

RNAseqReanalysis01132021a

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Introduction

Bladder cancer

Library for rstudio

Setup

setup working directory

```
setwd("/Users/xinqiaozhang/Desktop/PRJNA382834/01142021")
listMarts()
```

```
## Ensembl site unresponsive, trying asia mirror
```

```
##           biomart           version
## 1 ENSEMBL_MART_ENSEMBL      Ensembl Genes 102
## 2  ENSEMBL_MART_MOUSE       Mouse strains 102
## 3   ENSEMBL_MART_SNP       Ensembl Variation 102
## 4 ENSEMBL_MART_FUNCGEN     Ensembl Regulation 102
```

```
if(interactive()){
  mart <- useEnsembl("ensembl")
  humanmart <- useEnsembl(biomart = "ensembl", mirror = "useast")
}
humanmart = useEnsembl(biomart = "ensembl", dataset = "hsapiens_gene_ensembl", mirror = "useast")
```

Input data

```
counts <- read.csv("PRJNA382834strandnocount.csv", stringsAsFactors = FALSE)
counts <- counts[, c(1:6, 8,9,13)]
counts <- data.frame(counts[,-1], row.names = counts[,1])
head(counts, n=6)
```

```
##           HT1197 HT1376  J82  T24 x253JP RT112  RT4  UC3
## ENSG00000000003  1460   740 1421 1011  2530  3650  7688 1250
## ENSG00000000005    0     0   0   0     0     0     0     0
```

```
## ENSG00000000457      414      567      294      477      391      677      709      405
## ENSG00000000460      794     1842      903     1182      879     1449      609      731
## ENSG00000000938         1         5         2         0         1         27         81         0
## ENSG00000000971         5         22         52         23         63     1409    16905      293
```

```
samples <- read.csv("conditionPRJNA382834strandno.csv", stringsAsFactors = FALSE)
samples <- samples[1:8,]
samples <- data.frame(samples[,-1], row.names = samples[,1])
head(samples, n=8)
```

```
##           condition replicate
## HT1197          R           1
## HT1376          R           2
## J82             R           3
## T24            R           4
## x253JP          S           1
## x5637          S           2
## RT112          S           3
## RT4            S           4
```

```
colnames(counts) <- c(rownames(samples))
all(rownames(samples) == colnames(counts))
```

```
## [1] TRUE
```

DESeq2 analysis

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

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

```
dds <- estimateSizeFactors(dds)
sizeFactors(dds)
```

```
##      HT1197      HT1376      J82      T24      x253JP      x5637      RT112      RT4
## 0.9845122 1.0230076 0.9010706 1.1620333 1.1918203 0.9425694 0.9545667 1.0372552
```

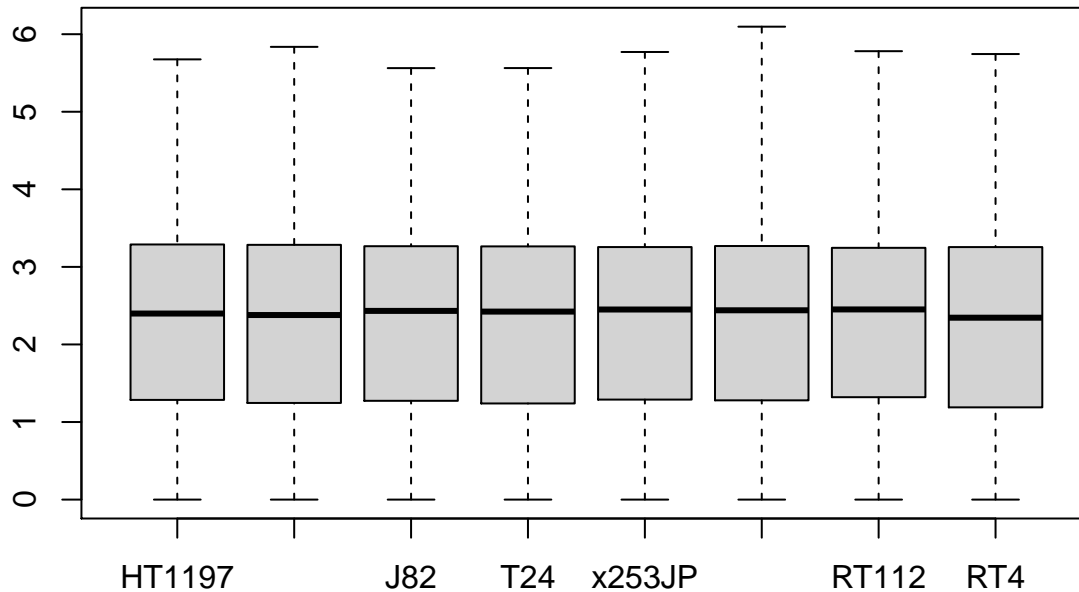
```
colData(dds)
```

```
## DataFrame with 8 rows and 3 columns
##           condition replicate sizeFactor
##           <factor> <integer> <numeric>
## HT1197          R           1  0.984512
## HT1376          R           2  1.023008
## J82             R           3  0.901071
## T24            R           4  1.162033
## x253JP          S           1  1.191820
## x5637          S           2  0.942569
## RT112          S           3  0.954567
## RT4            S           4  1.037255
```

```
keep <- rowSums(counts(dds) >= 5) >= 4
table(keep)
```

```
## keep
## FALSE TRUE
## 36026 19414
```

```
dds <- dds[keep,]
normalized_counts <- counts(dds, normalized=TRUE)
boxplot(log10(counts(dds, normalized=TRUE)+1))
```



```
vsd <- vst(dds)
```

```
## -- note: fitType='parametric', but the dispersion trend was not well captured by the
## function: y = a/x + b, and a local regression fit was automatically substituted.
## specify fitType='local' or 'mean' to avoid this message next time.
```

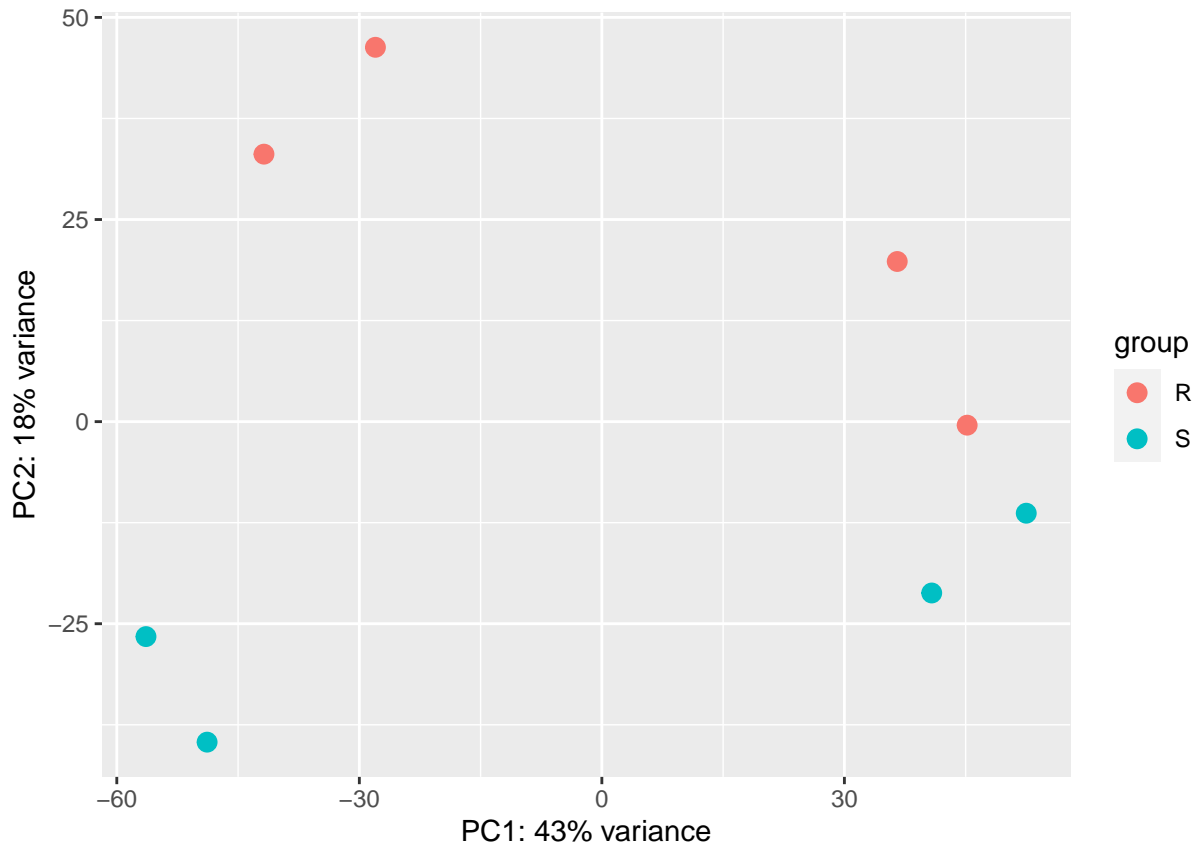
```
class(vsd)
```

```
## [1] "DESeqTransform"
## attr(,"package")
## [1] "DESeq2"
```

```
assay(vsd)[1:3, 1:8]
```

```
##           HT1197  HT1376      J82      T24  x253JP  x5637
## ENSG00000000003 10.651548  9.848954 10.725604 10.043988 11.094746 11.902333
## ENSG000000000457  9.321348  9.581646  9.096379  9.299386  9.101097  9.841628
## ENSG000000000460  9.962911 10.887914 10.198737 10.215316  9.869072 10.694782
##           RT112      RT4
## ENSG00000000003 12.975876 10.407563
## ENSG000000000457  9.876405  9.254231
## ENSG000000000460  9.720845  9.822061
```

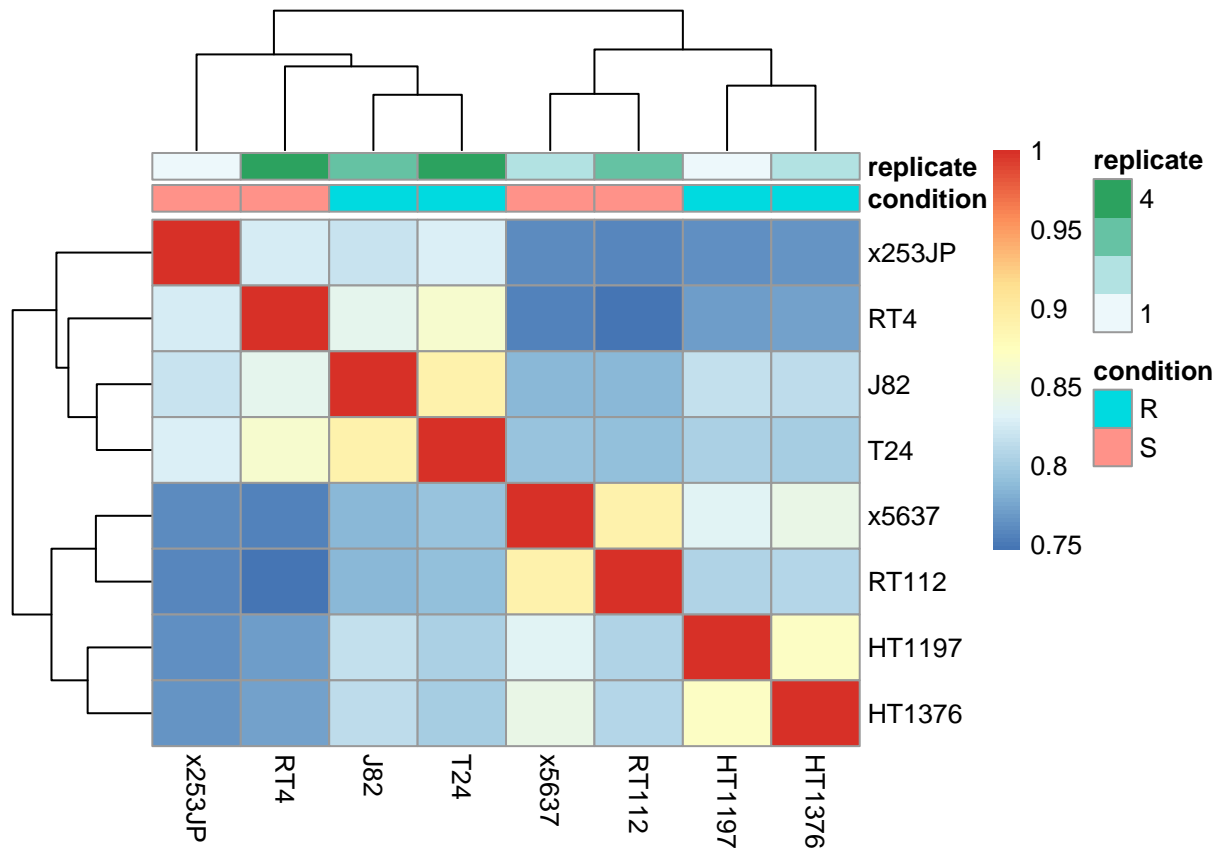
```
plotPCA(vsd, "condition")
```



```
plotPCA(vsd, "condition", returnData=TRUE)
```

```
##           PC1           PC2 group condition  name
## HT1197 -28.00838  46.3037010    R      R   HT1197
## HT1376 -41.79575  33.0875849    R      R   HT1376
## J82     36.53862  19.8098940    R      R     J82
## T24     45.18675  -0.4378793    R      R     T24
## x253JP  40.80287 -21.1969259    S      S   x253JP
## x5637  -56.39084 -26.5916600    S      S   x5637
## RT112  -48.82928 -39.6456937    S      S   RT112
## RT4    52.49602 -11.3290211    S      S     RT4
```

```
vsd %>%
  assay() %>%
  cor() %>%
  pheatmap(annotation=samples[,c("condition", "replicate")])
```



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

```
## using pre-existing size factors
```

```
## estimating dispersions
```

```
## gene-wise dispersion estimates
```

```
## mean-dispersion relationship
```

```
## final dispersion estimates
```

```
## fitting model and testing
```

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

```
## [1] "Intercept"          "condition_S_vs_R"
```

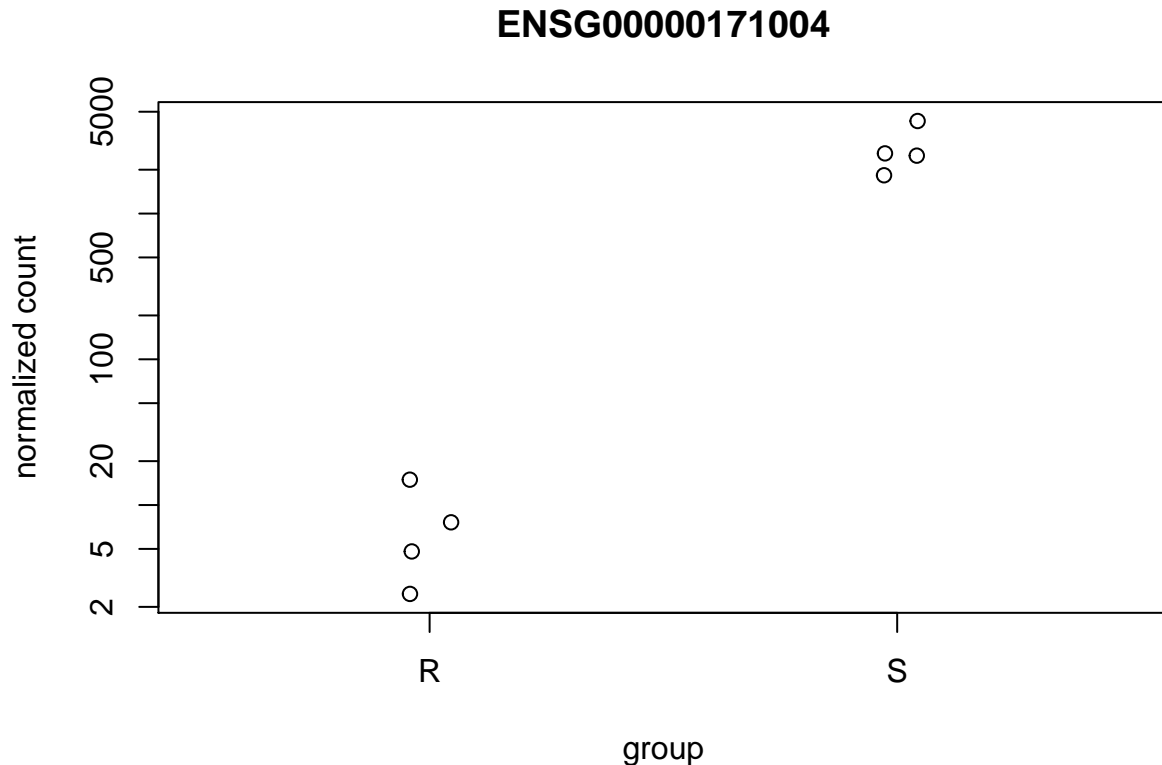
```
head(res[order(res$padj), ], n=60)
```

```
## log2 fold change (MLE): condition S vs R
```

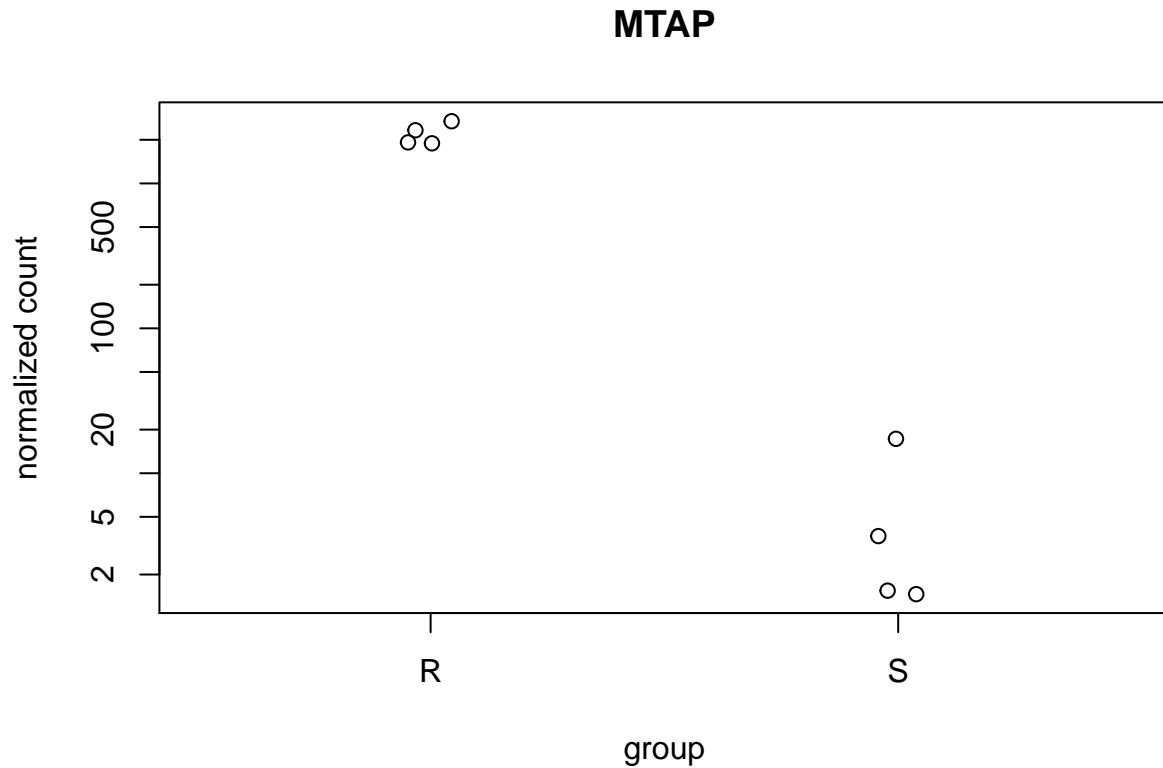
```
## Wald test p-value: condition S vs R
```

```
## DataFrame with 60 rows and 6 columns
##           baseMean log2FoldChange   lfcSE   stat   pvalue
##           <numeric> <numeric> <numeric> <numeric> <numeric>
## ENSG00000171004  1407.34    8.68073  0.619538  14.01163  1.32336e-44
## ENSG00000149582   537.20   -5.64295  0.505749 -11.15762  6.57283e-29
## ENSG00000099810  1105.68   -8.61934  0.856420 -10.06438  7.93861e-24
## ENSG00000165949  1949.55   -6.90973  0.688394 -10.03747  1.04321e-23
## ENSG00000166147 16350.49   -8.11368  0.840921  -9.64856  4.98529e-22
## ...
## ENSG00000074410  5928.905    7.30609  1.358011   5.37999  7.44892e-08
## ENSG00000064787  2653.990    9.41571  1.753535   5.36956  7.89286e-08
## ENSG00000076706   865.793   -7.70499  1.436826  -5.36251  8.20741e-08
## ENSG00000143369  2258.291   -5.18343  0.968615  -5.35138  8.72857e-08
## ENSG00000130758  1061.319    1.46541  0.274099   5.34630  8.97725e-08
##           padj
##           <numeric>
## ENSG00000171004  2.26599e-40
## ENSG00000149582  5.62733e-25
## ENSG00000099810  4.46570e-20
## ENSG00000165949  4.46570e-20
## ENSG00000166147  1.70726e-18
## ...
## ENSG00000074410  2.27764e-05
## ENSG00000064787  2.37104e-05
## ENSG00000076706  2.42303e-05
## ENSG00000143369  2.53321e-05
## ENSG00000130758  2.54505e-05
```

```
plotCounts(dds, which.min(res$padj), "condition")
```



```
plotCounts(dds, gene = "ENSG00000099810", "condition", main="MTAP")
```



```
summary(res)
```

```
##  
## out of 19414 with nonzero total read count  
## adjusted p-value < 0.1  
## LFC > 0 (up)      : 277, 1.4%  
## LFC < 0 (down)    : 350, 1.8%  
## outliers [1]      : 785, 4%  
## low counts [2]    : 1506, 7.8%  
## (mean count < 9)  
## [1] see 'cooksCutoff' argument of ?results  
## [2] see 'independentFiltering' argument of ?results
```

```
sum(res$padj <0.1, na.rm = TRUE)
```

```
## [1] 627
```

```
sum(res$padj <0.05, na.rm = TRUE)
```

```
## [1] 446
```

```
sum(res$padj <0.01, na.rm = TRUE)
```

```
## [1] 233
```

```
res1 <- results(dds, name = "condition_S_vs_R")
res1 <- results(dds, contrast = c("condition", "S", "R"))
resLFC <- lfcShrink(dds, coef = "condition_S_vs_R", type="apeglm")
```

```
## using 'apeglm' for LFC shrinkage. If used in published research, please cite:
##   Zhu, A., Ibrahim, J.G., Love, M.I. (2018) Heavy-tailed prior distributions for
##   sequence count data: removing the noise and preserving large differences.
##   Bioinformatics. https://doi.org/10.1093/bioinformatics/bty895
```

```
resLFC
```

```
## log2 fold change (MAP): condition S vs R
## Wald test p-value: condition S vs R
## DataFrame with 19414 rows and 5 columns
##
```

	baseMean	log2FoldChange	lfcSE	pvalue	padj
##	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
## ENSG00000000003	2488.448	0.8658578	1.236514	9.13667e-03	0.164410405
## ENSG000000000457	486.381	0.1202203	0.247501	3.79943e-01	0.811809642
## ENSG000000000460	1030.492	-0.0991357	0.251677	4.32892e-01	0.843179196
## ENSG000000000971	2455.485	6.9794622	1.514278	1.14249e-07	0.000031553
## ENSG00000001036	3683.040	-0.2075753	0.308397	1.66731e-01	0.636387722
##
## ENSG00000283041	6073.4798	0.0149682	0.227995	0.9064768	0.984796
## ENSG00000283050	407.4790	-0.0724409	0.248527	0.5447904	0.887176
## ENSG00000283064	11.6368	-0.1445268	0.317970	0.0796848	0.492223
## ENSG00000283078	36.2228	-0.0643776	0.259935	0.5313379	0.882368
## ENSG00000283103	150.3640	0.2732784	0.424769	0.0514922	0.402398

```
resLFCa <- resLFC[order(resLFC$padj),]
resLFCb <- resLFCa[1:60,]
resLFCb
```

```
## log2 fold change (MAP): condition S vs R
## Wald test p-value: condition S vs R
## DataFrame with 60 rows and 5 columns
##
```

	baseMean	log2FoldChange	lfcSE	pvalue	padj
##	<numeric>	<numeric>	<numeric>	<numeric>	<numeric>
## ENSG00000171004	1407.34	8.59263	0.624035	1.32336e-44	2.26599e-40
## ENSG00000149582	537.20	-5.54878	0.509533	6.57283e-29	5.62733e-25
## ENSG00000099810	1105.68	-8.44030	0.860692	7.93861e-24	4.46570e-20
## ENSG00000165949	1949.55	-6.76511	0.695394	1.04321e-23	4.46570e-20
## ENSG00000166147	16350.49	-7.92542	0.850526	4.98529e-22	1.70726e-18
##
## ENSG00000074410	5928.905	6.773197	1.436804	7.44892e-08	2.27764e-05
## ENSG00000064787	2653.990	8.711776	1.878490	7.89286e-08	2.37104e-05
## ENSG00000076706	865.793	-0.117002	0.313760	8.20741e-08	2.42303e-05
## ENSG00000143369	2258.291	-4.780574	1.018371	8.72857e-08	2.53321e-05
## ENSG00000130758	1061.319	1.364209	0.283355	8.97725e-08	2.54505e-05

Top 60 Gene

```
meta <- samples
meta <- meta %>%
  rownames_to_column(var = "samplename") %>%
  as_tibble()

normalized_counts1 <- normalized_counts
normalized_counts1 <- normalized_counts1 %>%
  data.frame() %>%
  rownames_to_column(var = "gene")

resA <- res
resA <- data.frame(resA)
res_table_tb <- resA
res_table_tb <- data.frame(res_table_tb) %>%
  rownames_to_column(var = "gene") %>%
  as_tibble()
top30_sig_genes <- res_table_tb %>%
  arrange(padj) %>%
  pull(gene) %>%
  head(n=30)

top30_sig_genes
```

```
## [1] "ENSG00000171004" "ENSG00000149582" "ENSG00000099810" "ENSG00000165949"
## [5] "ENSG00000166147" "ENSG00000114270" "ENSG00000147889" "ENSG00000141574"
## [9] "ENSG00000147883" "ENSG00000128591" "ENSG00000139910" "ENSG00000142619"
## [13] "ENSG00000115414" "ENSG00000151640" "ENSG00000117525" "ENSG00000131737"
## [17] "ENSG00000239697" "ENSG00000255874" "ENSG00000162734" "ENSG00000184489"
## [21] "ENSG00000265194" "ENSG00000103064" "ENSG00000079931" "ENSG00000185885"
## [25] "ENSG00000168386" "ENSG00000101335" "ENSG00000221968" "ENSG00000151388"
## [29] "ENSG00000177096" "ENSG00000170848"
```

Session information

```
sessionInfo()
```

```
## R version 4.0.3 (2020-10-10)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS Big Sur 10.16
##
## Matrix products: default
## BLAS: /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRblas.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
```

```

## [1] parallel stats4 stats graphics grDevices utils datasets
## [8] methods base
##
## other attached packages:
## [1] vsn_3.58.0 tximportData_1.18.0
## [3] tximport_1.18.0 tximeta_1.8.3
## [5] tidyverse_1.3.0 tidyr_1.1.2
## [7] tibble_3.0.4 testthat_3.0.1
## [9] stringr_1.4.0 Rsubread_2.4.2
## [11] rmarkdown_2.6 readr_1.4.0
## [13] Rcpp_1.0.5 RColorBrewer_1.1-2
## [15] purrr_0.3.4 pheatmap_1.0.12
## [17] pbapply_1.4-3 magrittr_2.0.1
## [19] locfit_1.5-9.4 knitr_1.30
## [21] IHW_1.18.0 httr_1.4.2
## [23] gplots_3.1.1 goseq_1.42.0
## [25] geneLenDataBase_1.26.0 geneplotter_1.68.0
## [27] annotate_1.68.0 XML_3.99-0.5
## [29] lattice_0.20-41 genefilter_1.72.0
## [31] forcats_0.5.0 EnsDb.Hsapiens.v86_2.99.0
## [33] ensemblDb_2.14.0 GenomicFeatures_1.42.1
## [35] EnhancedVolcano_1.8.0 ggrepel_0.9.0
## [37] ggplot2_3.3.2 dplyr_1.0.2
## [39] DOSE_3.16.0 devtools_2.3.2
## [41] usethis_2.0.0 DESeq2_1.30.0
## [43] DEGreport_1.26.0 data.table_1.13.4
## [45] clusterProfiler_3.18.0 biomaRt_2.46.0
## [47] BiasedUrn_1.07 ashR_2.2-47
## [49] apeglm_1.12.0 AnnotationHub_2.22.0
## [51] BiocFileCache_1.14.0 dbplyr_2.0.0
## [53] AnnotationFilter_1.14.0 AnnotationDbi_1.52.0
## [55] annotables_0.1.91 airway_1.10.0
## [57] SummarizedExperiment_1.20.0 Biobase_2.50.0
## [59] GenomicRanges_1.42.0 GenomeInfoDb_1.26.2
## [61] IRanges_2.24.1 S4Vectors_0.28.1
## [63] BiocGenerics_0.36.0 MatrixGenerics_1.2.0
## [65] matrixStats_0.57.0
##
## loaded via a namespace (and not attached):
## [1] rappdirs_0.3.1 rtracklayer_1.50.0
## [3] coda_0.19-4 bit64_4.0.5
## [5] irlba_2.3.3 DelayedArray_0.16.0
## [7] RCurl_1.98-1.2 generics_0.1.0
## [9] preprocessCore_1.52.0 callr_3.5.1
## [11] cowplot_1.1.0 RSQLite_2.2.1
## [13] shadowtext_0.0.7 bit_4.0.4
## [15] enrichplot_1.10.1 lubridate_1.7.9.2
## [17] xml2_1.3.2 httpuv_1.5.4
## [19] assertthat_0.2.1 viridis_0.5.1
## [21] xfun_0.19 hms_0.5.3
## [23] evaluate_0.14 promises_1.1.1
## [25] fansi_0.4.1 progress_1.2.2
## [27] readxl_1.3.1 caTools_1.18.0
## [29] igraph_1.2.6 DBI_1.1.0

```

```

## [31] tmvnsim_1.0-2          reshape_0.8.8
## [33] ellipsis_0.3.1        backports_1.2.1
## [35] vctrs_0.3.6           remotes_2.2.0
## [37] Cairo_1.5-12.2        withr_2.3.0
## [39] ggforce_0.3.2         lasso2_1.2-21.1
## [41] bdsmatrix_1.3-4       GenomicAlignments_1.26.0
## [43] fdrtool_1.2.15       prettyunits_1.1.1
## [45] mnormt_2.0.2          cluster_2.1.0
## [47] lazyeval_0.2.2        crayon_1.3.4
## [49] labeling_0.4.2        slam_0.1-48
## [51] edgeR_3.32.0          pkgconfig_2.0.3
## [53] tweenr_1.0.1          nlme_3.1-151
## [55] vipor_0.4.5           pkgload_1.1.0
## [57] ProtGenerics_1.22.0   rlang_0.4.9
## [59] lifecycle_0.2.0       downloader_0.4
## [61] affyio_1.60.0         extrafontdb_1.0
## [63] modelr_0.1.8          invgamma_1.1
## [65] cellranger_1.1.0      ggrastr_0.2.1
## [67] rprojroot_2.0.2       polyclip_1.10-0
## [69] Matrix_1.3-0          lpsymphony_1.18.0
## [71] reprex_0.3.0          beeswarm_0.2.3
## [73] GlobalOptions_0.1.2   processx_3.4.5
## [75] png_0.1-7             viridisLite_0.3.0
## [77] rjson_0.2.20          bitops_1.0-6
## [79] ConsensusClusterPlus_1.54.0 KernSmooth_2.23-18
## [81] Biostrings_2.58.0     blob_1.2.1
## [83] shape_1.4.5           mixsqp_0.3-43
## [85] SQUAREM_2020.5        qvalue_2.22.0
## [87] scales_1.1.1          memoise_1.1.0
## [89] plyr_1.8.6            zlibbioc_1.36.0
## [91] compiler_4.0.3        scatterpie_0.1.5
## [93] bbmle_1.0.23.1        ash_1.0-15
## [95] clue_0.3-58           affy_1.68.0
## [97] Rsamtools_2.6.0       cli_2.2.0
## [99] XVector_0.30.0        ps_1.5.0
## [101] mgcv_1.8-33           MASS_7.3-53
## [103] tidyselect_1.1.0      stringi_1.5.3
## [105] proj4_1.0-10          emdbook_1.3.12
## [107] yaml_2.2.1            GOSemSim_2.16.1
## [109] askpass_1.1           grid_4.0.3
## [111] fastmatch_1.1-0       tools_4.0.3
## [113] circlize_0.4.11      logging_0.10-108
## [115] gridExtra_2.3         farver_2.0.3
## [117] ggraph_2.0.4          digest_0.6.27
## [119] rvcheck_0.1.8        BiocManager_1.30.10
## [121] shiny_1.5.0           broom_0.7.3
## [123] ggalt_0.4.0           BiocVersion_3.12.0
## [125] later_1.1.0.1         Nozzle.R1_1.1-1
## [127] ggdendro_0.1.22      ComplexHeatmap_2.6.2
## [129] psych_2.0.12          colorspace_2.0-0
## [131] rvest_0.3.6           fs_1.5.0
## [133] truncnorm_1.0-8       splines_4.0.3
## [135] graphlayouts_0.7.1    sessioninfo_1.1.1
## [137] xtable_1.8-4          jsonlite_1.7.2

```

```
## [139] tidygraph_1.2.0          R6_2.5.0
## [141] pillar_1.4.7                htmltools_0.5.0
## [143] mime_0.9                     glue_1.4.2
## [145] fastmap_1.0.1               BiocParallel_1.24.1
## [147] interactiveDisplayBase_1.28.0 maps_3.3.0
## [149] fgsea_1.16.0                pkgbuild_1.2.0
## [151] mvtnorm_1.1-1               numDeriv_2016.8-1.1
## [153] curl_4.3                     ggbeeswarm_0.6.0
## [155] gtools_3.8.2                GO.db_3.12.1
## [157] openssl_1.4.3               Rttf2pt1_1.3.8
## [159] survival_3.2-7              limma_3.46.0
## [161] desc_1.2.0                   munsell_0.5.0
## [163] DO.db_2.9                    GetoptLong_1.0.5
## [165] GenomeInfoDbData_1.2.4      haven_2.3.1
## [167] reshape2_1.4.4              gtable_0.3.0
## [169] extrafont_0.17
```