Capstone Project - Roberto Atallah

Integrated analysis of IncRNA-miRNA-mRNA ceRNA network in human renal cell carcinoma

Code

```
deg.R
```

```
# Differential expression analysis with limma
#Set the working directory to where the R scripts are found
setwd("~/Documents/capstone/rscripts/")
#Install BiocManager package to be able to then install the GEOquery and limma packages
if (!require("BiocManager", quietly = TRUE))
 install.packages("BiocManager")
BiocManager::install("GEOquery")
BiocManager::install("limma")
#load them
library(GEOquery)
library(limma)
library(dplyr)
# load series and platform data from GEO, this platform is for total RNA
gset <- getGEO("GSE16441", GSEMatrix =TRUE, AnnotGPL=TRUE)</pre>
if (length(gset) > 1) idx <- grep("GPL6480", attr(gset, "names")) else idx <- 1</pre>
gset <- gset[[idx]]</pre>
pData(gset)$data_processing[1] #Data has already been LOWESS normalized and log2 transformed, it is indicated in
file.
# make proper column names to match toptable
fvarLabels(gset) <- make.names(fvarLabels(gset))</pre>
# group membership for all samples
sml <- strsplit(gsms, split="")[[1]]</pre>
# assign samples to groups and set up design matrix
gs <- factor(sml)</pre>
groups <- make.names(c("RCC","Normal"))</pre>
levels(gs) <- groups</pre>
gset$group <- gs</pre>
design <- model.matrix(~group + 0, gset)</pre>
colnames(design) <- levels(gs)</pre>
#fit linear model
fit <- lmFit(gset, design)</pre>
# set up contrasts of interest and recalculate model coefficients
cts <- paste(groups[1], groups[2], sep="-")</pre>
cont.matrix <- makeContrasts(contrasts=cts, levels=design)</pre>
fit2 <- contrasts.fit(fit, cont.matrix)</pre>
# compute statistics and table of top significant genes
fit2 <- eBayes(fit2, 0.01)</pre>
tT <- topTable(fit2, adjust="fdr", sort.by="B", number=Inf)</pre>
tT <- subset(tT, select=c("ID", "adj.P.Val", "P.Value", "t", "B", "logFC", "Gene.symbol", "Gene.title", "Gene.ID"))
#Take table and mutate it based on cut off condition to extract DEGs, remove the genes with no name (undefined)
deg <- tT %>%
 mutate(condition = abs(logFC) > 2 & adj.P.Val < 0.05 & Gene.symbol != "")</pre>
# summarize test results
sum(deg$condition == TRUE) #differentially regulated
sum(deg$condition== TRUE & deg$logFC > 0) #upregulated
sum(deg$condition== TRUE & deg$logFC < 0) #downregulated</pre>
deg <- subset(deg, deg$condition == TRUE)</pre>
#having a file in the capstone folder, called files to store all necessary tables and txt files
write.csv(deg, file="../files/DEG_table.csv", row.names=FALSE)
write.table(deg$Gene.ID, file= "../files/DEG_IDs.txt", row.names = FALSE, col.names = FALSE)
write.table(deg$Gene.symbol, file= "../files/DEG_symbols.txt", row.names = FALSE, col.names = FALSE)
#REGEX to take long non coding RNA from the data, it should start by LINC
write.table(grep("LINC.*", deg$Gene.symbol, value = TRUE), "../files/LNC.txt", row.names = FALSE, col.names = FAL
SE)
#finding top 20 DEGs significant and top 6 LINC, to display it in the report
head_deg <- subset(head(deg, n=20), select= c("Gene.symbol", "adj.P.Val", "P.Value", "logFC"))
lnc <- deg[grep("LINC.*", deg$Gene.symbol), ]</pre>
lnc_head <- subset(head(lnc, n=6), select= c("Gene.symbol", "adj.P.Val", "P.Value", "logFC"))</pre>
#writing to file
write.table(head_deg, file="../files/DEG_table_head.csv", row.names=FALSE)
write.table(lnc_head, file="../files/LNC_table_head.csv", row.names=FALSE)
# volcano plot (log P-value vs log fold change)
library(ggplot2)
tT %>%
 mutate(Significant = adj.P.Val < 0.05 & abs(logFC) > 2) %>%
 ggplot(aes(x = logFC, y = -log10(adj.P.Val), col=Significant)) + geom_point()
ex <- exprs(gset)</pre>
# box-and-whisker plot
ord <- order(qs)</pre>
palette(c("#1B9E77", "#7570B3", "#E7298A", "#E6AB02", "#D95F02", "#66A61E", "#A6761D", "#B32424", "#B324B3", "#666
666"))
par(mar=c(7,4,2,1))
title <- paste ("GSE16441", "/", annotation(gset), sep ="")</pre>
boxplot(ex[,ord], boxwex=0.6, notch=T, main=title, outline=FALSE, las=2, col=gs[ord])
legend("topleft", groups, fill=palette(), bty="n")
```

load series and platform data from GEO, this platform is for miRNA gset <- getGEO("GSE16441", GSEMatrix =TRUE, AnnotGPL=TRUE)</pre> if (length(gset) > 1) idx <- grep("GPL8659", attr(gset, "names")) else idx <- 1</pre>

dem.R

#load packages

library(limma) library(dplyr)

library(GEOquery)

Differential expression analysis with limma

setwd("~/Documents/capstone/rscripts/")

```
gset <- gset[[idx]]</pre>
 # make proper column names to match toptable
 fvarLabels(gset) <- make.names(fvarLabels(gset))</pre>
 # group membership for all samples
 sml <- strsplit(gsms, split="")[[1]]</pre>
 #Data has already been LOWESS normalized and log2 transformed, it is indicated in file.
 # assign samples to groups and set up design matrix
 gs <- factor(sml)</pre>
 groups <- make.names(c("RCC","Normal"))</pre>
 levels(gs) <- groups</pre>
 gset$group <- gs</pre>
 design <- model.matrix(~group + 0, gset)</pre>
 colnames(design) <- levels(gs)</pre>
 # fit linear model
 fit <- lmFit(gset, design)</pre>
 # set up contrasts of interest and recalculate model coefficients
 cts <- paste(groups[1], groups[2], sep="-")</pre>
 cont.matrix <- makeContrasts(contrasts=cts, levels=design)</pre>
 fit2 <- contrasts.fit(fit, cont.matrix)</pre>
 # compute statistics and table of top significant genes
 fit2 <- eBayes(fit2, 0.01)</pre>
 tT <- topTable(fit2, adjust="fdr", sort.by="B", number=Inf)</pre>
 tT <- subset(tT, select=c("ID", "adj.P.Val", "P.Value", "t", "B", "logFC", "miRNA_ID", "SPOT_ID"))
 dem <- tT %>%
   mutate(condition = abs(logFC) > 1 & adj.P.Val < 0.05)</pre>
 # summarize test results
 sum(dem$condition == TRUE)
 sum(dem$condition== TRUE & dem$logFC > 0)
 sum(dem$condition== TRUE & dem$logFC < 0)</pre>
 dem <- subset(dem, dem$condition == TRUE)</pre>
 write.csv(dem, file="../files/dem_table.csv", row.names=FALSE)
 write.table(dem$miRNA_ID, file= "../files/dem_symbols.txt", row.names = FALSE, col.names = FALSE)
 #finding top 20 DEMs significant
 head_dem <- subset(head(dem, n=20), select= c("miRNA_ID", "adj.P.Val", "P.Value", "logFC"))
 write.table(head_dem, file="../files/dem_table_head.csv", row.names=FALSE)
 # summarize test results
 # volcano plot (log P-value vs log fold change)
 library(ggplot2)
 tT %>%
   mutate(Significant = adj.P.Val < 0.05 & abs(logFC) > 1) %>%
   ggplot(aes(x = logFC, y = -log10(adj.P.Val), col=Significant)) + geom_point()
 ex <- exprs(gset)</pre>
 # box-and-whisker plot
 ord <- order(gs)
 palette(c("#1B9E77", "#7570B3", "#E7298A", "#E6AB02", "#D95F02", "#66A61E", "#A6761D", "#B32424", "#B324B3", "#666
 par(mar=c(7,4,2,1))
 title <- paste ("GSE16441", "/", annotation(gset), sep ="")
 boxplot(ex[,ord], boxwex=0.6, notch=T, main=title, outline=FALSE, las=2, col=gs[ord])
 legend("topleft", groups, fill=palette(), bty="n")
code.R
 library(dplyr)
 setwd("~/Documents/capstone/rscripts/")
 deg <- read.csv("../files/DEG table.csv")</pre>
 dem <- read.csv("../files/dem table.csv")</pre>
```

#then taking predicted miRNA ids without duplicates (because multiple lncRNA can interact with the same miRNA)

#We got 3 miRNAs found in both DEM and predicted. "hsa-miR-532-5p" "hsa-miR-140-5p" and "hsa-miR-28-5p" stored

write.table(unique(lncmiR\$miRNA), "../files/miRNA predicted.txt", row.names = FALSE, col.names = FALSE)

#Reading file of lncRNA miRNA interactions that was taken from DIANA lnc Base

#in this code, the predictions are made by using only mirWalk database

lncmiR <- read.csv("../files/lncRNA miRNA.csv")</pre>

#in the file miRNA intersection

miRNA predicted <- read.table("../files/miRNA predicted.txt") #Finding intersections of miRNA in DEM file and predicted ones miRNA intersection <- intersect(dem\$miRNA ID, miRNA predicted\$V1) write.table(miRNA intersection, "../files/miRNA intersection.txt", row.names = FALSE, col.names=FALSE)

```
#MIRWalk Target of miRNA prediction, after using the mirWalk website to do it
 mirwalk pred <- read.csv(file = "../files/mirwalk mirNA Targets-3.csv")</pre>
 #taking only columns of miRNA and target
 mirwalk pred <- mirwalk_pred[, c(1,3)]</pre>
 #taking the intersections of mRNA predicted with the DEmRNAs found befire
 intersection table <- subset(mirwalk pred, mirwalk pred[,2] %in% deg$Gene.symbol)
 #removing duplicate rows because some miRNA mRNA interactions are written multiple times (multiple binding sequen
 intersection_table <- distinct(intersection_table)</pre>
 #writing to table
 write.csv(intersection table, "../files/intersection table miRNA mRNA.csv", row.names = FALSE)
 length(unique(mirwalk pred$genesymbol)) #the number of predicted mRNA
 length(unique(intersection_table$genesymbol)) #the number of final mRNA
 miRNA values <- c("hsa-miR-140-5p", "hsa-miR-532-5p", "hsa-miR-28-5p")
 #the lncRNA interacting with these 3 miRNA
 lncRNA miRNA filtered <- subset(lncmiR, miRNA %in% miRNA values)</pre>
 write.table(lncRNA_miRNA_filtered, "../files/lncRNA_miRNA_filtered.csv", row.names = FALSE)
 colnames(lncRNA_miRNA_filtered)[1:2] <- c("source", "target")</pre>
 colnames(intersection_table)[1:2] <- c("source", "target")</pre>
 #adding column called regulation to see if up or down regulated for further color in cytoscape
 deg updated <- deg
 deg updated$regulation <- ifelse(deg updated$logFC > 0, "up", "down")
 dem updated <- dem
 dem updated$regulation <- ifelse(dem updated$logFC > 0, "up", "down")
 #for loop to see if gene up or down regulated
 for (i in 1:nrow(intersection table)) {
   gene <- intersection table[i, 2]</pre>
   match row <- which(deg updated$Gene.symbol == gene)</pre>
   if (length(match row) > 0) {
     regulation <- deg updated[match_row[1], "regulation"]</pre>
   } else {
     regulation <- "NA"
   intersection table[i, "regulation"] <- regulation</pre>
 for (i in 1:nrow(lncRNA miRNA filtered)) {
   mirna <- lncRNA miRNA filtered[i, 2]</pre>
   match_row <- which(dem_updated$miRNA_ID == mirna)</pre>
   if (length(match row) > 0) {
     regulation <- dem updated[match_row[1], "regulation"]</pre>
   } else {
     regulation <- "NA"
   lncRNA miRNA filtered[i, "regulation"] <- regulation</pre>
 cytoscape <- rbind(intersection table, lncRNA miRNA filtered)</pre>
 write.table(cytoscape, "../files/cytoscape to import.csv", row.names = FALSE)
 #this is the table to be imported to cytoscape in order to draw the network
 #these are the final mRNA that will be used for further PPI network and enrichment analysis
 write.table(unique(intersection_table$target), "../files/deg_intersection.txt", row.names = FALSE, col.names = FA
 LSE)
 #getting top 30 hub genes from both PPI and ceRNA network,
 ppi hub <- read.csv("../files/PPI hub.csv", header = TRUE) #exported from cytoscape
 cerna hub <- read.csv("../files/ceRNA hub.csv", header = TRUE) #exported from cytoscape
 cerna hub$name <- gsub('"', '', cerna hub$name)</pre>
 #The 3 key genes found (venn diagramm showed) by intersecting PPI and ceRNA hub genes
 result <- merge(ppi hub, cerna hub, by.x = "display.name", by.y = "name")
 result$display.name
 #to use in venn diagram
 write.csv(ppi hub$display.name, "../files/ppi hub genes.txt", row.names = FALSE, col.names = FALSE)
 write.csv(cerna_hub$name, "../files/cerna_hub_genes.txt", row.names = FALSE, col.names = FALSE)
code2.R
 library(dplyr)
 setwd("~/Documents/capstone/rscripts/")
 #This is the code that I tried to find the genes that are targets to the three miRNA using 3 DB
 #the search was done using the websites of the databases and the tables were downloaded and used here
 pred_mirDB <- read.csv("../files/mirDB.csv")</pre>
 pred_mirTarBase <- read.csv("../files/search_result-2.csv")</pre>
 target1<- read.csv("../files/Copy of TargetScan8.0 miR-140-5p.predicted targets.csv")
 target2 <- read.csv("../files/Copy of TargetScan8.0__miR-28-5p_708-5p.predicted_targets.csv")</pre>
 target3 <- read.csv("../files/Copy of TargetScan8.0 miR-532-5p.predicted targets.csv")</pre>
 target2 <- subset(target2, target2$Representative.miRNA == "hsa-miR-28-5p")</pre>
 target1 <- target1[, -9]
 target2 <- target2[, -9]</pre>
 target3 <- target3[, -9]</pre>
 pred_targetScan1 <- union(target1, target2)</pre>
 pred_targetScan <- union(pred_targetScan1, target3)</pre>
 #the one of target scan I performed it one by one so I had to do the union
 #here renaming the columns to be the same so I could later join them
```

colnames(pred_targetScan)[colnames(pred_targetScan)=="Target.gene"] <- "Target"</pre> colnames(pred_targetScan)[colnames(pred_targetScan)=="Representative.miRNA"] <- "miRNA"</pre> #taking only miRNA and target from tables because i am only interested in them pred_mirDB <- pred_mirDB[, c("miRNA", "Target")]</pre>

colnames(pred_mirDB)[colnames(pred_mirDB)=="Gene.Symbol"] <- "Target"</pre> colnames(pred_mirDB)[colnames(pred_mirDB)=="miRNA.Name"] <- "miRNA"</pre>

pred_mirTarBase <- pred_mirTarBase[, c("miRNA", "Target")]</pre> pred_targetScan <- pred_targetScan[, c("miRNA", "Target")]</pre> #Combining all three tables together, without duplicates

```
combined_table <- rbind(pred_mirDB, pred_mirTarBase, pred_targetScan)</pre>
 combined_table<-unique(combined_table)</pre>
 #adding rows to take count if one row is found in a table or not
 combined_table$mirDB <- integer(nrow(combined_table))</pre>
 combined_table$targetScan <- integer(nrow(combined_table))</pre>
 combined_table$mirTarBase <- integer(nrow(combined_table))</pre>
 #conditions, 1 found, 0 not found
 combined table$mirDB <- ifelse(paste(combined_table$Target, combined_table$miRNA) %in%
                                    paste(pred_mirDB$Target, pred_mirDB$miRNA), 1, 0)
 combined_table$targetScan <- ifelse(paste(combined_table$Target, combined_table$miRNA) % in %
                                         paste(pred_targetScan$Target, pred_targetScan$miRNA), 1, 0)
 combined_table$mirTarBase <- ifelse(paste(combined_table$Target, combined_table$miRNA) % in %
                                         paste(pred_mirTarBase$Target, pred_mirTarBase$miRNA), 1, 0)
 #summing, then taking only ones with sum greater or equal than 2
 combined_table$Sum <- rowSums(combined_table[,3:5])</pre>
 subset_table <- subset(combined_table, Sum >= 2)
 temp <- subset(head(subset_table, n=20))</pre>
 #genes intersection predicted vs DEGs, found 17, not enough
 subset_table_intersection <- subset(subset_table, subset_table[,2] %in% deg$Gene.symbol)
 array<- unique(subset_table_intersection$Target)</pre>
 write.table(unique(subset_table_intersection)$Target, "../files/genes_intersection.txt", row.names = FALSE, col.n
 ames = FALSE)
 x <- subset_table_intersection</pre>
 y <- lncRNA_miRNA_filtered[, 1:2]</pre>
 merged <- merge(x, y, by.x = "miRNA", by.y = "target")</pre>
 merged <- merged[, c("source", "miRNA", "Target", "mirDB", "targetScan", "mirTarBase", "Sum")]</pre>
 colnames(merged)[1] <- "lncRNA"</pre>
 write.table(merged, "../files/cytoscape_to_import2.csv", row.names = FALSE)
 #table to view network
enrichment.R
 setwd("~/Documents/capstone/rscripts/")
 load(dplyr)
 #This table was downloaded from DAVID database, and taking top 5 of each category by P value
 #PS the data of table is already ordered by P.value
 chart<-read.table("../files/chart_4CBB66A493291681316993753.txt", sep="\t", header = TRUE)
 cc <- chart[grep("^GOTERM_CC", chart$Category), ][1:5, ]</pre>
 bp <- chart[grep("^GOTERM_BP", chart$Categorry), ][1:5, ]</pre>
 mf <- chart[grep("^GOTERM_MF", chart$Category), ][1:5, ]</pre>
 kegg <- chart[grep("^KEGG", chart$Category), ][1:5, ]</pre>
 #Restructuring the tables
 CC <- CC %>%
   mutate(Category = "CC") %>%
   select(Category, everything())
 cc < -cc[,c(1:3, 5)]
 bp <- bp %>%
   mutate(Category = "BP") %>%
   select(Category, everything())
 bp < -bp[,c(1:3, 5)]
```

load(dplyr)

mf <- mf %>%

mf < -mf[,c(1:3, 5)]

kegg <- kegg %>%

kegg < -kegg[,c(1:3, 5)]

temp <- enrichement

xlab("Count") + ylab("Row") +

enrichment2.R

library(ggplot2)

#Binding all rows together

mutate(Category = "MF") %>%

select(Category, everything())

mutate(Category = "KEGG") %>% select(Category, everything())

enrichement<- rbind(cc, bp, mf, kegg)</pre>

temp\$Term <- substr(temp\$Term, 1, 10)</pre>

geom_bar(stat = "identity") +

create a bar plot of the enrichment data

ggtitle("Enrichement Data by Row Name")

setwd("~/Documents/capstone/rscripts/")

write.csv(enrichement, "../files/enrichement.csv")

#PS the data of table is already ordered by P.value

write.csv(enrichement, "../files/enrichement2.csv")

ggplot(enrichement, aes(x = Count, y = reorder(Term, Count), fill = Category)) +

#This table was downloaded from DAVID database, and taking top 5 of each category by P value

chart<-read.table("../files/chart_14FC45D11CE21683469320556.txt", sep="\t", header = TRUE)</pre>

```
cc <- chart[grep("^GOTERM_CC", chart$Category), ][1:5, ]</pre>
bp <- chart[grep("^GOTERM BP", chart$Category), ][1:5, ]</pre>
mf <- chart[grep("^GOTERM_MF", chart$Category), ][1:5, ]</pre>
kegg <- chart[grep("^KEGG", chart$Category), ][1:5, ]</pre>
#Restructuring the tables
cc <- cc %>%
  mutate(Category = "CC") %>%
  select(Category, everything())
cc < -cc[,c(1:3, 5)]
bp <- bp %>%
 mutate(Category = "BP") %>%
  select(Category, everything())
bp < -bp[,c(1:3, 5)]
mf <- mf %>%
  mutate(Category = "MF") %>%
  select(Category, everything())
mf < -mf[,c(1:3, 5)]
kegg <- kegg %>%
  mutate(Category = "KEGG") %>%
  select(Category, everything())
kegg < -kegg[,c(1:3, 5)]
#Binding all rows together
enrichement<- rbind(cc, bp, mf, kegg)</pre>
temp <- enrichement
temp$Term <- substr(temp$Term, 1, 10)</pre>
library(ggplot2)
# create a bar plot of the enrichment data
ggplot(enrichement, aes(x = Count, y = reorder(Term, Count), fill = Category)) +
  geom bar(stat = "identity") +
  xlab("Count") +
  ylab("Row") +
  ggtitle("Enrichement Data by Row Name")
```