

Class12

Dennis Kim

Already submitted the first part but lost the r project file, time to catch up on some old data lost

```
library(BiocManager)
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 objects are masked from 'package:base':

expand.grid, I, unname

Loading required package: IRanges

Attaching package: 'IRanges'

The following object is masked from 'package:grDevices':

windows

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,  
rowWeightedSds, 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
```

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

```
library(DESeq2)
```

```
dds <- DESeqDataSetFromMatrix(countData=counts,  
                              colData=metadata,  
                              design=~dex)
```

converting counts to integer mode

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

```
dds
```

```
class: DESeqDataSet
dim: 38694 8
metadata(1): version
assays(1): counts
rownames(38694): ENSG000000000003 ENSG000000000005 ... ENSG00000283120
               ENSG00000283123
rowData names(0):
colnames(8): SRR1039508 SRR1039509 ... SRR1039520 SRR1039521
colData names(4): id dex celltype geo_id
```

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

estimating size factors

estimating dispersions

gene-wise dispersion estimates

mean-dispersion relationship

final dispersion estimates

fitting model and testing

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

log2 fold change (MLE): dex treated vs control

Wald test p-value: dex treated vs control

DataFrame with 38694 rows and 6 columns

	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG000000000003	747.1942	-0.3507030	0.168246	-2.084470	0.0371175
ENSG000000000005	0.0000	NA	NA	NA	NA
ENSG000000000419	520.1342	0.2061078	0.101059	2.039475	0.0414026

```

ENSG00000000457 322.6648 0.0245269 0.145145 0.168982 0.8658106
ENSG00000000460 87.6826 -0.1471420 0.257007 -0.572521 0.5669691
...
ENSG00000283115 0.000000 NA NA NA NA
ENSG00000283116 0.000000 NA NA NA NA
ENSG00000283119 0.000000 NA NA NA NA
ENSG00000283120 0.974916 -0.668258 1.69456 -0.394354 0.693319
ENSG00000283123 0.000000 NA NA NA NA
      padj
<numeric>
ENSG00000000003 0.163035
ENSG00000000005 NA
ENSG00000000419 0.176032
ENSG00000000457 0.961694
ENSG00000000460 0.815849
...
ENSG00000283115 NA
ENSG00000283116 NA
ENSG00000283119 NA
ENSG00000283120 NA
ENSG00000283123 NA

```

```
summary(res)
```

```

out of 25258 with nonzero total read count
adjusted p-value < 0.1
LFC > 0 (up)      : 1563, 6.2%
LFC < 0 (down)    : 1188, 4.7%
outliers [1]      : 142, 0.56%
low counts [2]    : 9971, 39%
(mean count < 10)
[1] see 'cooksCutoff' argument of ?results
[2] see 'independentFiltering' argument of ?results

```

```

res05 <- results(dds, alpha=0.05)
summary(res05)

```

```
out of 25258 with nonzero total read count
```

```

adjusted p-value < 0.05
LFC > 0 (up)      : 1236, 4.9%
LFC < 0 (down)    : 933, 3.7%
outliers [1]      : 142, 0.56%
low counts [2]    : 9033, 36%
(mean count < 6)
[1] see 'cooksCutoff' argument of ?results
[2] see 'independentFiltering' argument of ?results

```

done catching up, time to work on the new lab

```

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

```

We can use the `mapIds()` function to add individual columns to our results table. We provide the row names of our results table as a key, and specify that `keytype=ENSEMBL`. The `column` argument tells the `mapIds()` function which information we want, and the `multiVals` argument tells the function what to do if there are multiple possible values for a single input value. Here we ask to just give us back the first one that occurs in the database.

Q11. Run the `mapIds()` function two more times to add the Entrez ID and UniProt accession and GENENAME as new columns called `res$entrez`, `res$uniprot` and `res$genename`.

```

res$symbol <- mapIds(org.Hs.eg.db,
  keys=row.names(res), # Our genenames
  keytype="ENSEMBL",   # The format of our genenames
  column="SYMBOL",     # The new format we want to add
  multiVals="first")

```

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

```
res$entrez <- mapIds(org.Hs.eg.db,
                     keys=row.names(res), # Our genenames
                     keytype="ENSEMBL",   # The format of our genenames
                     column="ENTREZID",   # The new format we want to add
                     multiVals="first")
```

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

```
res$uniprot <- mapIds(org.Hs.eg.db,
                     keys=row.names(res), # Our genenames
                     keytype="ENSEMBL",   # The format of our genenames
                     column="UNIPROT",     # The new format we want to add
                     multiVals="first")
```

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

```
res$genename <- mapIds(org.Hs.eg.db,
                      keys=row.names(res), # Our genenames
                      keytype="ENSEMBL",   # The format of our genenames
                      column="GENENAME",    # The new format we want to add
                      multiVals="first")
```

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

```
head(res)
```

log2 fold change (MLE): dex treated vs control

Wald test p-value: dex treated vs control

DataFrame with 6 rows and 10 columns

	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG000000000003	747.194195	-0.3507030	0.168246	-2.084470	0.0371175
ENSG000000000005	0.000000	NA	NA	NA	NA
ENSG000000000419	520.134160	0.2061078	0.101059	2.039475	0.0414026
ENSG000000000457	322.664844	0.0245269	0.145145	0.168982	0.8658106

	padj	symbol	entrez	uniprot
	<numeric>	<character>	<character>	<character>
ENSG000000000460	87.682625	-0.1471420	0.257007	-0.572521 0.5669691
ENSG000000000938	0.319167	-1.7322890	3.493601	-0.495846 0.6200029
ENSG000000000003	0.163035	TSPAN6	7105	AOA024RCIO
ENSG000000000005	NA	TNMD	64102	Q9H2S6
ENSG000000000419	0.176032	DPM1	8813	O60762
ENSG000000000457	0.961694	SCYL3	57147	Q8IZE3
ENSG000000000460	0.815849	C1orf112	55732	AOA024R922
ENSG000000000938	NA	FGR	2268	P09769
	genename			
	<character>			
ENSG000000000003	tetraspanin 6			
ENSG000000000005	tenomodulin			
ENSG000000000419	dolichyl-phosphate m..			
ENSG000000000457	SCY1 like pseudokina..			
ENSG000000000460	chromosome 1 open re..			
ENSG000000000938	FGR proto-oncogene, ..			

```
ord <- order( res$padj )
#View(res[ord,])
head(res[ord,])
```

log2 fold change (MLE): dex treated vs control
Wald test p-value: dex treated vs control
DataFrame with 6 rows and 10 columns

	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG00000152583	954.771	4.36836	0.2371268	18.4220	8.74490e-76
ENSG00000179094	743.253	2.86389	0.1755693	16.3120	8.10784e-60
ENSG00000116584	2277.913	-1.03470	0.0650984	-15.8944	6.92855e-57
ENSG00000189221	2383.754	3.34154	0.2124058	15.7319	9.14433e-56
ENSG00000120129	3440.704	2.96521	0.2036951	14.5571	5.26424e-48
ENSG00000148175	13493.920	1.42717	0.1003890	14.2164	7.25128e-46
	padj	symbol	entrez	uniprot	
	<numeric>	<character>	<character>	<character>	
ENSG00000152583	1.32441e-71	SPARCL1	8404	AOA024RDE1	
ENSG00000179094	6.13966e-56	PER1	5187	O15534	
ENSG00000116584	3.49776e-53	ARHGEF2	9181	Q92974	
ENSG00000189221	3.46227e-52	MAOA	4128	P21397	
ENSG00000120129	1.59454e-44	DUSP1	1843	B4DU40	


```

ENSG00000148175 1.83034e-42      STOM      2040      F8VSL7
                        genename
                        <character>
ENSG00000152583      SPARC like 1
ENSG00000179094 period circadian reg..
ENSG00000116584 Rho/Rac guanine nucl..
ENSG00000189221      monoamine oxidase A
ENSG00000120129 dual specificity pho..
ENSG00000148175      stomatin

```

```
write.csv(res[ord,], "deseq_results.csv")
```

Now we can load the packages and setup the KEGG data-sets we need. The gageData package has pre-compiled databases mapping genes to KEGG pathways and GO terms for common organisms. kegg.sets.hs is a named list of 229 elements. Each element is a character vector of member gene Entrez IDs for a single KEGG pathway.

```
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 license agreement (details at <http://www.kegg.jp/kegg/legal.html>).

```
#####
```

```
library(gage)
```

```
library(gageData)
```

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

```
# Examine the first 2 pathways in this kegg set for humans
```

```
head(kegg.sets.hs, 2)
```

```
$`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" "54490" "54575" "54576"
[25] "54577" "54578" "54579" "54600" "54657" "54658" "54659" "54963"
[33] "574537" "64816" "7083" "7084" "7172" "7363" "7364" "7365"
[41] "7366" "7367" "7371" "7372" "7378" "7498" "79799" "83549"
[49] "8824" "8833" "9" "978"
```

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

```
7105 64102 8813 57147 55732 2268
-0.35070302 NA 0.20610777 0.02452695 -0.14714205 -1.73228897
```

Run gage pathway analysis

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

```
$names
```

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

```
# Look at the first three down (less) pathways
head(keggres$less, 3)
```

	p.geomean	stat.mean	p.val
hsa05332 Graft-versus-host disease	0.0004250461	-3.473346	0.0004250461
hsa04940 Type I diabetes mellitus	0.0017820293	-3.002352	0.0017820293
hsa05310 Asthma	0.0020045888	-3.009050	0.0020045888
	q.val	set.size	exp1

hsa05332 Graft-versus-host disease	0.09053483	40	0.0004250461
hsa04940 Type I diabetes mellitus	0.14232581	42	0.0017820293
hsa05310 Asthma	0.14232581	29	0.0020045888

Let's pull up the highlighted pathways and show our differentially expressed genes on the pathway. I will use the "hsa" KEGG id to get the pathway from KEGG and my 'foldchange' vector to show my genes.

```
pathview(gene.data=foldchanges, pathway.id="hsa05310")
```

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

Info: Working in directory C:/Users/denjk/OneDrive/Desktop/BioInformatics Lab/Class12/Class1

Info: Writing image file hsa05310.pathview.png

Put this image into my document

