

class 14

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Data Import

Data Tidying

DESeq setup and analysis

Add annotation data

Save my results

Visualization

```
library(BiocManager)
```

Bioconductor version '3.17' is out-of-date; the current release version '3.18' is available with R version '4.3'; see <https://bioconductor.org/install>

```
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,
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
```

```
metaFile <- "GSE37704_metadata.csv"
countFile <- "GSE37704_featurecounts.csv"
# 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
```

```
counts<-read.csv("GSE37704_featurecounts.csv", row.names=1)
metadata<-read.csv("GSE37704_metadata.csv")
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
SRR493371						
ENSG00000186092	0					
ENSG00000279928	0					
ENSG00000279457	46					
ENSG00000278566	0					
ENSG00000273547	0					
ENSG00000187634	258					

```
metadata
```

	id	condition
1	SRR493366	control_sirna
2	SRR493367	control_sirna
3	SRR493368	control_sirna
4	SRR493369	hoxa1_kd
5	SRR493370	hoxa1_kd
6	SRR493371	hoxa1_kd

```
# Import countdata
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
	SRR493371					
ENSG00000186092	0					
ENSG00000279928	0					
ENSG00000279457	46					
ENSG00000278566	0					
ENSG00000273547	0					
ENSG00000187634	258					

Q. Complete the code below to remove the troublesome first column from countData

```
counts<-countData[,-1]
all(colnames(counts) == metadata$id)
```

[1] TRUE

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).

Tip: What will rowSums() of countData return and how could you use it in this context?

```
nrow(countData)
```

[1] 19808

```
to.rm.ind<-rowSums(counts) == 0
counts<-counts[!to.rm.ind,]

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

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

```
dds
```

```
class: DESeqDataSet
dim: 15975 6
metadata(1): version
assays(1): counts
rownames(15975): ENSG00000279457 ENSG00000187634 ... ENSG00000276345
               ENSG00000271254
rowData names(0):
colnames(6): SRR493366 SRR493367 ... SRR493370 SRR493371
colData names(2): id condition
```

```
summary(dds)
```

```
[1] "DESeqDataSet object of length 15975 with 0 metadata columns"
```

```
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)
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

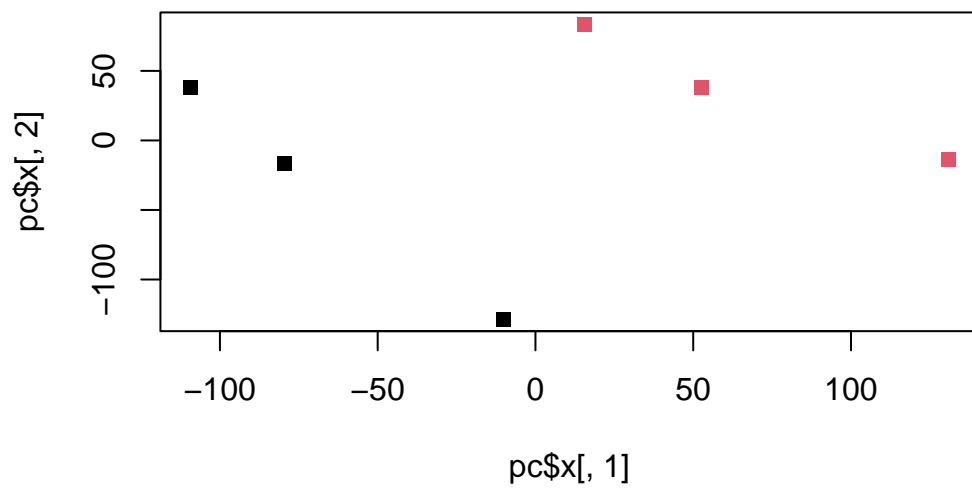
	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG00000279457	29.9136	0.1792571	0.3248216	0.551863	5.81042e-01
ENSG00000187634	183.2296	0.4264571	0.1402658	3.040350	2.36304e-03
ENSG00000188976	1651.1881	-0.6927205	0.0548465	-12.630158	1.43990e-36
ENSG00000187961	209.6379	0.7297556	0.1318599	5.534326	3.12428e-08
ENSG00000187583	47.2551	0.0405765	0.2718928	0.149237	8.81366e-01
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				

```
pc <- prcomp(t(counts), scale=T)
summary(pc)
```

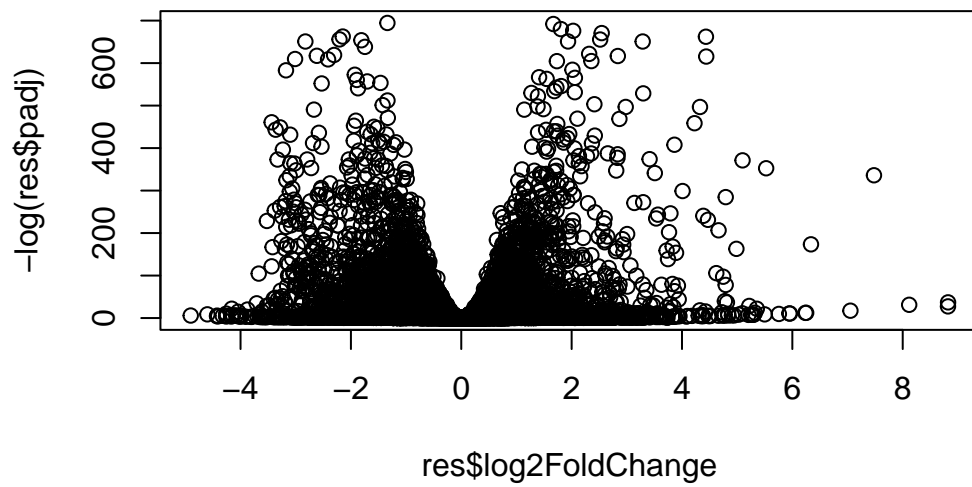
Importance of components:

	PC1	PC2	PC3	PC4	PC5	PC6
Standard deviation	87.7211	73.3196	32.89604	31.15094	29.18417	7.373e-13
Proportion of Variance	0.4817	0.3365	0.06774	0.06074	0.05332	0.000e+00
Cumulative Proportion	0.4817	0.8182	0.88594	0.94668	1.00000	1.000e+00

```
plot(pc$x[,1], pc$x[,2], col=as.factor(metadata$condition), pch=15)
```

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



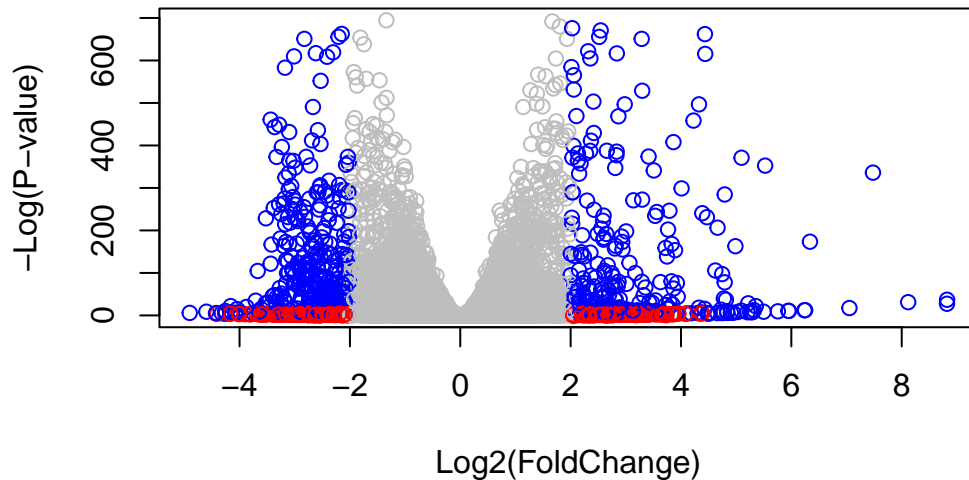
Q. Improve this plot by completing the below code, which adds color and axis labels

```
# Make a color vector for all genes
mycols <- rep("gray", nrow(res) )

# Color red the genes with absolute fold change above 2
mycols[ abs(res$log2FoldChange) > 2 ] <- "red"

# Color blue those with adjusted p-value less than 0.01
# and absolute fold change more than 2
inds <- (res$padj < 0.01) & (abs(res$log2FoldChange) > 2 )
mycols[ inds ] <- "blue"

plot( res$log2FoldChange, -log(res$padj), col=mycols, xlab="Log2(FoldChange)", ylab="-Log(
```



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

```
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"       "UCSCKG"
[26] "UNIPROT"
```

```
res$symbol = mapIds(org.Hs.eg.db,
  keys=row.names(counts),
  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(counts),
  keytype="ENSEMBL",
  column="ENTREZID",
  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 8 columns

	baseMean	log2FoldChange	lfcSE	stat	pvalue
	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
ENSG00000279457	29.913579	0.1792571	0.3248216	0.551863	5.81042e-01
ENSG00000187634	183.229650	0.4264571	0.1402658	3.040350	2.36304e-03
ENSG00000188976	1651.188076	-0.6927205	0.0548465	-12.630158	1.43990e-36
ENSG00000187961	209.637938	0.7297556	0.1318599	5.534326	3.12428e-08
ENSG00000187583	47.255123	0.0405765	0.2718928	0.149237	8.81366e-01

ENSG00000187642	11.979750	0.5428105	0.5215598	1.040744	2.97994e-01
ENSG00000188290	108.922128	2.0570638	0.1969053	10.446970	1.51282e-25
ENSG00000187608	350.716868	0.2573837	0.1027266	2.505522	1.22271e-02
ENSG00000188157	9128.439422	0.3899088	0.0467163	8.346304	7.04321e-17
ENSG00000237330	0.158192	0.7859552	4.0804729	0.192614	8.47261e-01
	padj	symbol	entrez		
	<numeric>	<character>	<character>		
ENSG00000279457	6.86555e-01	NA	NA		
ENSG00000187634	5.15718e-03	SAMD11	148398		
ENSG00000188976	1.76549e-35	NOC2L	26155		
ENSG00000187961	1.13413e-07	KLHL17	339451		
ENSG00000187583	9.19031e-01	PLEKHN1	84069		
ENSG00000187642	4.03379e-01	PERM1	84808		
ENSG00000188290	1.30538e-24	HES4	57801		
ENSG00000187608	2.37452e-02	ISG15	9636		
ENSG00000188157	4.21963e-16	AGRN	375790		
ENSG00000237330	NA	RNF223	401934		

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.

```
#|message: false
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.
```

```
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).
```

```
#####
```

```

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

```

```

      <NA>      148398      26155      339451      84069      84808
0.17925708 0.42645712 -0.69272046 0.72975561 0.04057653 0.54281049

```

```

# Get the results
data(kegg.sets.hs)
data(sigmet.idx.hs)
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	1.246882e-03	-3.059466
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	1.246882e-03	0.065461279
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	1.246882e-03
hsa03440 Homologous recombination	28	3.066756e-03
hsa04114 Oocyte meiosis	102	3.784520e-03

```

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

```

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


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

Info: Working in directory /Users/kalodiahtoma/Desktop/phd 2023/bggn 213/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
```

```
[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/kalodiahtoma/Desktop/phd 2023/bggn 213/class 14

Info: Writing image file hsa04060.pathview.png

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

Info: Working in directory /Users/kalodiahtoma/Desktop/phd 2023/bggn 213/class 14

Info: Writing image file hsa05323.pathview.png

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

Info: Working in directory /Users/kalodiahtoma/Desktop/phd 2023/bggn 213/class 14

Info: Writing image file hsa05146.pathview.png

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

```
Info: Working in directory /Users/kalodiahtoma/Desktop/phd 2023/bggg 213/class 14
```

```
Info: Writing image file hsa05332.pathview.png
```

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

```
Info: Working in directory /Users/kalodiahtoma/Desktop/phd 2023/bggg 213/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?

Gene Ontology

```
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	expl
GO:0048285 organelle fission	5.841698e-12	376	1.536227e-15
GO:0000280 nuclear division	5.841698e-12	352	4.286961e-15
GO:0007067 mitosis	5.841698e-12	352	4.286961e-15
GO:0000087 M phase of mitotic cell cycle	1.195672e-11	362	1.169934e-14
GO:0007059 chromosome segregation	1.658603e-08	142	2.028624e-11
GO:0000236 mitotic prometaphase	1.178402e-07	84	1.729553e-10

Reactome

We will use the online version of Reactome. It wants a list of your genes. We will write it out from R here.

```
sig_genes <- res[res$padj <= 0.05 & !is.na(res$padj), "symbol"]
write.table(sig_genes,
            file="significant_genes.txt",
            row.names=FALSE, col.names=FALSE, quote=FALSE)
print(paste("Total number of significant genes:", length(sig_genes)))
```

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

