

RNA-seq Differential Expression

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Introduction

This analysis investigates hypoxia-induced transcriptional changes in prostate cancer cell lines LNCaP (androgen-sensitive) and PC3 (androgen-independent). Hypoxia is a major driver of tumor aggressiveness and lineage plasticity.

The goal is to identify:

- [1]Differentially expressed genes (DEGs) under hypoxia
- [2]Global expression patterns using PCA
- [3]Hypoxia-regulated biological pathways
- [4]Hallmark biological programs altered in LNCaP cells

Loading required libraries

These packages implement:

Differential expression statistics (DESeq2)
Data manipulation (tidyverse)
Visualization (ggplot2, pheatmap)
Pathway analysis (clusterProfiler, ReactomePA, fgsea)

Packages must be installed before knitting

```
library(DESeq2)
library(tidyverse)
library(pheatmap)
library(RColorBrewer)
library(matrixStats)
library(clusterProfiler)
library(ReactomePA)
library(org.Hs.eg.db)
library(fgsea)
```

Loading raw count matrix

```
# Read count matrix

counts <- read.csv("GSE106305_counts_matrix.csv",
row.names = "Geneid",
```

```

stringsAsFactors = FALSE)

counts <- counts[, sort(colnames(counts))]
# View the first few rows
head(counts)

##          LNCAP_Hypoxia_S1 LNCAP_Hypoxia_S2 LNCAP_Normoxia_S1
## ENSG000000000003      604           691           367
## ENSG000000000005       0            0            0
## ENSG000000000419     1995          2302          2160
## ENSG000000000457      554           607           433
## ENSG000000000460      275           350           379
## ENSG000000000938        2            2            2
##          LNCAP_Normoxia_S2 PC3_Hypoxia_S1 PC3_Hypoxia_S2 PC3_Normoxia_S1
## ENSG000000000003      380          1059          332           352
## ENSG000000000005       0            0            0            0
## ENSG000000000419     2454          1974          693           747
## ENSG000000000457      518           88            26            29
## ENSG000000000460      349           390           155           189
## ENSG000000000938        1            0            0            0
##          PC3_Normoxia_S2
## ENSG000000000003      971
## ENSG000000000005       0
## ENSG000000000419     1761
## ENSG000000000457      83
## ENSG000000000460     438
## ENSG000000000938        1

```

Creating sample metadata

```

condition <- factor(c(
rep("LNCAP_Hypoxia", 2),
rep("LNCAP_Normoxia", 2),
rep("PC3_Hypoxia", 2),
rep("PC3_Normoxia", 2)
))

colData <- data.frame(condition)
rownames(colData) <- colnames(counts)
head(colData)

```

```

##          condition
## LNCAP_Hypoxia_S1    LNCAP_Hypoxia
## LNCAP_Hypoxia_S2    LNCAP_Hypoxia
## LNCAP_Normoxia_S1   LNCAP_Normoxia
## LNCAP_Normoxia_S2   LNCAP_Normoxia
## PC3_Hypoxia_S1      PC3_Hypoxia
## PC3_Hypoxia_S2      PC3_Hypoxia

```

DESeq2 dataset

```

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

# Keep genes with at least 10 reads in at least 2 samples
keep <- rowSums(counts(dds) >= 10) >= 2
dds <- dds[keep, ]

```

Differential expression analysis

```

dds <- DESeq(dds)

#Normalized counts
norm_counts <- counts(dds, normalized = TRUE)
write.csv(norm_counts, "Normalized_counts.csv")

```

Exploratory analysis: PCA

```

#Variance-stabilizing transformation
vsd <- vst(dds, blind = TRUE)

pca_data <- plotPCA(vsd, intgroup = "condition", returnData = TRUE)
percentVar <- round(100 * attr(pca_data, "percentVar"))

p_pca <- ggplot(pca_data, aes(PC1, PC2, color = condition)) +
  geom_point(size = 3) +
  labs(x = paste0("PC1: ", percentVar[1], "%"),
       y = paste0("PC2: ", percentVar[2], "%"),
       title = "PCA (VST transformed)") +
  theme_minimal()

ggsave("PCA_plot.png", p_pca, width = 6, height = 5, dpi = 300)

```

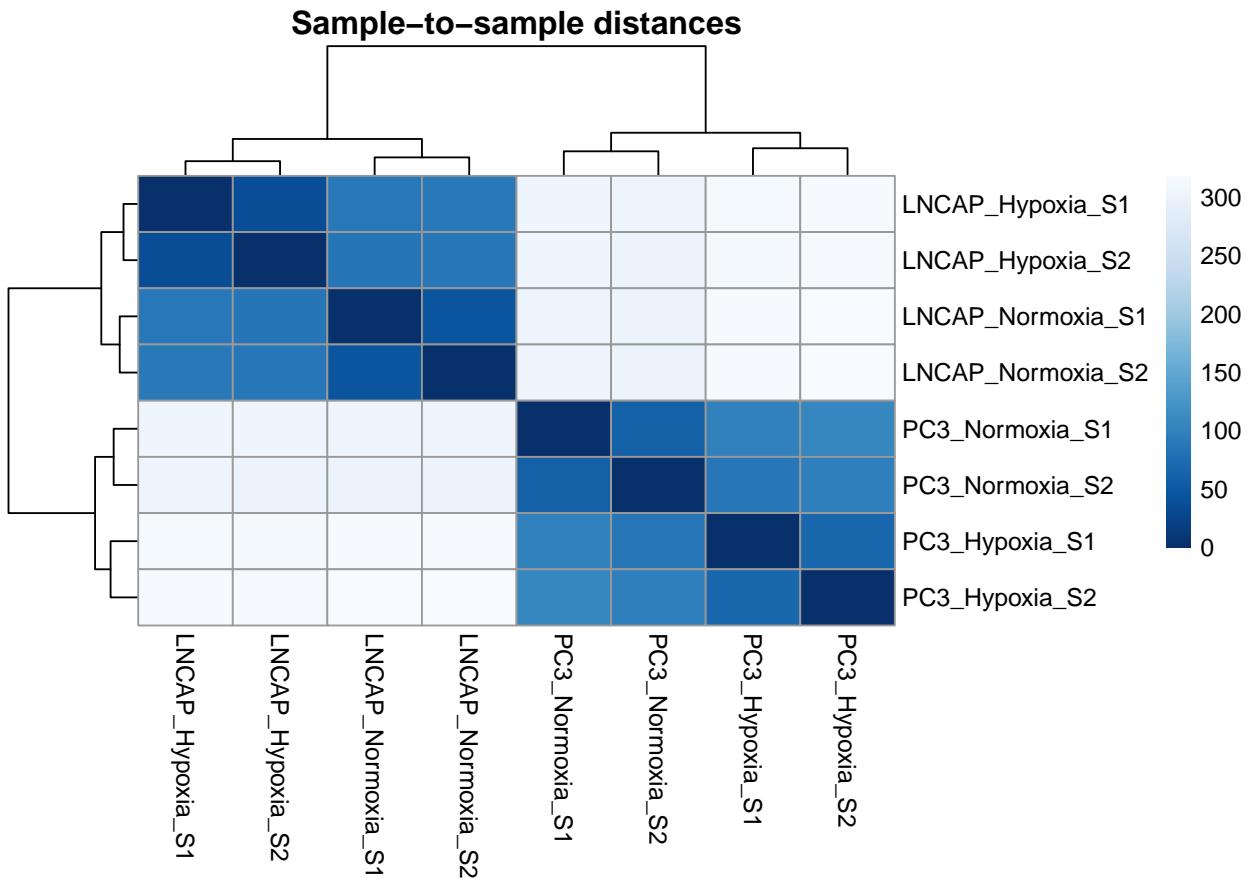
Sample-to-sample distance heatmap

```

sample_dists <- dist(t(assay(vsd)))
dist_matrix <- as.matrix(sample_dists)

png("sample_distance_heatmap.png", width = 1200, height = 1000, res = 300)
pheatmap(dist_matrix,
         clustering_distance_rows = sample_dists,
         clustering_distance_cols = sample_dists,
         col = colorRampPalette(rev(brewer.pal(9, "Blues")))(255),
         main = "Sample-to-sample distances")

```



```
dev.off()
```

```
## png
##   3
```

Expression distribution diagnostics

```
png("density_raw_vs_vst.png", width = 2000, height = 2000, res = 300)
par(mfrow = c(2, 2))
```

```
plot(density(counts(dds)[,1]), main = "Raw counts", xlab = "Expression")
plot(density(assay(vsd)[,1]), main = "VST counts", xlab = "Expression")
```

```
dev.off()
```

```
## pdf
##   2
```

Highly variable gene heatmap

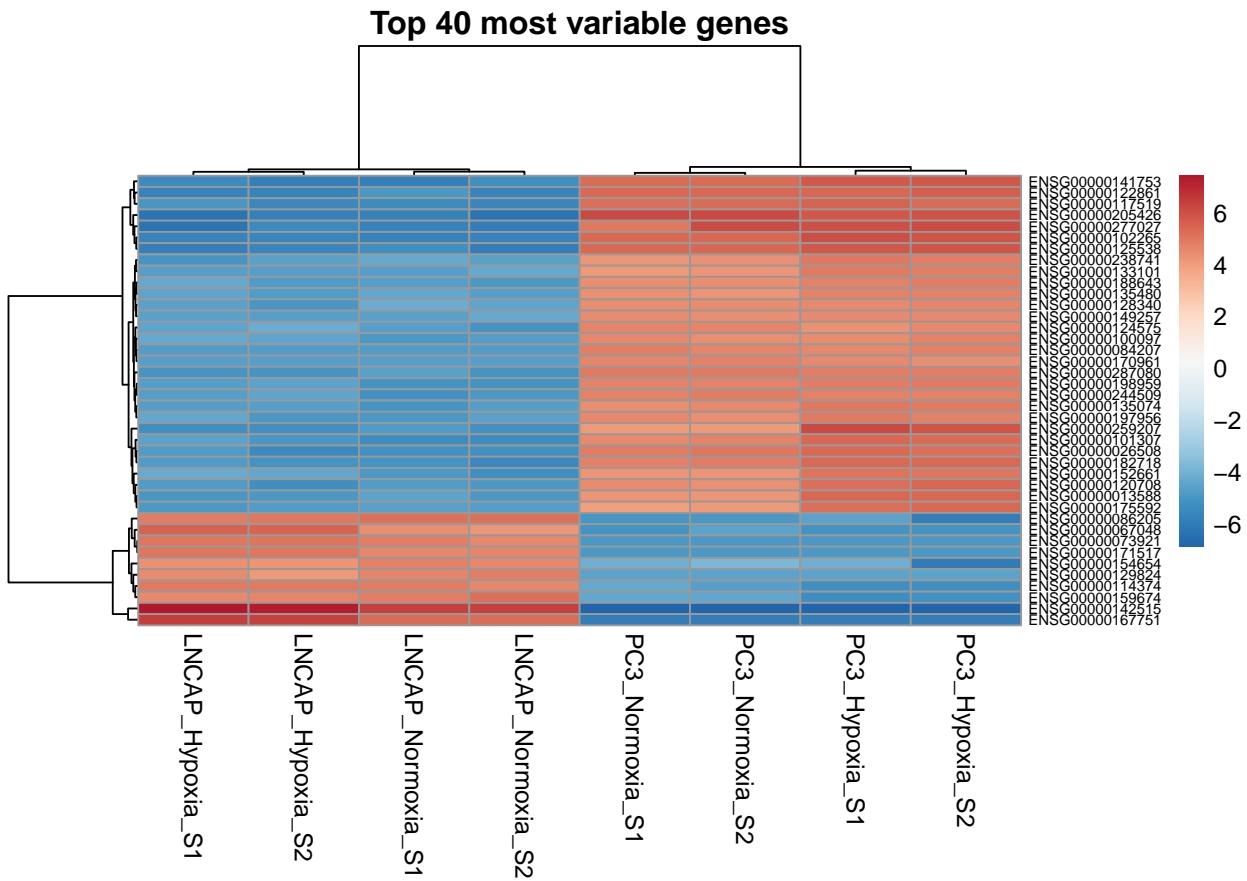
```

rv <- rowVars(assay(vsd))
top_genes <- order(rv, decreasing = TRUE)[1:40]

mat <- assay(vsd)[top_genes, ]
mat <- mat - rowMeans(mat)

png("top_variable_genes_heatmap.png", width = 1200, height = 1200, res = 300)
pheatmap(mat,
color = colorRampPalette(rev(brewer.pal(9, "RdBu"))))(255),
fontsize_row = 6,
main = "Top 40 most variable genes")

```



```

dev.off()

## png
##   3

dds_lncap <- dds[, grep("LNCAP", colnames(dds))]
dds_lncap$condition <- droplevels(dds_lncap$condition)
dds_lncap$condition <- relevel(dds_lncap$condition, ref = "LNCAP_Normoxia")

```

Extract differential expression results

```

dds_lncap <- DESeq(dds_lncap)

res_lncap <- results(dds_lncap,
contrast = c("condition",
"LNCAP_Hypoxia",
"LNCAP_Normoxia"))
write.csv(as.data.frame(res_lncap), "DEGs_LNCAP.csv")
head(res_lncap)

## log2 fold change (MLE): condition LNCAP_Hypoxia vs LNCAP_Normoxia
## Wald test p-value: condition LNCAP Hypoxia vs LNCAP Normoxia
## DataFrame with 6 rows and 6 columns
##           baseMean log2FoldChange      lfcSE      stat      pvalue
## <numeric>     <numeric> <numeric> <numeric> <numeric>
## ENSG00000000003  257.560    0.88827897  0.201949  4.398534 1.08985e-05
## ENSG00000000419 1110.185   -0.00297024  0.128404 -0.023132 9.81545e-01
## ENSG00000000457  264.313    0.39379117  0.192937  2.041031 4.12478e-02
## ENSG00000000460  169.022   -0.13896323  0.248008 -0.560318 5.75262e-01
## ENSG00000001036  146.122   -0.84216624  0.248853 -3.384187 7.13893e-04
## ENSG00000001084  612.416    0.78838940  0.165656  4.759190 1.94372e-06
##           padj
## <numeric>
## ENSG00000000003 1.15282e-04
## ENSG00000000419 9.91507e-01
## ENSG00000000457 1.21213e-01
## ENSG00000000460 7.45255e-01
## ENSG00000001036 4.30967e-03
## ENSG00000001084 2.46578e-05

```

MA plot

```

png("MA_plot_LNCAP.png", width = 800, height = 600, res = 150)
plotMA(res_lncap, ylim = c(-5, 5))
dev.off()

```

```

## pdf
## 2

```

Volcano plot

```

res_df <- as.data.frame(res_lncap) %>%
na.omit() %>%
mutate(regulation = case_when(
padj < 0.05 & log2FoldChange > 1 ~ "Upregulated",
padj < 0.05 & log2FoldChange < -1 ~ "Downregulated",
TRUE ~ "Not significant"
))

p_volcano <- ggplot(res_df, aes(log2FoldChange, -log10(padj), color = regulation)) +
geom_point(alpha = 0.6) +

```

```

theme_minimal() +
labs(title = "Volcano plot: LNCaP hypoxia")

ggsave("Volcano_plot_LNCAP.png", p_volcano, width = 6, height = 5, dpi = 300)

```

Pathway Analysis using GSEA

```

#Converting gene IDs (ENSEMBL → ENTREZ)

res_lncap_df <- as.data.frame(res_lncap)
res_lncap_df$ENSEMBL <- rownames(res_lncap)

id_map <- bitr(
  res_lncap_df$ENSEMBL,
  fromType = "ENSEMBL",
  toType = "ENTREZID",
  OrgDb = org.Hs.eg.db
)

head(res_lncap_df)

```

	baseMean	log2FoldChange	lfcSE	stat	pvalue
## ENSG00000000003	257.5595	0.888278972	0.2019489	4.39853403	1.089845e-05
## ENSG00000000419	1110.1847	-0.002970238	0.1284037	-0.02313203	9.815450e-01
## ENSG00000000457	264.3134	0.393791169	0.1929374	2.04103059	4.124779e-02
## ENSG00000000460	169.0223	-0.138963228	0.2480076	-0.56031842	5.752623e-01
## ENSG00000001036	146.1224	-0.842166237	0.2488533	-3.38418723	7.138930e-04
## ENSG00000001084	612.4156	0.788389402	0.1656562	4.75918976	1.943716e-06
##	padj	ENSEMBL			
## ENSG00000000003	1.152817e-04	ENSG00000000003			
## ENSG00000000419	9.915074e-01	ENSG00000000419			
## ENSG00000000457	1.212133e-01	ENSG00000000457			
## ENSG00000000460	7.452553e-01	ENSG00000000460			
## ENSG00000001036	4.309675e-03	ENSG00000001036			
## ENSG00000001084	2.465782e-05	ENSG00000001084			

```

#Merge mapping and removing duplicates
res_mapped <- res_lncap_df %>% left_join(id_map, by = "ENSEMBL") %>%
filter(!is.na(ENTREZID)) %>%
distinct(ENTREZID, .keep_all = TRUE)
head(res_mapped)

```

	baseMean	log2FoldChange	lfcSE	stat	pvalue	padj
## 1	257.5595	0.888278972	0.2019489	4.39853403	1.089845e-05	1.152817e-04
## 2	1110.1847	-0.002970238	0.1284037	-0.02313203	9.815450e-01	9.915074e-01
## 3	264.3134	0.393791169	0.1929374	2.04103059	4.124779e-02	1.212133e-01
## 4	169.0223	-0.138963228	0.2480076	-0.56031842	5.752623e-01	7.452553e-01
## 5	146.1224	-0.842166237	0.2488533	-3.38418723	7.138930e-04	4.309675e-03
## 6	612.4156	0.788389402	0.1656562	4.75918976	1.943716e-06	2.465782e-05
##	ENSEMBL	ENTREZID				
## 1	ENSG00000000003	7105				

```

## 2 ENSG00000000419      8813
## 3 ENSG00000000457      57147
## 4 ENSG00000000460      55732
## 5 ENSG00000001036      2519
## 6 ENSG00000001084      2729

```

Creating ranked gene list

```

gene_ranking <- res_mapped$log2FoldChange
names(gene_ranking) <- res_mapped$ENTREZID
gene_ranking <- sort(gene_ranking, decreasing = TRUE)

head(gene_ranking)

```

```

##      7040    441932     2681     79656 100128264     59350
## 7.680917 7.418113 7.264110 7.082984 6.712309 6.679098

```

Run GSEA using Reactome

```

gsea_reactome <- gsePathway(
  geneList = gene_ranking,
  organism = "human",
  pvalueCutoff = 0.05,
  verbose = FALSE
)

head(gsea_reactome@result)

```

```

##                               ID
## R-HSA-156842   R-HSA-156842
## R-HSA-9633012 R-HSA-9633012
## R-HSA-2408557 R-HSA-2408557
## R-HSA-9954716 R-HSA-9954716
## R-HSA-156827   R-HSA-156827
## R-HSA-9954714 R-HSA-9954714
##
## R-HSA-156842
## R-HSA-9633012
## R-HSA-2408557
## R-HSA-9954716 ZNF598 and the Ribosome-associated Quality Trigger (RQT) complex dissociate a ribosome
## R-HSA-156827
## R-HSA-9954714
##                               setSize enrichmentScore      NES      pvalue      p.adjust
## R-HSA-156842        90      -0.6450714 -2.436321 1.134338e-10 9.68191e-09
## R-HSA-9633012        98      -0.6327030 -2.423390 1.000000e-10 9.68191e-09
## R-HSA-2408557        92      -0.6375252 -2.420458 1.000000e-10 9.68191e-09
## R-HSA-9954716        95      -0.6324646 -2.420341 1.000000e-10 9.68191e-09
## R-HSA-156827        106      -0.6149729 -2.410724 1.000000e-10 9.68191e-09
## R-HSA-9954714        89      -0.6424837 -2.408874 1.000000e-10 9.68191e-09
##                               qvalue rank      leading_edge
## R-HSA-156842 8.962325e-09 5778 tags=89%, list=32%, signal=61%
## R-HSA-9633012 8.962325e-09 5778 tags=86%, list=32%, signal=59%

```

```

## R-HSA-2408557 8.962325e-09 5778 tags=86%, list=32%, signal=59%
## R-HSA-9954716 8.962325e-09 5778 tags=83%, list=32%, signal=57%
## R-HSA-156827 8.962325e-09 5778 tags=83%, list=32%, signal=57%
## R-HSA-9954714 8.962325e-09 5778 tags=88%, list=32%, signal=60%
##
## R-HSA-156842 51121/6154/6230/6139/6168/6205/6144/9045/61
## R-HSA-9633012 51121/6154/6230/1054/6139/8894/6168/6205/6144/1649/9045/6167/6
## R-HSA-2408557 51121/6154/6230/6139/6168/6205/6144/9045/
## R-HSA-9954716 51121/6154/6230/6139/6168/6205/6144/9045/
## R-HSA-156827 51121/6154/6230/3646/10480/6139/51386/8894/6168/6205/6144/1975/8664/9045/6167/8667/622
## R-HSA-9954714 51121/6154/6230/6139/6168/6205/61

```

Over-Representation Analysis (ORA)

```

#extracted significant genes
sig_genes <- res_mapped %>%
filter(padj < 0.05 & abs(log2FoldChange) > 1) %>%
pull(ENTREZID)

```

Run ORA using Reactome

```

ora.reactome <- enrichPathway(
  gene = sig_genes,
  organism = "human",
  pvalueCutoff = 0.05
)

```

Dotpot

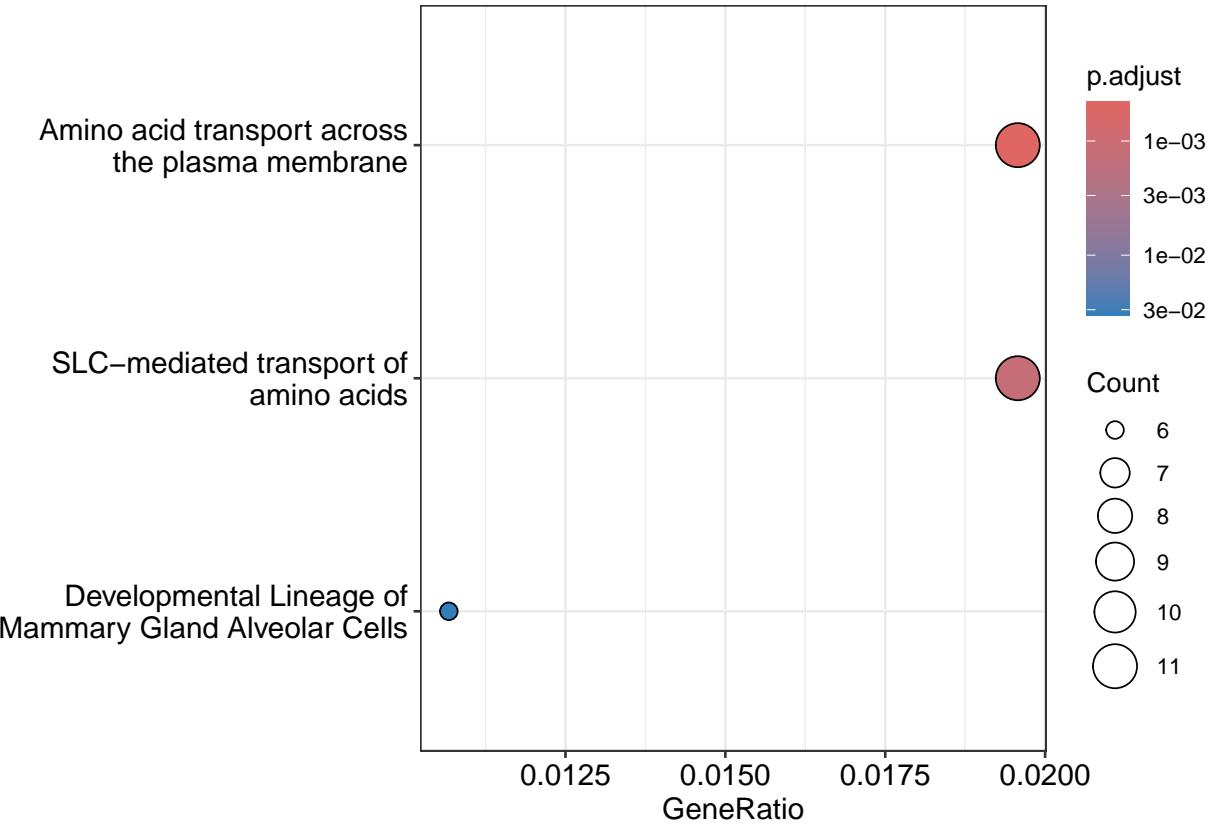
```

d <- dotplot(ora.reactome, showCategory = 20)

ggsave("ora.reactome_dotplot.png", plot = d, width = 10, height = 8, dpi = 300)

d

```



fgsea + Hallmark

Loaded fgsea and Hallmark gene sets

```
library(fgsea)

hallmark_sets <- gmtPathways("h.all.v7.0.symbols.gmt.txt")

# Convert ENSEMBL to SYMBOL
symbol_map <- bitr(
  res_lncap$ENSEMBL,
  fromType = "ENSEMBL",
  toType = "SYMBOL",
  OrgDb = org.Hs.eg.db
)

res_symbol <- res_lncap_df %>%
  left_join(symbol_map, by = c("ENSEMBL" = "ENSEMBL")) %>%
  filter(!is.na(SYMBOL)) %>%
  distinct(SYMBOL, .keep_all = TRUE)

ranked_symbols <- res_symbol$log2FoldChange
```

```

names(ranked_symbols) <- res_symbol$SYMBOL

ranked_symbols <- sort(ranked_symbols, decreasing = TRUE)

```

Run fgsea

```

fgsea_res <- fgsea(
pathways = hallmark_sets,
stats = ranked_symbols,
minSize = 15,
maxSize = 500,
nperm = 1000
)

fgsea_res[order(padj), .(pathway, NES, padj)][1:6]

```

```

##      pathway   NES   padj
##      <char> <num> <num>
## 1:    <NA>     NA     NA
## 2:    <NA>     NA     NA
## 3:    <NA>     NA     NA
## 4:    <NA>     NA     NA
## 5:    <NA>     NA     NA
## 6:    <NA>     NA     NA

```

Visualized Hallmark enrichment

```

waterfall_plot <- function(fgsea_res, graph_title) {

  fgsea_df <- as.data.frame(fgsea_res)

  p <- fgsea_df %>%
    mutate(short_name = str_split_fixed(pathway, "_", 2)[,2]) %>%
    ggplot(aes(x = reorder(short_name, NES), y = NES)) +
    geom_bar(stat = "identity", aes(fill = padj < 0.05)) +
    coord_flip() +
    labs(
      x = "Hallmark Pathway",
      y = "Normalized Enrichment Score",
      title = graph_title
    ) +
    theme(
      axis.text.y = element_text(size = 7),
      plot.title = element_text(hjust = 0.5)
    )

  return(p)
}

```

```
p <- waterfall_plot(  
  fgsea_res,  
  "Hallmark pathways altered by hypoxia in LNCaP cells"  
)  
  
ggsave(  
  filename = "Hallmark_fgsea_waterfall_LNCaP.png",  
  plot = p,  
  width = 8,  
  height = 6,  
  dpi = 300  
)
```