1)
  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	2	258				

We need to remove the first column from countData.

```
# Note we need to remove the odd first $length col
countData <- as.matrix(countData[,-1])
head(countData)</pre>
```

	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

```
# Filter count data where you have 0 read count across all samples.
countData = countData[rowSums(countData) != 0, ]
head(countData)
```

	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

Running DESeq2

Setting up the DESeqDataSet object that is required for the DESeq() function and then run the DESeq pipeline.

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

```
dds = DESeq(dds)

estimating size factors

estimating dispersions

gene-wise dispersion estimates

mean-dispersion relationship

final dispersion estimates

fitting model and testing
```

dds

```
class: DESeqDataSet
dim: 15975 6
metadata(1): version
assays(4): counts mu H cooks
rownames(15975): ENSG00000279457 ENSG00000187634 ... ENSG00000276345
   ENSG00000271254
rowData names(22): baseMean baseVar ... deviance maxCooks
colnames(6): SRR493366 SRR493367 ... SRR493370 SRR493371
colData names(2): condition sizeFactor
```

Get results for the HoxA1 knockdown versus control siRNA.Labeled as "hoxa1_kd" and "control_sirna" in our original colData metaFile input to DESeq.

```
res = results(dds, contrast=c("condition", "hoxa1_kd", "control_sirna"))
```

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.

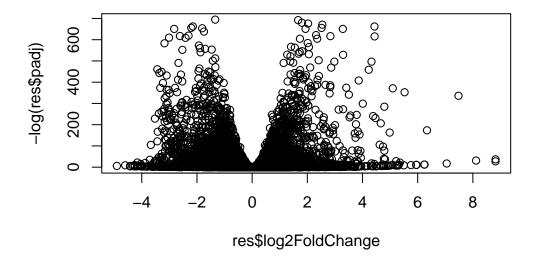
```
summary(res)
```

```
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>
```

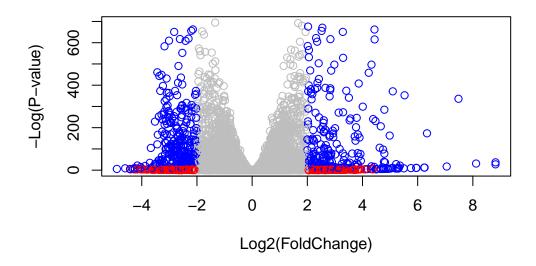
Volcano Plot

We will make a volcano plot to visually assess the data. A plot of log2 fold change vs -log adjusted p-value.

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



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



Adding gene annotation

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

```
library("AnnotationDbi")
Warning: package 'AnnotationDbi' was built under R version 4.3.2
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"
                                   "ONTOLOGYALL" "PATH"
                    "ONTOLOGY"
                                                                 "PFAM"
[21] "PMID"
                    "PROSITE"
                                   "REFSEQ"
                                                  "SYMBOL"
                                                                 "UCSCKG"
[26] "UNIPROT"
  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$genename = 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
                                                                     pvalue
                                                           stat
                  <numeric>
                                 <numeric> <numeric> <numeric>
                                                                  <numeric>
                                 0.1792571 0.3248216 0.551863 5.81042e-01
ENSG00000279457
                  29.913579
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
                 47.255123
ENSG00000187583
                                 0.0405765 0.2718928
                                                       0.149237 8.81366e-01
                  11.979750
                                 0.5428105 0.5215598 1.040744 2.97994e-01
ENSG00000187642
                                 2.0570638 0.1969053 10.446970 1.51282e-25
ENSG00000188290 108.922128
                                 0.2573837 0.1027266
                                                       2.505522 1.22271e-02
ENSG00000187608 350.716868
ENSG00000188157 9128.439422
                                 0.3899088 0.0467163
                                                      8.346304 7.04321e-17
ENSG00000237330
                   0.158192
                                 0.7859552 4.0804729
                                                       0.192614 8.47261e-01
                                 symbol
                       padj
                                             entrez
                                                                   genename
                  <numeric> <character> <character>
                                                               <character>
ENSG00000279457 6.86555e-01
                                     NA
                                                                         NA
                                                 NΑ
ENSG00000187634 5.15718e-03
                                             148398 sterile alpha motif ..
                                 SAMD11
ENSG00000188976 1.76549e-35
                                  NOC2L
                                              26155 NOC2 like nucleolar ...
ENSG00000187961 1.13413e-07
                                             339451 kelch like family me..
                                 KLHL17
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
                                               9636 ISG15 ubiquitin like..
                                  ISG15
ENSG00000188157 4.21963e-16
                                             375790
                                   AGRN
                                                                      agrin
ENSG00000237330
                         NA
                                 RNF223
                                             401934 ring finger protein ...
```

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")
```

Pathway Analysis

Here we are going to use the gage and pathview packages for pathway analysis.

```
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

library(gage)

```
library(gageData)
  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"
                        "1577"
              "1576"
                                 "1806"
                                          "1807"
                                                    "1890"
                                                             "221223" "2990"
[17] "3251"
              "3614"
                        "3615"
                                 "3704"
                                          "51733"
                                                    "54490"
                                                             "54575"
                                                                      "54576"
[25] "54577"
              "54578"
                       "54579"
                                 "54600"
                                          "54657"
                                                    "54658"
                                                             "54659"
                                                                      "54963"
                                                             "7364"
                                                                       "7365"
[33] "574537" "64816"
                        "7083"
                                 "7084"
                                          "7172"
                                                    "7363"
[41] "7366"
              "7367"
                        "7371"
                                 "7372"
                                          "7378"
                                                    "7498"
                                                             "79799"
                                                                      "83549"
[49] "8824"
              "8833"
                        "9"
                                 "978"
$`hsa00230 Purine metabolism`
  [1] "100"
               "10201"
                        "10606"
                                                     "10623"
                                                              "107"
                                                                       "10714"
                                  "10621"
                                           "10622"
  [9] "108"
               "10846"
                        "109"
                                  "111"
                                           "11128"
                                                     "11164"
                                                              "112"
                                                                       "113"
 [17] "114"
               "115"
                        "122481" "122622" "124583" "132"
                                                              "158"
                                                                       "159"
 [25] "1633"
               "171568" "1716"
                                  "196883" "203"
                                                     "204"
                                                              "205"
                                                                       "221823"
 [33] "2272"
               "22978"
                        "23649"
                                  "246721" "25885"
                                                     "2618"
                                                              "26289"
                                                                       "270"
                                  "2766"
                        "272"
                                           "2977"
                                                     "2982"
                                                              "2983"
 [41] "271"
               "27115"
                                                                       "2984"
 [49] "2986"
               "2987"
                        "29922"
                                  "3000"
                                           "30833"
                                                     "30834"
                                                              "318"
                                                                       "3251"
```

```
"4830"
[57] "353"
               "3614"
                         "3615"
                                  "3704"
                                            "377841" "471"
                                                                        "4831"
 [65] "4832"
               "4833"
                         "4860"
                                  "4881"
                                            "4882"
                                                     "4907"
                                                               "50484"
                                                                        "50940"
                                                                        "5140"
                                            "5137"
                                                               "5139"
 [73] "51082"
               "51251"
                         "51292"
                                  "5136"
                                                     "5138"
[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"
[105] "5424"
               "5425"
                         "5426"
                                  "5427"
                                            "5430"
                                                     "5431"
                                                               "5432"
                                                                        "5433"
                                  "5437"
                                                     "5439"
                                                               "5440"
[113] "5434"
               "5435"
                         "5436"
                                            "5438"
                                                                        "5441"
[121] "5471"
               "548644" "55276"
                                  "5557"
                                            "5558"
                                                     "55703"
                                                               "55811"
                                                                        "55821"
                         "56655"
[129] "5631"
               "5634"
                                  "56953"
                                                     "57804"
                                                               "58497"
                                                                        "6240"
                                            "56985"
[137] "6241"
               "64425"
                         "646625" "654364"
                                            "661"
                                                     "7498"
                                                               "8382"
                                                                        "84172"
[145] "84265"
               "84284"
                         "84618"
                                  "8622"
                                            "8654"
                                                     "87178"
                                                               "8833"
                                                                        "9060"
                                                     "955"
[153] "9061"
                         "953"
                                  "9533"
                                            "954"
                                                               "956"
                                                                        "957"
               "93034"
[161] "9583"
               "9615"
```

The main gage() function requires a named vector of fold changes, where the names of the values are the Entrez gene IDs.

We will use mapIDs() function for this.

```
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
```

Now we can run the gage pathway analysis.

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

Look at the results.

```
attributes(keggres)

$names
[1] "greater" "less" "stats"
```

Look at the first few down (less) pathways head(keggres\$less)

```
p.geomean stat.mean
                                                                  p.val
hsa04110 Cell cycle
                                    8.995727e-06 -4.378644 8.995727e-06
hsa03030 DNA replication
                                     9.424076e-05 -3.951803 9.424076e-05
hsa03013 RNA transport
                                     1.375901e-03 -3.028500 1.375901e-03
hsa03440 Homologous recombination
                                     3.066756e-03 -2.852899 3.066756e-03
hsa04114 Oocyte meiosis
                                     3.784520e-03 -2.698128 3.784520e-03
hsa00010 Glycolysis / Gluconeogenesis 8.961413e-03 -2.405398 8.961413e-03
                                           q.val set.size
                                                                 exp1
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
                                                     28 3.066756e-03
                                     0.121861535
                                     0.121861535 102 3.784520e-03
hsa04114 Oocyte meiosis
hsa00010 Glycolysis / Gluconeogenesis 0.212222694
                                                     53 8.961413e-03
```

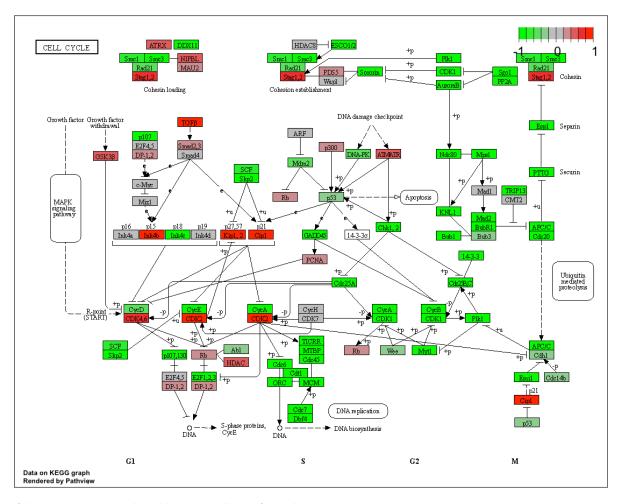
Using the pathview() function to make a pathway plot with our RNA-Seq expression results shown in color. We'll manually supply a pathway.id ("hsa04110 Cell cycle") that we could see from above.

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

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

Info: Working in directory /Users/aishamohamed/Documents/BIMM 143/Class14

Info: Writing image file hsa04110.pathview.png



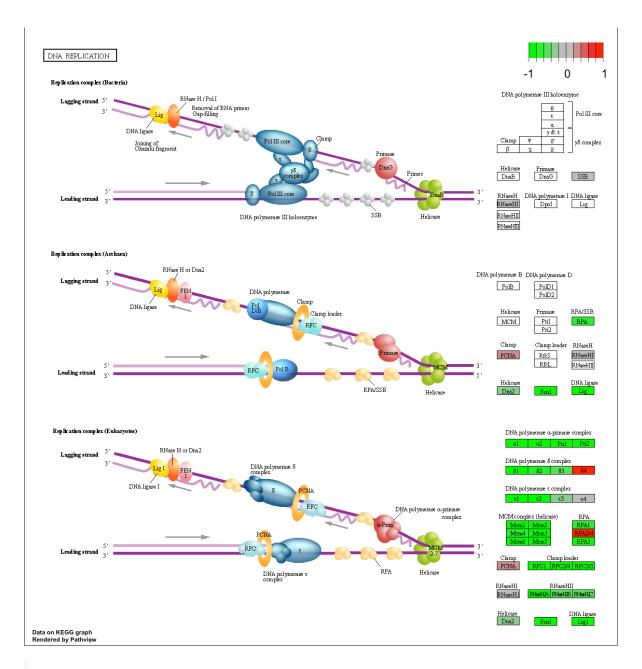
Also want to visualize "hsa03030" DNA replication.

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

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

Info: Working in directory /Users/aishamohamed/Documents/BIMM 143/Class14

Info: Writing image file hsa03030.pathview.png



head(keggres\$greater)

p.geomean stat.mean p.val
hsa04640 Hematopoietic cell lineage 0.002822776 2.833362 0.002822776
hsa04630 Jak-STAT signaling pathway 0.005202070 2.585673 0.005202070
hsa00140 Steroid hormone biosynthesis 0.007255099 2.526744 0.007255099
hsa04142 Lysosome 0.010107392 2.338364 0.010107392

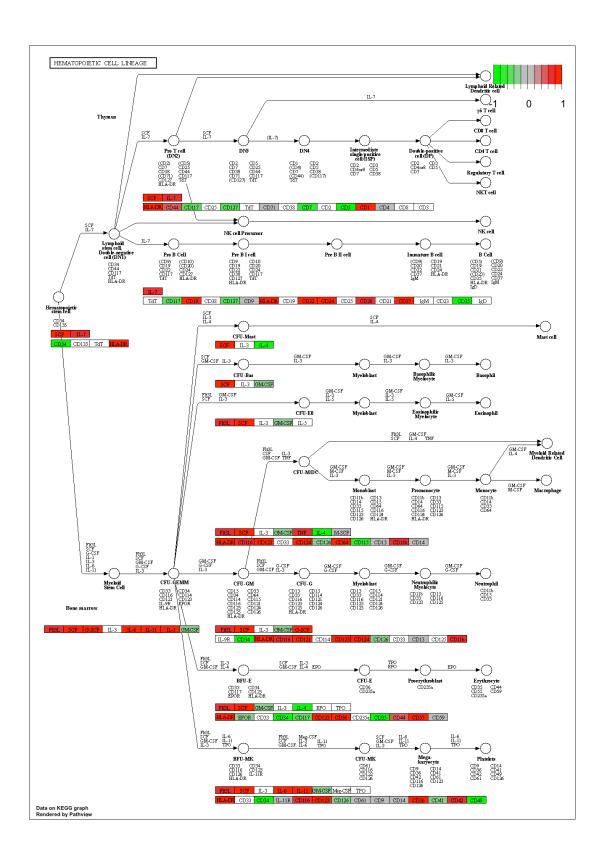
```
hsa04330 Notch signaling pathway
                                    0.018747253 2.111725 0.018747253
hsa04916 Melanogenesis
                                    0.019399766 2.081927 0.019399766
                                        q.val set.size
                                                             exp1
hsa04640 Hematopoietic cell lineage
                                    0.3893570
                                                  55 0.002822776
hsa04630 Jak-STAT signaling pathway
                                    0.3893570
                                                 109 0.005202070
hsa00140 Steroid hormone biosynthesis 0.3893570
                                                  31 0.007255099
hsa04142 Lysosome
                                    0.4068225
                                                 118 0.010107392
                                                  46 0.018747253
hsa04330 Notch signaling pathway
                                    0.4391731
hsa04916 Melanogenesis
                                    0.4391731
                                                  90 0.019399766
```

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

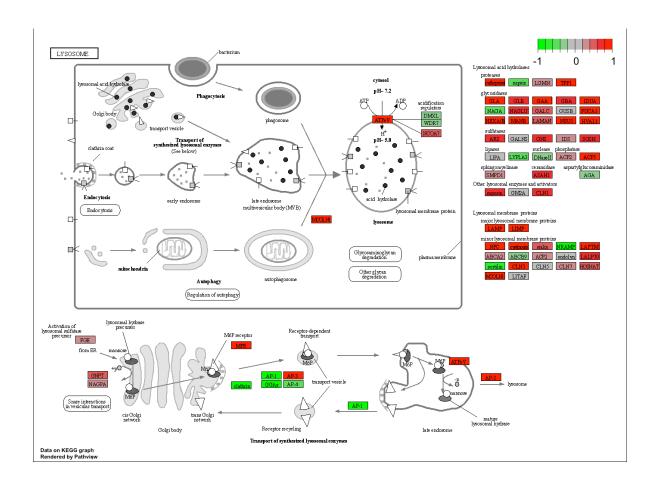
# Extract the 8 character long IDs part of each string
keggresids = substr(keggrespathways, start=1, stop=8)
keggresids</pre>
```

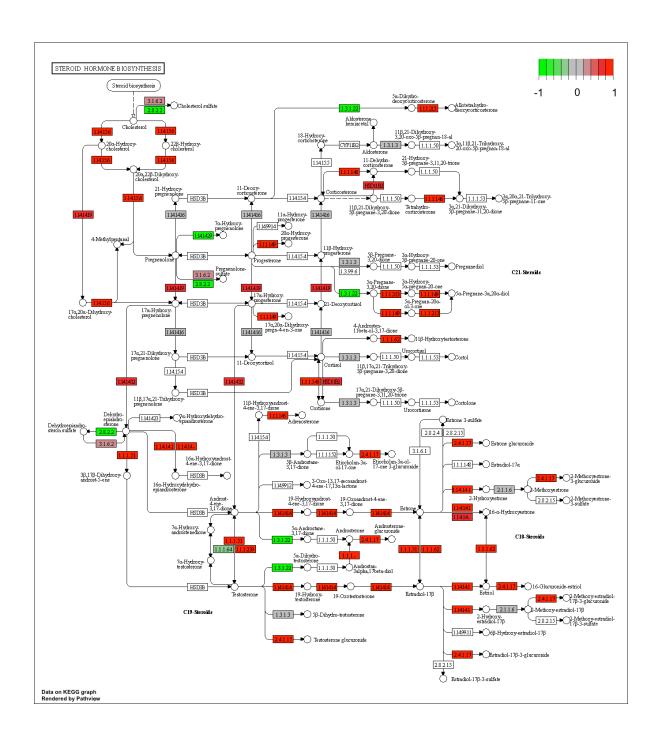
[1] "hsa04640" "hsa04630" "hsa00140" "hsa04142" "hsa04330"

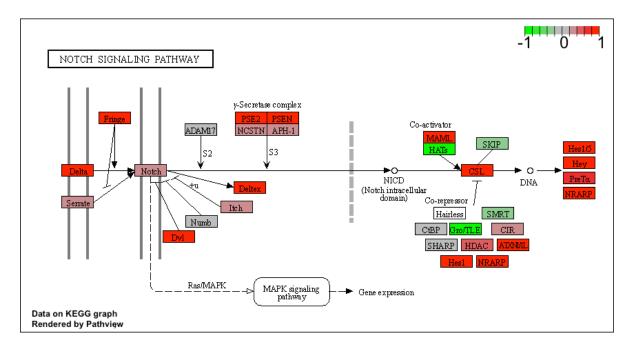
Upregulated pathways (Top 5):











Q. Can you do the same procedure as above to plot the pathview figures for the top 5 down-reguled pathways?

```
## Focus on top 5 down-regulated pathways
keggrespathways.less <- rownames(keggres$less)[1:5]

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

[1] "hsa04110" "hsa03030" "hsa03013" "hsa03440" "hsa04114"

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

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

Info: Working in directory /Users/aishamohamed/Documents/BIMM 143/Class14

Info: Writing image file hsa04110.pathview.png

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

Info: Working in directory /Users/aishamohamed/Documents/BIMM 143/Class14

Info: Writing image file hsa03030.pathview.png

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

Info: Working in directory /Users/aishamohamed/Documents/BIMM 143/Class14

Info: Writing image file hsa03013.pathview.png

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

Info: Working in directory /Users/aishamohamed/Documents/BIMM 143/Class14

Info: Writing image file hsa03440.pathview.png

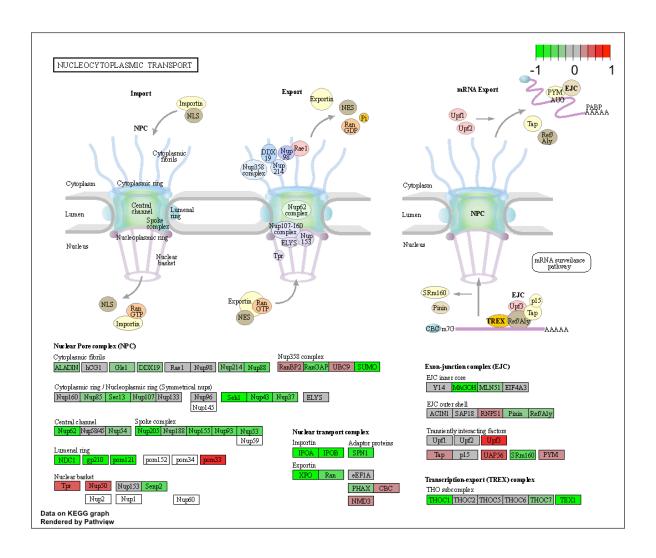
'select()' returned 1:1 mapping between keys and columns

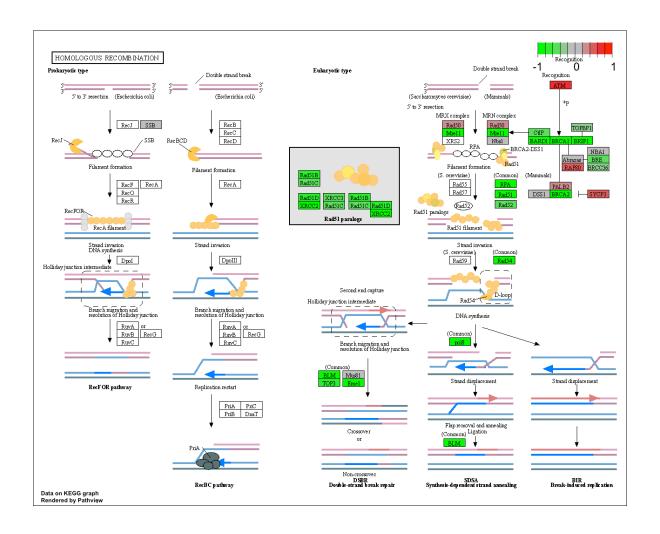
Info: Working in directory /Users/aishamohamed/Documents/BIMM 143/Class14

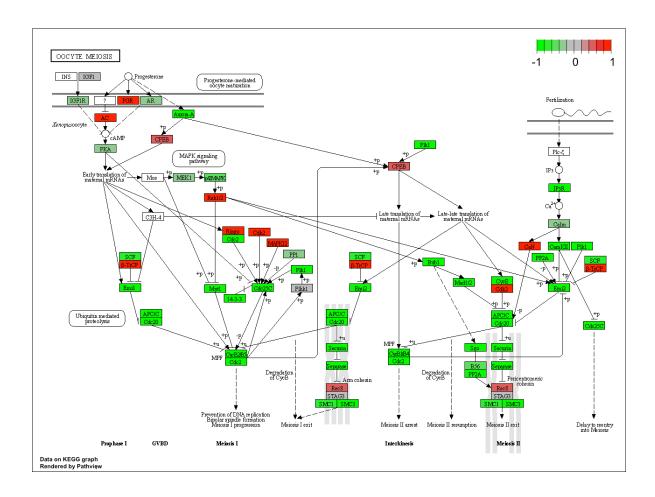
Info: Writing image file hsa04114.pathview.png

Since "hsa04110" and "hsa03030" have already been visualized up above, only the 3 other IDs will be printed out.

Downregulated Pathways:







Gene Ontology (G0)

We can also do a similar procedure with gene ontology. Similar to above, go.sets.hs has all GO terms. go.subs.hs is a named list containing indexes for the BP, CC, and MF ontologies. Let's focus on BP (a.k.a Biological Process) here.

```
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)
```

lapply(gobpres, head)

\$greater

φβιεατεί			
	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 se	t.size	exp1
GO:0007156 homophilic cell adhesion	0.1952430	113 8.5	19724e-05
GO:0002009 morphogenesis of an epithelium	0.1952430	339 1.39	96681e-04
GO:0048729 tissue morphogenesis	0.1952430	424 1.43	32451e-04
GD:0007610 behavior	0.1968058	426 1.92	25222e-04
GO:0060562 epithelial tube morphogenesis	0.3566193	257 5.93	32837e-04
GO:0035295 tube development	0.3566193	391 5.9	53254e-04
\$less			
	p.geomean	stat.mean	p.val
GO:0048285 organelle fission	1.536227e-15		-
•	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
-	2.028624e-11		
	1.729553e-10		
1		set.size	
GO:0048285 organelle fission	5.843127e-12		.536227e-15
_	5.843127e-12		.286961e-15
	5.843127e-12		.286961e-15
GO:0000087 M phase of mitotic cell cycle		362 1	.169934e-14
-	1.659009e-08		.028624e-11
3 3	1.178690e-07	84 1	.729553e-10
reconstruction for the second			
\$stats			
450405	stat.mean	exp1	
GO:0007156 homophilic cell adhesion	3.824205 3.8	-	
GO:0002009 morphogenesis of an epithelium			
GO:0048729 tissue morphogenesis	3.643242 3.0		
G0:0007610 behavior	3.565432 3.5		
G0:0060562 epithelial tube morphogenesis	3.261376 3.3		
G0:0035295 tube development	3.253665 3.3		
do.0000200 tube development	0.200000 0	200000	

Reactome Analysis

Reactome is database consisting of biological molecules and their relation to pathways and processes.

Output the list of significant genes at the 0.05 level

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