### CSE185-LAB4-README

July 17, 2023

## 1 CSE185 Lab 4 Report - Code Documentation (10 pts)

- Document any commands used or additional analysis steps below!
- You should include enough detail that the instructors (or your future self) could come back to this several months from now and know exactly what you did and why you did it.
- We will not run this notebook, but will look back to see what you did especially if you end up with different answers.

For grading purposes only - Do not copy or edit this cell!

#### Question 1: Command Documentation

I ran the following commands which open the .genes.results files and then check for the TPM field (fifth field) which are greater than 0 then output the number of lines that follow that criteria. I made sure I was in the  $\sim$ /public/lab4 directory and ran the following commands:

```
[1]: cat Chow_Rep1.genes.results | awk '$5>0' | wc -1 cat Chow_Rep2.genes.results | awk '$5>0' | wc -1 cat Chow_Rep3.genes.results | awk '$5>0' | wc -1 cat HFD_Rep1.genes.results | awk '$5>0' | wc -1 cat HFD_Rep2.genes.results | awk '$5>0' | wc -1 cat HFD_Rep3.genes.results | awk '$5>0' | wc -1 cat HFD_Rep3.genes.results | awk '$5>0' | wc -1
```

#### Question 2: R Code

Code for Scatter Plots comparing log10 TPM values for Chow Rep 1 vs Chow Rep 2 & HFD Rep 1 vs HFD Rep 2

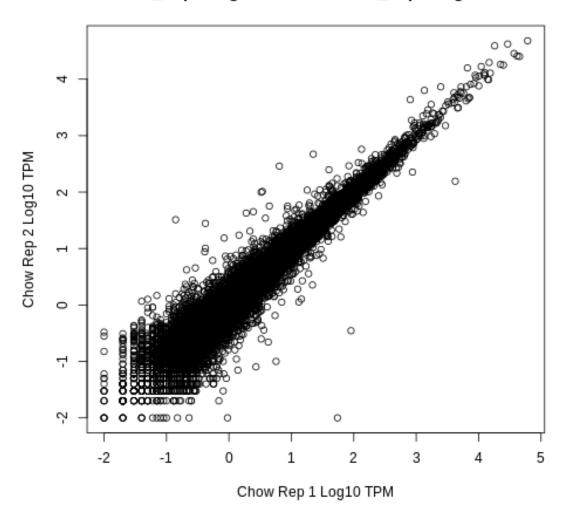
```
[2]: # Run this to allow using the %%R cell magic
%load_ext rpy2.ipython

# Suppress warnings (most notable from rpy2)
import warnings
```

```
warnings.filterwarnings('ignore')
```

```
[3]: \%\R
     # Read in Chow Rep 1 and 2 delim files
     data1 <- read.delim("Chow_Rep1.genes.results", header = TRUE, sep = "\t")</pre>
     data2 <- read.delim("Chow_Rep2.genes.results", header = TRUE, sep = "\t")</pre>
     # Convert into 1og10 TPM values
     TPM1 <- log10(data1$TPM)</pre>
     TPM2 <- log10(data2$TPM)</pre>
     # Scatter Plot with Chow Rep 1 as x-axis and Chow Rep 2 as y-axis
     plot(TPM1, TPM2, main = "Chow_Rep1 Log10 TPM vs. Chow_Rep2 Log10 TPM", xlab = Log10 TPM vs. Chow_Rep2 Log10 TPM", xlab
      →"Chow Rep 1 Log10 TPM", ylab = "Chow Rep 2 Log10 TPM")
     # Read in HFD Rep 1 and 2 delim files
     hfd1 <- read.delim("HFD_Rep1.genes.results", header = TRUE, sep = "\t")</pre>
     hfd2 <- read.delim("HFD_Rep2.genes.results", header = TRUE, sep = "\t")</pre>
     #Convert into log10 TPM values
     tpm1 <- log10(hfd1$TPM)</pre>
     tpm2 <- log10(hfd2$TPM)</pre>
     # Scatter plot with HFD Rep1 as x-axis and HFD Rep2 as y-axis
     plot(tpm1, tpm2, main = "HFD_Rep1 Log10 TPM vs. HFD_Rep2 Log10 TPM", xlab =
      →"HFD Rep 1 Log10 TPM", ylab = "HFD Rep 2 Log10 TPM")
```

## Chow\_Rep1 Log10 TPM vs. Chow\_Rep2 Log10 TPM



Part 2: Differential Expression Analysis

```
##### Load the libraries we need #####
library("DESeq2")
library("tximport")

# This cell is an example. We recommend putting the R code you use for this
# in the CSE185-LAB4-README.ipynb notebook

# You might find some of the code below helpful!
# Or, you can ignore what we have below and follow
```

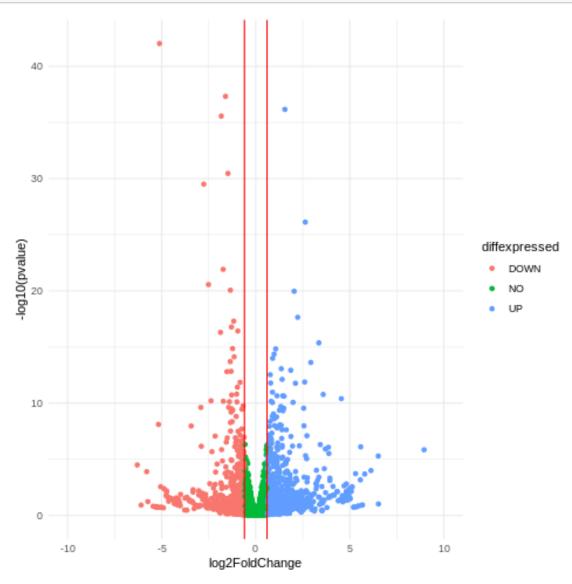
```
# http://bioconductor.org/packages/devel/bioc/vignettes/DESeq2/inst/doc/DESeq2.
 ∽html
##### List the files and set up metadata #####
# Note, you should change this to use the files in your home directory
#files <- c("Chow Rep1.genes.results",</pre>
          "Chow Rep2.genes.results",
#
          "Chow_Rep3.genes.results",
#
          "HFD_Rep1.genes.results",
#
          "HFD_Rep2.genes.results",
          "HFD_Rep3.genes.results")
#conditions <- c(rep("Chow", 3), rep("HFD", 3))</pre>
#samples <- data.frame("run"=c("Chow_Rep1", "Chow_Rep2", "Chow_Rep3", "</pre>

¬"HFD_Rep1", "HFD_Rep2", "HFD_Rep3"),
                      "condition"=conditions)
#names(files) = samples$run
sampleFiles <- c("Chow_Rep1.genes.results", "Chow_Rep2.genes.results", "</pre>
 ⇔"Chow_Rep3.genes.results",
                  "HFD_Rep1.genes.results", "HFD_Rep2.genes.results", "HFD_Rep3.
⇔genes.results")
sampleNames <- c("Chow_Rep1", "Chow_Rep2", "Chow_Rep3", "HFD_Rep1", "HFD_Rep2", "
→"HFD Rep3")
samplePaths <- file.path(getwd(), sampleFiles)</pre>
# Define a function to read in RSEM files and extract the transcript IDs and
read rsem file <- function(filepath) {</pre>
 rsem_data <- read.table(filepath, header=TRUE, stringsAsFactors=FALSE)</pre>
 transcript_ids <- rsem_data$transcript_id</pre>
 counts <- rsem_data$expected_count</pre>
 return(list(transcript_ids=transcript_ids, counts=counts))
}
# Read in the RSEM data for all samples and combine into a single matrix
sample_data <- lapply(samplePaths, read_rsem_file)</pre>
transcript_ids <- sample_data[[1]]$transcript_ids</pre>
counts_matrix <- sapply(sample_data, function(x) round(x$counts))</pre>
colnames(counts_matrix) <- sampleNames</pre>
rownames(counts_matrix) <- transcript_ids</pre>
# Create a data frame with the sample names and condition labels
sample info <- data.frame(sampleName=sampleNames,</pre>

¬condition=rep(c("Chow","HFD"), each=3))
# Create the DESeq2 data object
```

```
# Filter out genes with low counts
      dds <- dds[ rowSums(counts(dds)) > 10, ]
      # Run DESeq2 analysis
      dds <- DESeq(dds)
      # Extract the differential expression results
      results <- results(dds)
      # Write the results to a file
      write.csv(as.data.frame(results), file = "chow_vs_hfd_deseq2.csv")
     R[write to console]: converting counts to integer mode
     R[write to console]: estimating size factors
     R[write to console]: estimating dispersions
     R[write to console]: gene-wise dispersion estimates
     R[write to console]: mean-dispersion relationship
     R[write to console]: final dispersion estimates
     R[write to console]: fitting model and testing
     Question 4: R Code for Volcano Plot
[25]: \%\R
      library(ggplot2)
      # Read in CSV file
      data <- read.csv("chow_vs_hfd_deseq2.csv")</pre>
      # add column of NAs
      data$diffexpressed <- "NO"
      # if log2FoldChange > 0.6, set as "UP"
      data$diffexpressed[data$log2FoldChange > 0.6] <- "UP"</pre>
      #if log2FoldChange < -0.6, set as "DOWN"
      data$diffexpressed[data$log2FoldChange < -0.6] <- "DOWN"
      ggplot(data, aes(x = log2FoldChange, y = -log10(pvalue), col = diffexpressed)) +
```

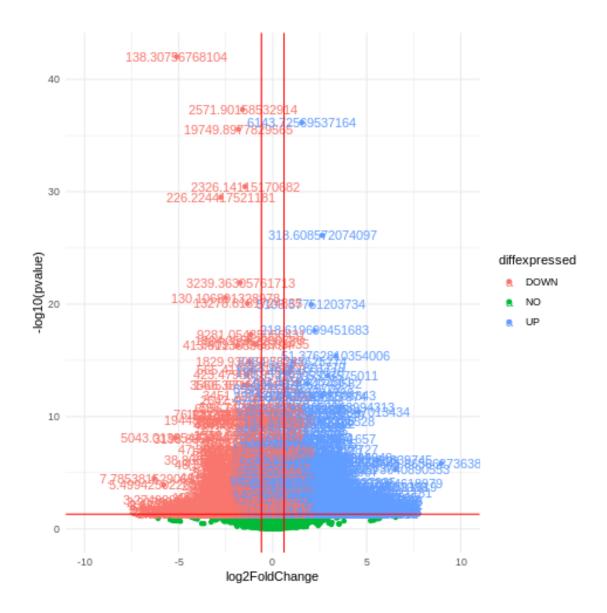
```
geom_point() +
theme_minimal() +
geom_vline(xintercept = c(-0.6, 0.6), col='red') +
xlim(-10, 10)
```



#### Question 5:

I used the same code as before but this time I added the p-value threshold of 0.05 (5%) and also added labels for the baseMean for the most differentially expressed genes. Then, I went through the plot starting at the top gene (most differentially expressed) and went down through 10 genes. Which can be used to go through and see which gene has that baseMean. Then, I can use this to search for that gene using grep and find out its log2fold change and p-value and also grep again to find the gene name in GRCm38.75.gene\_names.

# [26]: %%R library(ggplot2) # Read in CSV file data <- read.csv("chow\_vs\_hfd\_deseq2.csv")</pre> # add column of NAs data\$diffexpressed <- "NO"</pre> # if log2FoldChange > 0.6 and p-value < 0.05, set as "UP"</pre> data\$diffexpressed[data\$log2FoldChange > 0.6 & data\$pvalue < 0.05] <- "UP" #if log2FoldChange < -0.6 and pvalue < 0.05, set as "DOWN" data\$diffexpressed[data\$log2FoldChange < -0.6 & data\$pvalue < 0.05] <- "DOWN" data\$dlabel <- NA data\$dlabel[data\$diffexpressed != "NO"] <- data\$baseMean[data\$diffexpressed !=\_\_ →"NO"] ggplot(data, aes(x = log2FoldChange, y = -log10(pvalue), col = diffexpressed, u $\hookrightarrow$ label = dlabel)) + geom\_point() + theme\_minimal() + geom\_text() + geom\_vline(xintercept = c(-0.6, 0.6), col='red') + geom\_hline(yintercept = -log10(0.05), col = 'red') + xlim(-10, 10)



Question 6: GO Analysis I went to the GO Analysis Tool DAVID which can be found here: https://david.ncifcrf.gov/tools.jsp Then, I used the following tutorial to perform GO analysis via DAVID and just followed the directions by each slide: https://david.ncifcrf.gov/helps/tutorial.pdf Question 7: For Question 7, I just made a table of the returned enriched categories and added their

corresponding p-value