## Glycolysis and Gluconeogenesis Pathway Heatmap

## Introduction

This report will explain the steps used to make a heatmap produce of average log2(CPM) expression of selected glycolysis and gluconeogenesis genes across all non-diabetic single cell samples. All non-diabetic samples were included in the heat map excluding "none" and "multiple" classified samples.

```
# Load in libraries
suppressPackageStartupMessages(library(Biobase))
suppressPackageStartupMessages(library(pheatmap))
suppressPackageStartupMessages(library(RColorBrewer))
suppressPackageStartupMessages(library(edgeR))
suppressPackageStartupMessages(library(gplots))
suppressPackageStartupMessages(library(ggplot2))
library(Biobase)
library(pheatmap)
library(gplots)
library(ggplot2)
library(edgeR)
library(RColorBrewer)
rm(list=ls())
# Load in data
setwd("/Users/lawlon/Documents/Final_RNA_Seq_3/Data/")
load("nonT2D.rdata")
s.anns <- pData(cnts.eset)</pre>
p.anns <- as(featureData(cnts.eset), "data.frame")</pre>
counts <- exprs(cnts.eset)</pre>
# Calculate the cpm of the data
cpms \leftarrow cpm(x = counts)
data <- log2(cpms+1)
# Get cell types in order
s.1 <- s.anns[s.anns$cell.type %in% c("INS"),]</pre>
s.2 <- s.anns[s.anns$cell.type %in% c("GCG"),]
s.3 <- s.anns[s.anns$cell.type %in% c("SST"),]</pre>
s.4 <- s.anns[s.anns$cell.type %in% c("PPY"),]</pre>
s.5 <- s.anns[s.anns$cell.type %in% c("PRSS1"),]</pre>
s.6 <- s.anns[s.anns$cell.type %in% c("KRT19"),]</pre>
s.7 <- s.anns[s.anns$cell.type %in% c("COL1A1"),]
# Get Expression matrices and average mean expression
f.1 <- data[, rownames(s.1)]</pre>
avg1 <- rowMeans(f.1)</pre>
f.2 <- data[, rownames(s.2)]</pre>
avg2 <- rowMeans(f.2)</pre>
f.3 <- data[, rownames(s.3)]</pre>
avg3 <- rowMeans(f.3)
f.4 <- data[, rownames(s.4)]</pre>
avg4 <- rowMeans(f.4)
f.5 <- data[, rownames(s.5)]</pre>
avg5 <- rowMeans(f.5)</pre>
```

```
f.6 <- data[, rownames(s.6)]</pre>
avg6 <- rowMeans(f.6)</pre>
f.7 <- data[, rownames(s.7)]</pre>
avg7 <- rowMeans(f.7)</pre>
# Match up cell type with hormone marker
namelist <- c(INS="Beta", GCG="Alpha", SST="Delta", PPY="Gamma",
               GHRL="Epsilon", COL1A1="Stellate", PRSS1="Acinar",
               KRT19="Ductal", none="None")
# Combine all cell expression data into one matrix
mat.orig \leftarrow cbind(f.1, f.2, f.3, f.4, f.5, f.6, f.7)
mat.avg <- cbind(avg1, avg2, avg3, avg4, avg5, avg6, avg7)</pre>
colnames(mat.avg) <- c("Beta", "Alpha", "Delta", "Gamma",</pre>
                         "Acinar", "Ductal", "Stellate")
# Transpose avq matrix
trans.mat <- t(mat.avg)</pre>
# Change colnames of mat to gene symbol
colnames(trans.mat) <- p.anns$Associated.Gene.Name</pre>
# Glycolysis pathway genes
genes.sel <- c("SLC2A1", "SLC2A2", "SLC2A3", "GCK", "HK1", "HK2",</pre>
                "GPI", "PFKL", "PFKM", "PFKP", "ALDOA", "ALDOB", "ALDOC",
                "GAPDH", "PGK1", "PGAM1", "PGAM5", "ENO1", "ENO2", "ENO3",
                "ENO4", "PKM", "LDHA", "LDHB", "LDHC",
                "G6PC2", "PFKFB2", "GPD1", "GPD2")
# Find ids of genes
genes.ids <- NULL
for (i in 1:length(genes.sel)) {
  idx <- which(colnames(trans.mat) == genes.sel[i])</pre>
  genes.ids <- c(genes.ids, idx)</pre>
# Extract genes of interest
mat.sel <- trans.mat[, genes.ids]</pre>
# transpose again
mat.sel <- t(mat.sel)</pre>
# Mean center the data
# Mean center by column (gene)
center_apply <- function(x) {</pre>
  apply(x, 1, function(y) y - mean(y))
}
mat.center <- center_apply(mat.sel)</pre>
# Scale the data between -1 and 1
nor.min.max <- function(x) {</pre>
  if (is.numeric(x) == FALSE) {
    stop("Please input numeric for x")
  x.min \leftarrow min(x)
  x.max \leftarrow max(x)
```

```
x \leftarrow 2*((x - x.min) / (x.max - x.min)) - 1
 return (x)
# Only use if want to scale data
mat.scale <- t(apply(mat.center, 2, nor.min.max))</pre>
# Annotation matrix
annotation_col = data.frame(Cell_Type = colnames(mat.scale))
rownames(annotation_col) <- colnames(mat.scale)</pre>
# Specify cell type colors
grey <- brewer.pal(n=9, name="Greys")</pre>
ann_colors <- list(</pre>
 Cell_Type = c(Beta="#e41a1c", Alpha = "#377eb8",
                Delta = "#4daf4a", Gamma = "#984ea3",
                Acinar = "#525252", Ductal = "#969696", Stellate = "#000000"))
# Make heatmap
pheatmap(mat = mat.sel, cluster_rows = FALSE, cluster_cols = FALSE,
         color = colorRampPalette(brewer.pal(n = 7, name = "Greys"))(20),
         annotation_colors = ann_colors, clustering_distance_rows = NULL,
         clustering method=NULL, show rownames=TRUE,
         show_colnames = TRUE, annotation_names_row = TRUE,
         annotation_names_col = FALSE, trace = "none", fontsize_row = 8,
         cellwidth = 16, cellheight = 16, annotation_legend = FALSE)
```

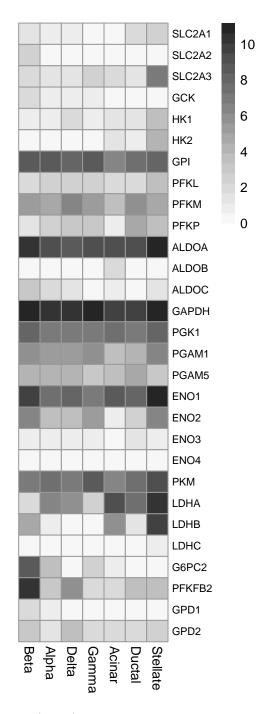


Figure 1: Heat map of average log2(CPM) expression of glycolysis and gluconeogenesis genes across non-diabetic endocrine single cell samples.

## ## Session Information

```
# Load libraries
suppressPackageStartupMessages(library(Biobase))
suppressPackageStartupMessages(library(pheatmap))
suppressPackageStartupMessages(library(RColorBrewer))
suppressPackageStartupMessages(library(edgeR))
suppressPackageStartupMessages(library(gplots))
suppressPackageStartupMessages(library(ggplot2))
library(Biobase)
library(pheatmap)
library(gplots)
library(ggplot2)
library(edgeR)
library(RColorBrewer)
sessionInfo()
## R version 3.3.0 (2016-05-03)
## Platform: x86_64-apple-darwin13.4.0 (64-bit)
## Running under: OS X 10.11.3 (El Capitan)
## 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 stats
                           graphics grDevices utils
                                                         datasets methods
## [8] base
## other attached packages:
## [1] ggplot2_2.1.0
                           gplots_3.0.1
                                               edgeR 3.14.0
## [4] limma_3.28.7
                           RColorBrewer_1.1-2 pheatmap_1.0.8
## [7] Biobase_2.32.0
                           BiocGenerics_0.18.0
##
## loaded via a namespace (and not attached):
## [1] Rcpp_0.12.5
                           knitr_1.13
                                              magrittr_1.5
## [4] munsell_0.4.3
                           colorspace_1.2-6
                                              stringr_1.0.0
## [7] plyr_1.8.4
                           caTools_1.17.1
                                              tools_3.3.0
## [10] grid_3.3.0
                           gtable_0.2.0
                                              KernSmooth_2.23-15
## [13] htmltools_0.3.5
                           gtools_3.5.0
                                              yaml_2.1.13
## [16] digest_0.6.9
                           formatR_1.4
                                              bitops_1.0-6
                           rmarkdown 0.9.6
                                              gdata 2.17.0
## [19] evaluate 0.9
## [22] stringi_1.1.1
                           scales_0.4.0
```