

# Sahand\_Adibnia\_NucleoTIDE\_Hackathon\_Project\_Documentation

The formation of neuromelanin in the brain is a phenomenon that is known to be unique to humans [1]. Animals do not accumulate the pigment. It is observed to form only in certain catecholaminergic neurons, namely the dopaminergic neurons of the substantia nigra pars compacta (SNpc) [2]. Interestingly, neuromelanin is closely linked to Parkinson's disease, as it is these dopaminergic neuromelanin-containing neurons that are exclusively lost in Parkinson's [2].

Currently, it is assumed that neuromelanin formed in dopaminergic neurons is a product of dopamine oxidation, a process not known to be catalyzed by any particular enzymes [3-5]. However, studies in animal models have failed to produce neuromelanin by solely increasing dopamine levels and dopamine oxidation [1]. Additionally, neuromelanin does not form in all catecholaminergic neurons. In fact, the dopaminergic neurons of the ventral tegmental area (VTA) generally lack neuromelanin [2], suggesting that some other molecular mechanism may be involved in neuromelanin formation.

The goal of this project is to use bioinformatics and gene expression data to give some insight into the formation of neuromelanin. Learning more about these molecular pathways may even provide some insight into the mechanisms that cause the selective death of neuromelanin-containing neurons in Parkinson's disease. The primary method by which this is achieved is by analyzing differences in gene expression between the SNpc, a brain region containing a large population of melanized dopaminergic neurons, and the VTA, a brain region containing a large population of non-melanized dopaminergic neurons.

The Allen Human Brain Atlas (<https://human.brain-map.org/microarray/search>) provides microarray expression data of numerous different brain regions in 6 neurologically and neuropathologically normal individuals, including the SNpc and VTA. It also provides a function that does differential expression analysis of different brain regions, outputting p-values and fold change values for each probe. However, it only outputs the fold change for probes that are upregulated in the target structure with respect to the contrast structure. Therefore, this search had to be done twice: once with the SNpc as the target structure and the VTA as the contrast structure, and another time with the VTA as the target structure and the SNpc as the contrast structure. This data could be directly downloaded from the site. However, it could only be downloaded 2000 probes at a time. This was rather tedious, and knowing that many probes are not significantly differentially expressed anyway, only the first 10000 probes with the highest fold change were downloaded, giving a total of 20000 probes.

READING IN P-VALUES AND FOLD CHANGE OF TOP 10000 DIFERENTIALLY EXPRESSED PROBES

```
library(readr)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag
```

```

## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union

# SNpc target, VTA contrast
files_1 <- list.files(path = "/Users/rezaadibnia/Desktop/PD NM Project/mDBR - VTA vs SNpc/SNpc target,
SNt_VTAc_probe_list <- lapply(files_1, read_csv)

##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),
##   'fold-change' = col_double()
## )

##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),
##   'fold-change' = col_double()
## )

##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),

```

```

## 'fold-change' = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),
##   'fold-change' = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),
##   'fold-change' = col_double()
## )

```

```

SNt_VTAc_probes <- data.frame(bind_rows(SNt_VTAc_probe_list))

```

```

# VTA target, SNpc contrast

```

```

files_2 <- list.files(path = "/Users/rezaadibnia/Desktop/PD NM Project/mDBR - VTA vs SNpc/VTA target, SNpc",
VTAt_SNc_probe_list <- lapply(files_2, read_csv)

```

```

##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),

```

```

## 'fold-change' = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),
##   'fold-change' = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),
##   'fold-change' = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   id = col_double(),
##   name = col_character(),
##   'gene-id' = col_double(),
##   'gene-symbol' = col_character(),
##   'gene-name' = col_character(),
##   'entrez-id' = col_double(),
##   chromosome = col_character(),
##   'start-position' = col_character(),
##   'end-position' = col_character(),
##   p = col_double(),
##   'fold-change' = col_double()
## )
##
##
## -- Column specification -----
## cols(

```

```
## id = col_double(),
## name = col_character(),
## 'gene-id' = col_double(),
## 'gene-symbol' = col_character(),
## 'gene-name' = col_character(),
## 'entrez-id' = col_double(),
## chromosome = col_character(),
## 'start-position' = col_character(),
## 'end-position' = col_character(),
## p = col_double(),
## 'fold-change' = col_double()
## )
```

```
VTAt_SNC_probes <- data.frame(bind_rows(VTAt_SNC_probe_list))
```

CORRECTING P-VALUES AND TRANSFORMING FOLD CHANGES The definition of the fold change of B with respect to A is  $B/A$ . Therefore, the fold change of A with respect to B is  $A/B$ . In other words, the fold change of gene expression in the SNpc with respect to the VTA is equal to  $1/(\text{the fold change of gene expression in the VTA with respect to the SNpc})$ . By taking the reciprocal of the fold change values with VTA target and SNpc contrast, we get the equivalent fold change for SNpc target and VTA contrast, allowing us to combine the differential expression analysis of all 20000 probes into one data frame.

```
# BH-correcting p-values (Benjamini-Hochberg Procedure)
SNT_VTAc_probes$p <- p.adjust(SNT_VTAc_probes$p, method="BH")
VTAt_SNC_probes$p <- p.adjust(VTAt_SNC_probes$p, method="BH")

# Converting VTA target, SNpc contrast fold change to SNpc target, VTA contrast fold change
VTAt_SNC_probes$fold.change <- 1/(VTAt_SNC_probes$fold.change)

# Combining the data frames containing p values and fold change
p_and_FC <- data.frame(bind_rows(SNT_VTAc_probes, VTAt_SNC_probes))
p_and_FC <- p_and_FC[,c(4,10:11)]
```

## VISUALIZING DIFFERENTIALLY EXPRESSED PROBES - VOLCANO PLOT

```
library(ggplot2)

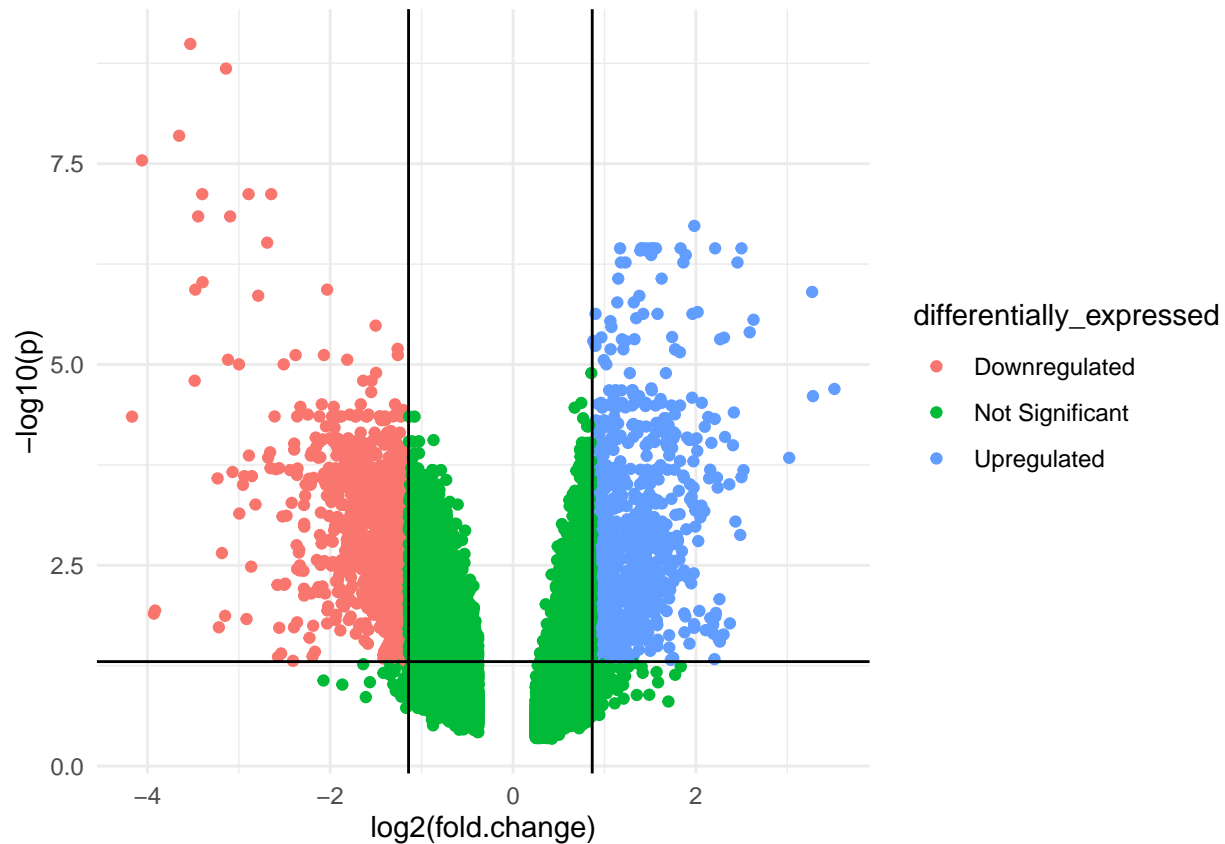
# Vertical lines for log2(fold change) thresholds, horizontal line for the p-value threshold
# Top 10% of upregulated probes and bottom 10% of downregulated probes are chosen
horz_min = quantile(log2(VTAt_SNC_probes$fold.change), 0.10)
horz_max = quantile(log2(SNT_VTAc_probes$fold.change), 0.90)

# Add a column to p_and_FC data frame indicating whether probe is upregulated, downregulated, or not significant
p_and_FC$differentially_expressed <- "Not Significant"
# If log2Foldchange > horz_max and pvalue < 0.05, set as "UP"
p_and_FC$differentially_expressed[log2(p_and_FC$fold.change) > horz_max & p_and_FC$p < 0.05] <- "Upregulated"
# If log2Foldchange < horz_min and pvalue < 0.05, set as "DOWN"
p_and_FC$differentially_expressed[log2(p_and_FC$fold.change) < horz_min & p_and_FC$p < 0.05] <- "Downregulated"

# Construct volcano plot
volcano_plot <- ggplot(data=p_and_FC, aes(x=log2(fold.change), y=-log10(p), col=differentially_expressed)) +
  geom_point() +
  theme_minimal() +
```

```
geom_vline(xintercept=c(horz_min, horz_max), col="black") +
geom_hline(yintercept=-log10(0.05), col="black")
```

*# The significantly differentially expressed probes are found in the upper-left and upper-right corners*  
volcano\_plot



FILTERING FOR  $P < 0.05$  and  $FC > 90\text{th PERCENTILE (TOP 10\%)}$  p-values were previously corrected when constructing the volcano plot

```
library(dplyr)
```

*# Filters for top 10% of genes with highest FC compared to all probes, not just significantly different*

*# SNpc target, VTA contrast*

```
SNt_VTAc_probes <- SNt_VTAc_probes %>% filter(SNt_VTAc_probes$p < 0.05 & SNt_VTAc_probes$fold.change > 0)
```

*# VTA target, SNpc contrast*

```
VTAt_SNC_probes <- VTAt_SNC_probes %>% filter(VTAt_SNC_probes$p < 0.05 & VTAt_SNC_probes$fold.change < 0)
```

READING IN EXPRESSION DATA FOR TOP 10000 DIFFERENTIALLY EXPRESSED PROBES This data is actually not necessary to overall analysis, but is good to have for reference.

```
library(readr)
library(dplyr)
```

```

read_csv_labeled_header <- function(file) {
  labels = c("Probe_IDs", "SN_1", "SN_2", "SN_3", "SN_4", "SN_5", "SN_6", "VTA_1", "VTA_2", "VTA_3", "VTA_4", "VTA_5", "VTA_6")
  read_csv(file, col_names = labels)
}

# SNpc target, VTA contrast
files_expr_1 <- list.files(path = "/Users/rezaadibnia/Desktop/PD NM Project/mDBR - VTA vs SNpc/SNpc target", pattern = "*.csv", full.names = TRUE)
Snt_VTAc_expr_list <- lapply(files_expr_1, read_csv_labeled_header)

```

```

##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),

```

```

## SN_5 = col_double(),
## SN_6 = col_double(),
## VTA_1 = col_double(),
## VTA_2 = col_double(),
## VTA_3 = col_double(),
## VTA_4 = col_double(),
## VTA_5 = col_double(),
## VTA_6 = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )

```

```

Snt_VTAc_expr <- data.frame(bind_rows(Snt_VTAc_expr_list))

```

```

# VTA target, SNpc contrast

```

```

files_expr_2 <- list.files(path = "/Users/rezaadibnia/Desktop/PD NM Project/mDBR - VTA vs SNpc/VTA targ

```

```

VTA_SNc_expr_list <- lapply(files_expr_2, read_csv_labeled_header)

```

```

##

```

```

## -- Column specification -----

```



```

## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
##
##
## -- Column specification -----

```

```
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
##
##
## -- Column specification -----
## cols(
##   Probe_IDs = col_double(),
##   SN_1 = col_double(),
##   SN_2 = col_double(),
##   SN_3 = col_double(),
##   SN_4 = col_double(),
##   SN_5 = col_double(),
##   SN_6 = col_double(),
##   VTA_1 = col_double(),
##   VTA_2 = col_double(),
##   VTA_3 = col_double(),
##   VTA_4 = col_double(),
##   VTA_5 = col_double(),
##   VTA_6 = col_double()
## )
```

```
VTAt_SNC_expr <- data.frame(bind_rows(VTAt_SNC_expr_list))
```

JOINING EXPRESSION AND PROBE DATA FRAMES INTO ONE Note that probe data frames contain p-values and fold changes.

```
library(dplyr)

SNt_VTAc_expr <- SNt_VTAc_expr %>% inner_join(SNt_VTAc_probes, by = c("Probe_IDs" = "id"))
VTAt_SNC_expr <- VTAt_SNC_expr %>% inner_join(VTAt_SNC_probes, by = c("Probe_IDs" = "id"))

expr <- data.frame(bind_rows(SNt_VTAc_expr, VTAt_SNC_expr))
```

After filtering for probes that had p-values < 0.05 and significant fold changes, the probes needed to be aligned to individual genes. To do this, the expression data for all probes in all brain regions in all 6 donors had to be downloaded from Allen Human Brain Atlas (AHBA). The probes lacking Entrez IDs were first filtered out. Conveniently, AHBA provides a presence/absence flag that determines whether a certain probe's expression is well above the background levels or not. This flag was used to filter out any probes that are not expressed above the background in less than 1% of all samples. There may be multiple probes that align to the same gene, but only one probe per gene can be selected. Therefore, a systematic method was used

to align probes to genes using the collapseRows function from the WGCNA package, similar to methods previously used in the literature [6]:

- If there is only one probe for a gene, that one is chosen
- If there are two probes for one gene, the one with the max variance across all samples is chosen (method = maxRowVariance)
- If there more than two probes for one gene, the probe with the highest connectivity (summed adjacency) is chosen (connectivityBasedCollapsing = TRUE)

## ALIGNING PROBES TO GENES

```
# Map probes to genes
# First download 'Complete normalized microarray datasets' for all six donors from http://human.brain-m
# Store the downloaded files in a AHBA download directory
library(WGCNA)

## Loading required package: dynamicTreeCut

## Loading required package: fastcluster

##
## Attaching package: 'fastcluster'

## The following object is masked from 'package:stats':
##
##      hclust

##

##
## Attaching package: 'WGCNA'

## The following object is masked from 'package:stats':
##
##      cor

##### Load data #####

# Set donor names
donorNames <- c("donor9861", "donor10021", "donor12876", "donor14380", "donor15496", "donor15697")
names(donorNames) <- donorNames

# Location of AHBA directories
ahba_download <- "/Users/rezaadibnia/Desktop/AHBA Full Data/Microarray" # Where downloaded files are st

# Read in probe info (files are the same for each donor)
probeInfo <- read.csv(paste0(ahba_download, "/normalized_microarray_donor9861/Probes.csv")) # Same for

# Read expression data for each donor (probes x samples)
brainExpr <- lapply(donorNames, function(d){
  file1 <- paste0(ahba_download, "/normalized_microarray_", d, "/MicroarrayExpression.csv")
```

```

e <- read.csv(file1, header = FALSE)
rownames(e) <- e[,1]
e <- e[,-1]
file2 <- paste0(ahba_download, "/normalized_microarray_", d, "/SampleAnnot.csv")
sample_annotation <- read.csv(file2)
colnames(e) <- sample_annotation$structure_id
e
})

##### Filter probes based on concatenated expression data from all donors #####

# Probes with missing Entrez IDs
probes_missing_entrezID <- is.na(probeInfo$entrez_id) # same in all donors
print(paste(sum(probes_missing_entrezID), "probes with missing entrez IDs"))

## [1] "10521 probes with missing entrez IDs"

# Probes with expression well above background in at least 1% of samples in all donors
pa_call <- lapply(donorNames, function(d){
  file <- paste0(ahba_download, "/normalized_microarray_", d, "/PACall.csv")
  pa <- read.csv(file, header = FALSE)
  rownames(pa) <- pa[,1]
  pa[, -1]
})
pa_concat <- Reduce(cbind, pa_call)
sum <- rowSums(pa_concat)
low_presence <- sum < 0.01*ncol(pa_concat)
print(paste(sum(low_presence), "probes present in <1% of samples"))

## [1] "6068 probes present in <1% of samples"

# Concatenate data across all donors and filter probes by ID
expr_concat <- Reduce(cbind, brainExpr)
expr_concat <- expr_concat[!(probes_missing_entrezID | low_presence),]
probeInfo <- probeInfo[!(probes_missing_entrezID | low_presence), ]

probe2gene <- collapseRows(expr_concat,
                           rowGroup = probeInfo$entrez_id,
                           rowID = probeInfo$probe_id,
                           method = "maxRowVariance",
                           connectivityBasedCollapsing = TRUE
                           )
selected_probes <- probe2gene$selectedRow
entrez_id <- probeInfo$entrez_id[selected_probes]
entrez_order <- order(entrez_id)

probeInfo <- probeInfo[selected_probes, ]
probeInfo <- probeInfo[entrez_order,]

```

ALIGNING SIGNIFICANTLY DIFFERENTIALLY EXPRESSED PROBES TO GENES

```
library(dplyr)

# Combine data frames by probe ID to filter out all probes not corresponding to a gene
final_expr <- expr %>% inner_join(probeInfo, by = c("Probe_IDs" = "probe_id"))

# Remove unnecessary and repeated columns
final_expr <- final_expr[,-c(14, 18:21, 24:29)]
final_expr$fold.change <- sort(final_expr$fold.change, decreasing = TRUE)

final_expr[,c(15, 17:18)]
```

##	gene.symbol	p	fold.change
## 1	AGTR1	2.017857e-05	11.44100000
## 2	FEZF1	1.250000e-06	9.65900000
## 3	LMOD3	1.452381e-04	8.11800000
## 4	NPNT	2.771875e-06	6.19600000
## 5	CELSR1	3.971429e-06	6.01800000
## 6	RSP02	2.067010e-04	5.74200000
## 7	PYGL	3.581818e-07	5.65300000
## 8	ATP2A3	9.025070e-04	5.40100000
## 9	OC90	3.955056e-05	5.33600000
## 10	SRPX2	1.012987e-04	5.30000000
## 11	SLC6A3	8.355197e-03	4.78400000
## 12	F2RL2	2.553991e-04	4.69400000
## 13	SLC18A2	1.238258e-02	4.65600000
## 14	CBLN1	3.581818e-07	4.62900000
## 15	ENPP1	4.760417e-05	4.61700000
## 16	CCL3	4.660821e-02	4.60700000
## 17	FOXA1	2.041451e-04	4.45900000
## 18	NTSR1	2.030387e-02	4.30700000
## 19	SLC35D3	6.686930e-04	4.26400000
## 20	TPBG	3.000000e-05	4.18400000
## 21	ALDH1A1	5.714286e-04	4.17400000
## 22	FAM19A3	8.000000e-04	4.14100000
## 23	HTR1F	5.796178e-04	4.06500000
## 24	SSTR1	2.223077e-06	4.05500000
## 25	RDH12	1.191358e-04	4.04300000
## 26	SDC1	6.574924e-04	3.98500000
## 27	SLC10A4	1.899718e-02	3.94700000
## 28	SMPX	3.409639e-04	3.93300000
## 29	GFRA1	1.603448e-04	3.92700000
## 30	MATN3	2.343333e-06	3.89900000
## 31	VAV3	4.333333e-07	3.70500000
## 32	RGAG1	2.157395e-02	3.67100000
## 33	ZAR1	1.256367e-02	3.66400000
## 34	LOC441052	2.567442e-04	3.65600000
## 35	CYP2S1	3.234568e-05	3.54900000
## 36	EBF3	1.588235e-04	3.54800000
## 37	KCNJ6	7.291667e-04	3.53700000
## 38	PROK2	2.684887e-03	3.53100000
## 39	RGS16	7.448071e-04	3.42900000
## 40	APOC1	7.365079e-05	3.42700000
## 41	RGS4	6.468085e-06	3.42000000

## 42	COL12A1	6.520833e-06	3.41900000
## 43	LRRRC55	4.188889e-05	3.41000000
## 44	FAM70A	8.308824e-05	3.37400000
## 45	IL1RAPL2	2.784810e-03	3.36400000
## 46	B3GALT1	4.555556e-06	3.34100000
## 47	DRD2	1.095541e-04	3.34100000
## 48	PTPN3	1.725838e-03	3.33200000
## 49	KLHL1	4.703704e-04	3.32400000
## 50	FST	3.763889e-03	3.31500000
## 51	SV2C	1.351515e-04	3.26900000
## 52	APOC2	3.085106e-03	3.26500000
## 53	CADPS2	5.669811e-05	3.25100000
## 54	LPO	1.177489e-02	3.20600000
## 55	HPGD	9.809783e-04	3.20100000
## 56	AKAP14	2.144112e-03	3.20000000
## 57	GALNTL6	1.575314e-03	3.19900000
## 58	FAM71F1	4.663158e-05	3.15000000
## 59	RESP18	5.129151e-03	3.13600000
## 60	IZUMO2	1.714851e-03	3.12100000
## 61	NEB	1.797732e-03	3.09800000
## 62	GPR161	9.615385e-04	3.09100000
## 63	ITPR1	8.523810e-07	3.08500000
## 64	LOC100130344	1.161369e-03	3.06200000
## 65	CYP27C1	3.804945e-03	3.05600000
## 66	GBE1	6.939227e-03	3.02300000
## 67	TMEM200C	2.041451e-04	3.01600000
## 68	EN1	2.668731e-02	2.99800000
## 69	CNIH3	2.343333e-06	2.99100000
## 70	PGR	3.664773e-03	2.98800000
## 71	FLJ30901	1.161369e-03	2.97500000
## 72	GSG1L	5.223368e-04	2.97100000
## 73	OSBPL3	1.533755e-03	2.96200000
## 74	RERG	3.581818e-07	2.95600000
## 75	ST8SIA6	1.128713e-03	2.93400000
## 76	RET	1.192740e-02	2.93100000
## 77	KLF5	1.124378e-03	2.92500000
## 78	BEND4	5.801887e-03	2.90900000
## 79	TUB	4.710145e-04	2.89000000
## 80	LMX1B	1.359551e-03	2.86800000
## 81	KLHL13	2.096774e-05	2.86400000
## 82	SCUBE1	5.653207e-03	2.86100000
## 83	SIAH3	1.981818e-05	2.85300000
## 84	CAB39L	4.333333e-07	2.85100000
## 85	GPR26	5.894118e-03	2.84700000
## 86	C16orf93	3.581818e-07	2.83600000
## 87	FNDC1	3.674856e-02	2.83000000
## 88	GPRC5A	5.392039e-03	2.80800000
## 89	SLC2A13	9.225352e-05	2.78700000
## 90	PRKAR2B	3.815385e-07	2.75000000
## 91	CHRNA3	1.739766e-02	2.74500000
## 92	CLDN1	7.314410e-03	2.74200000
## 93	EFCAB5	3.017467e-04	2.73800000
## 94	SLITRK4	3.000000e-05	2.73700000
## 95	ST3GAL1	3.000000e-05	2.69600000

## 96	ZNF503	2.343333e-06	2.68400000
## 97	VGLL3	1.474026e-03	2.67800000
## 98	SNCA	1.731141e-03	2.67300000
## 99	TRIM58	6.387316e-03	2.65200000
## 100	GABRA4	2.784810e-03	2.63400000
## 101	LMO3	3.815385e-07	2.62600000
## 102	NCAPG	1.020725e-03	2.62400000
## 103	ZW10	2.612500e-02	2.61400000
## 104	OXTR	1.880819e-03	2.61200000
## 105	PCDH8	8.926554e-04	2.57400000
## 106	GPR98	1.927273e-03	2.55000000
## 107	PITX1	2.647607e-02	2.54400000
## 108	OPN3	2.814465e-03	2.52500000
## 109	EYA4	3.803596e-03	2.52000000
## 110	ANK1	3.012987e-05	2.51900000
## 111	DPYS	4.443022e-03	2.51700000
## 112	LOC100131532	5.012376e-03	2.51300000
## 113	C17orf82	4.846154e-06	2.51200000
## 114	SLC2A4	4.032476e-03	2.51200000
## 115	CACNB3	4.076923e-04	2.50900000
## 116	LOC100144602	9.419890e-04	2.50000000
## 117	C9orf135	4.180929e-02	2.49700000
## 118	RGNEF	1.899718e-02	2.49200000
## 119	SEZ6L	8.696884e-04	2.48500000
## 120	OSBPL10	1.772031e-03	2.48300000
## 121	INSM2	2.671614e-02	2.47800000
## 122	SLIT1	2.382182e-02	2.47500000
## 123	IGFBP4	2.209688e-02	2.47400000
## 124	ZWINT	2.026279e-02	2.47300000
## 125	SNN	5.669811e-05	2.47200000
## 126	KCND3	3.240964e-05	2.47000000
## 127	CHRNA6	1.359551e-03	2.46900000
## 128	SYT17	2.096774e-05	2.46500000
## 129	EPS8L1	2.164879e-02	2.46400000
## 130	PRMT8	2.261670e-02	2.45400000
## 131	LXN	2.074176e-02	2.44200000
## 132	TRPC6	1.420500e-02	2.43500000
## 133	PLEKHA5	1.279630e-05	2.42500000
## 134	RIMS1	3.767313e-03	2.41500000
## 135	FKBP1B	4.262136e-02	2.41400000
## 136	BNC2	2.085048e-02	2.41200000
## 137	KANK4	1.579167e-03	2.40400000
## 138	ARHGAP11A	6.939227e-03	2.40300000
## 139	CLPTM1L	2.477932e-02	2.39500000
## 140	PID1	9.463087e-05	2.39300000
## 141	DRP2	1.684211e-02	2.39100000
## 142	OLFM3	3.098291e-04	2.37600000
## 143	OR7G1	2.870813e-02	2.36900000
## 144	PKIB	5.796178e-04	2.36600000
## 145	GLA	4.393723e-02	2.36100000
## 146	AMPH	2.163624e-02	2.36100000
## 147	NTN1	1.511777e-03	2.34800000
## 148	CORO2A	5.368421e-07	2.34700000
## 149	VAT1	3.423529e-05	2.34300000

## 150	HIST1H2AM	1.000000e-03	2.33700000
## 151	CLTCL1	2.700893e-04	2.33000000
## 152	TNFRSF19	1.349315e-03	2.32700000
## 153	UNC13C	1.313084e-03	2.32700000
## 154	ST6GALNAC5	3.327769e-02	2.31500000
## 155	CRYGD	5.245098e-03	2.30900000
## 156	INSM1	4.464875e-02	2.30000000
## 157	UGGT1	1.440789e-03	2.28900000
## 158	POU3F2	4.853659e-06	2.28300000
## 159	NEFH	1.927273e-03	2.28100000
## 160	DOK6	2.096774e-05	2.27600000
## 161	ROBO2	3.056277e-04	2.27300000
## 162	B4GALT6	1.473913e-03	2.27200000
## 163	CLSTN2	6.856187e-03	2.27200000
## 164	FAM135B	2.684887e-03	2.25900000
## 165	CCDC85A	5.796178e-04	2.25100000
## 166	MCC	3.737255e-04	2.24700000
## 167	CPLX2	1.349315e-03	2.23900000
## 168	PRKCD	1.265823e-02	2.23700000
## 169	C18orf34	9.126679e-03	2.23400000
## 170	NOV	1.739766e-02	2.22900000
## 171	PBX1	5.420000e-05	2.21900000
## 172	GRAMD4	4.076923e-04	2.21500000
## 173	PADI1	1.407767e-02	2.20900000
## 174	GPRIN3	2.095563e-02	2.20200000
## 175	KMO	3.177711e-03	2.20000000
## 176	PDZD7	3.295964e-02	2.19300000
## 177	CSDC2	1.938406e-03	2.19200000
## 178	ENO3	3.012987e-05	2.19100000
## 179	ATP6V1B2	1.910946e-03	2.18800000
## 180	C22orf25	3.883197e-02	2.18700000
## 181	KIF26B	2.087719e-05	2.17500000
## 182	PLA2G7	3.327769e-02	2.17100000
## 183	ELAVL2	1.362416e-03	2.17100000
## 184	CTXN3	5.704142e-03	2.17000000
## 185	LINC00261	4.499295e-02	2.16600000
## 186	CUX2	2.348039e-04	2.16100000
## 187	COMTD1	4.663278e-03	2.14900000
## 188	MMD	1.103275e-03	2.14900000
## 189	LRRC3B	2.597015e-05	2.14700000
## 190	NRXN3	2.718978e-02	2.13800000
## 191	ANKRD29	5.859649e-05	2.13400000
## 192	C3orf80	2.825746e-03	2.13000000
## 193	LOC730091	2.628099e-03	2.12900000
## 194	KIF3C	3.151515e-03	2.12400000
## 195	SOX6	3.284024e-03	2.12200000
## 196	FABP3	1.927273e-03	2.11900000
## 197	SEMA3E	3.594080e-02	2.11700000
## 198	TAL2	8.120301e-05	2.11100000
## 199	ZNF385D	1.190821e-03	2.11100000
## 200	RNF144B	3.228700e-03	2.10900000
## 201	DAPL1	2.397172e-02	2.10800000
## 202	STYK1	3.036131e-02	2.10600000
## 203	FUT9	2.848297e-03	2.10400000



##	204	PCSK1	8.815920e-03	2.10400000
##	205	PIP5K1B	1.684211e-02	2.10100000
##	206	RELL2	2.161247e-02	2.10000000
##	207	SLC44A5	1.069182e-02	2.09900000
##	208	EPN3	3.548559e-02	2.09600000
##	209	SH3BGRL2	6.468085e-06	2.09600000
##	210	SYT1	5.796178e-04	2.09100000
##	211	GPX3	1.731141e-03	2.09000000
##	212	TRPM2	4.556962e-02	2.08700000
##	213	RNF43	7.660695e-03	2.08000000
##	214	MYOM1	2.292079e-04	2.07900000
##	215	CNTN4	3.258929e-03	2.07800000
##	216	ANKRD34A	2.223693e-02	2.07300000
##	217	VAV2	5.051903e-04	2.06600000
##	218	BSN	7.663158e-03	2.06300000
##	219	GLRA3	9.026290e-03	2.05600000
##	220	TPGS2	5.555556e-04	2.05300000
##	221	FXYP7	2.684887e-03	2.05000000
##	222	LOC158696	1.984626e-02	2.04600000
##	223	RCAN2	1.927273e-03	2.04100000
##	224	TTC39A	5.281174e-03	2.04000000
##	225	WIPF3	3.142530e-02	2.04000000
##	226	DYNC1I1	3.375551e-02	2.03900000
##	227	CHEK1	1.984626e-02	2.03600000
##	228	CREB3L1	1.857651e-02	2.03500000
##	229	TMEM159	2.700893e-04	2.03300000
##	230	DENND2A	1.020725e-03	2.03300000
##	231	PPM1J	1.851852e-02	2.03000000
##	232	CERS6	1.731141e-03	2.02800000
##	233	NEDD4L	3.409639e-04	2.02500000
##	234	TMEM241	8.189300e-03	2.02500000
##	235	FAM126A	8.240081e-03	2.02300000
##	236	SLC36A1	3.161180e-02	2.02000000
##	237	C15orf59	5.859649e-05	2.01900000
##	238	HMMR	1.376221e-02	2.01600000
##	239	CALN1	2.700893e-04	2.01500000
##	240	MAN1C1	5.669811e-05	2.01500000
##	241	RGS6	5.012376e-03	2.01100000
##	242	CRYBB1	4.359712e-02	2.01000000
##	243	EPB49	4.735632e-02	2.00800000
##	244	STK17A	1.684211e-02	2.00300000
##	245	AHCYL2	1.268908e-02	2.00100000
##	246	FAM207A	2.367056e-02	1.99800000
##	247	MCM10	8.922925e-03	1.99600000
##	248	NYAP2	8.588589e-03	1.99600000
##	249	COL11A1	4.959184e-05	1.99500000
##	250	PGM2L1	7.234973e-03	1.99500000
##	251	OR2W3	2.512594e-02	1.99300000
##	252	SOGA2	8.820000e-06	1.99100000
##	253	GBA2	1.131742e-02	1.99100000
##	254	TMEM182	1.933896e-02	1.99000000
##	255	CMAS	1.324885e-03	1.98500000
##	256	MGC50722	2.074176e-02	1.98300000
##	257	LINC00152	9.025070e-04	1.98000000

##	258	C1orf135	2.089041e-02	1.97900000
##	259	RSP04	1.013021e-03	1.97900000
##	260	EPHA5	4.781421e-02	1.97600000
##	261	SPCS3	4.731183e-04	1.97100000
##	262	TEX15	1.871013e-02	1.96900000
##	263	VWA5A	3.760446e-03	1.95900000
##	264	RTN2	1.734019e-02	1.95800000
##	265	SLC7A8	2.268041e-03	1.95700000
##	266	QPRT	5.859649e-05	1.95600000
##	267	CAMK1G	8.587174e-03	1.95200000
##	268	CADPS	4.076923e-04	1.95200000
##	269	TTPAL	6.897590e-04	1.94800000
##	270	RARRES1	3.556976e-02	1.94700000
##	271	KCNQ3	3.411079e-03	1.93900000
##	272	C10orf10	3.497831e-02	1.93900000
##	273	FAM54A	2.373868e-02	1.93500000
##	274	NPM2	4.894366e-04	1.93300000
##	275	TMEM132B	4.501880e-02	1.93300000
##	276	SEPT5	9.382716e-03	1.93000000
##	277	CDK14	2.292079e-04	1.92900000
##	278	C2orf55	2.223693e-02	1.92800000
##	279	RAB3A	5.842167e-03	1.92800000
##	280	FABP7	5.542312e-03	1.92600000
##	281	RAB3C	7.756813e-03	1.92400000
##	282	JAKMIP1	5.957983e-05	1.92000000
##	283	SPATS2	1.280335e-02	1.91800000
##	284	GALT	4.006039e-02	1.91600000
##	285	RUNDC3A	4.052974e-02	1.91600000
##	286	FAM110C	3.840991e-02	1.91500000
##	287	LRRC8A	2.348039e-04	1.91100000
##	288	SNX16	1.691837e-03	1.90600000
##	289	DPCD	2.897694e-02	1.90500000
##	290	OR7E156P	3.511039e-02	1.90500000
##	291	GARS	1.684211e-02	1.89900000
##	292	PPTC7	4.729242e-04	1.89900000
##	293	SLITRK5	1.508991e-02	1.89800000
##	294	CACNA2D2	8.922925e-03	1.89300000
##	295	KIAA1377	3.702665e-03	1.89200000
##	296	CEACAM4	7.352941e-03	1.89100000
##	297	MAP3K5	1.691837e-03	1.89000000
##	298	WNT3	2.402707e-03	1.89000000
##	299	TWIST1	8.954057e-03	1.88900000
##	300	LPGAT1	3.522788e-02	1.88600000
##	301	ANKRD34C	9.126679e-03	1.88400000
##	302	PCL0	3.308715e-03	1.88300000
##	303	ALPK3	3.589608e-02	1.88100000
##	304	DLGAP3	1.476826e-02	1.87800000
##	305	CYB5R1	5.202454e-03	1.87800000
##	306	HK1	6.180160e-03	1.87600000
##	307	LIN7A	1.869159e-03	1.87600000
##	308	ALDH4A1	6.071429e-03	1.87500000
##	309	LOC100129781	8.954057e-03	1.87400000
##	310	ANXA6	3.512337e-03	1.87300000
##	311	GRIA3	2.343333e-06	1.87000000

## 312	C1S	5.374150e-04	1.86600000
## 313	TAB3	5.043263e-03	1.86500000
## 314	APOL2	2.411802e-02	1.86500000
## 315	SMARCA1	3.388248e-02	1.85900000
## 316	SHOC2	8.029046e-03	1.85400000
## 317	SH3BGRL3	5.796178e-04	1.85200000
## 318	CCDC69	2.897694e-02	1.85100000
## 319	PDE6H	4.761905e-02	1.84900000
## 320	MSL3P1	1.268908e-02	1.84900000
## 321	ATP6V1G1	1.160494e-03	1.84700000
## 322	PSD3	1.511777e-03	1.84700000
## 323	MPZL1	4.106464e-04	1.84700000
## 324	RAP1GDS1	3.522788e-02	1.84500000
## 325	AHNAK2	3.874936e-02	1.84400000
## 326	FSTL4	8.827038e-03	1.84300000
## 327	CLTB	8.173375e-03	1.84300000
## 328	MSI2	3.804945e-03	1.84100000
## 329	ANKRD50	8.954057e-03	1.83900000
## 330	CDC6	4.057683e-02	1.83900000
## 331	GFER	1.230769e-02	1.83600000
## 332	LOC284801	1.047359e-02	1.83500000
## 333	GPR152	2.711761e-02	1.83500000
## 334	LOC100292489	3.701195e-04	1.83300000
## 335	SNX25	2.700893e-04	1.83200000
## 336	DPH3P1	2.691131e-02	1.83100000
## 337	DLGAP4	1.603448e-04	1.83100000
## 338	FLJ43860	1.539056e-02	1.82800000
## 339	PCNXL2	9.672131e-04	1.82600000
## 340	C11orf74	3.702665e-03	1.82500000
## 341	DIRAS3	1.420000e-08	0.45269353
## 342	PENK	1.592593e-05	0.45228403
## 343	CRH	1.169231e-06	0.45228403
## 344	SCGN	1.433333e-07	0.45105999
## 345	NEUROD2	2.226277e-03	0.44984256
## 346	CALB2	2.070000e-09	0.44943820
## 347	BAIAP3	1.433333e-07	0.44903458
## 348	CHODL	2.182266e-04	0.44822949
## 349	SST	3.147651e-04	0.44822949
## 350	PTPRT	7.571429e-08	0.44762757
## 351	IHH	3.290000e-03	0.44682752
## 352	KIT	1.448276e-04	0.44662796
## 353	NXPH2	1.952941e-04	0.44642857
## 354	MYH7	7.571429e-08	0.44622936
## 355	NPTX1	2.010526e-04	0.44603033
## 356	VIP	5.547945e-03	0.44503783
## 357	CARTPT	4.370925e-02	0.44503783
## 358	DMRTA2	1.898254e-02	0.44424700
## 359	SORCS3	9.956522e-06	0.44365572
## 360	ADRA1D	5.527523e-03	0.44365572
## 361	ARHGAP36	5.279188e-04	0.44326241
## 362	SLC6A4	1.870200e-02	0.44326241
## 363	HTR1E	9.642857e-05	0.44306602
## 364	EGFL6	7.631579e-06	0.44189129
## 365	F13A1	1.620295e-02	0.44169611

## 366	RPRM	2.030848e-03	0.44169611
## 367	NGB	3.363636e-05	0.44169611
## 368	CALB1	7.530864e-03	0.44150110
## 369	UNC5D	4.297521e-04	0.44130627
## 370	SLITRK6	3.147651e-04	0.44130627
## 371	KCNJ3	4.205128e-05	0.44111160
## 372	TPH2	4.292683e-02	0.44091711
## 373	RASGRP1	6.803170e-03	0.44091711
## 374	HIST1H1A	1.429577e-04	0.44033465
## 375	NDST3	6.978261e-04	0.44033465
## 376	KIAA1239	1.692525e-03	0.44014085
## 377	SYTL5	3.129032e-05	0.43975374
## 378	PLXNC1	8.561644e-05	0.43936731
## 379	SOX14	3.147651e-04	0.43917435
## 380	ADCYAP1	1.677445e-02	0.43917435
## 381	DRD5	1.169231e-06	0.43859649
## 382	ONECUT1	7.693920e-04	0.43840421
## 383	HTR2C	5.153061e-04	0.43840421
## 384	WNT4	4.431373e-05	0.43802015
## 385	KCNQ5	2.882788e-03	0.43725404
## 386	SORCS1	4.854881e-04	0.43687200
## 387	GULP1	2.362791e-04	0.43630017
## 388	NOS1	4.006116e-04	0.43572985
## 389	LPPR5	6.096774e-05	0.43572985
## 390	CALCR	5.004010e-03	0.43572985
## 391	NPTX2	4.085366e-04	0.43516101
## 392	FEV	6.803170e-03	0.43459365
## 393	NECAB2	3.211382e-03	0.43421624
## 394	PDGFD	5.582524e-04	0.43421624
## 395	COCH	5.593220e-04	0.43383948
## 396	C1QL3	1.027689e-02	0.43365134
## 397	GATA3	2.751606e-03	0.43365134
## 398	CHRNA3	2.081737e-03	0.43365134
## 399	CBLN4	3.479532e-03	0.43346337
## 400	TMEM215	2.038462e-03	0.43346337
## 401	DMBT1	8.778626e-04	0.43308792
## 402	COL21A1	8.291667e-05	0.43308792
## 403	DACT1	9.074468e-05	0.43308792
## 404	SPINK2	1.517592e-02	0.43252595
## 405	PCP4L1	8.084677e-04	0.43233895
## 406	S1PR5	4.205128e-05	0.43233895
## 407	KCNAB1	1.237705e-04	0.43177893
## 408	C10orf82	8.618677e-04	0.43159258
## 409	SLC30A3	2.255730e-03	0.43159258
## 410	LAMB1	8.212851e-04	0.43084877
## 411	BEAN1	1.358209e-04	0.42955326
## 412	GALR1	9.648799e-04	0.42936883
## 413	USH1C	1.653595e-04	0.42900043
## 414	GRM8	1.836735e-03	0.42863266
## 415	SPHKAP	1.271318e-04	0.42863266
## 416	LGR5	1.271318e-04	0.42844901
## 417	TSPAN8	2.720430e-03	0.42808219
## 418	CDH22	4.456140e-05	0.42771600
## 419	SLC17A6	9.463199e-03	0.42698548

## 420	SEMA3D	1.945455e-04	0.42680324
## 421	CNR1	3.987730e-04	0.42662116
## 422	PCDH11Y	5.536585e-04	0.42625746
## 423	MYO1E	4.693333e-04	0.42589438
## 424	OXR1	1.653595e-04	0.42589438
## 425	LRP2	9.074468e-05	0.42571307
## 426	NIPAL4	3.503185e-04	0.42553191
## 427	TCERG1L	4.881890e-04	0.42517007
## 428	DGKK	2.084158e-04	0.42480884
## 429	HN1L	1.392857e-04	0.42462845
## 430	MADCAM1	4.447439e-04	0.42426814
## 431	ZMAT4	2.447796e-03	0.42390844
## 432	SEMA5A	9.969697e-05	0.42372881
## 433	IGFBP3	2.223587e-03	0.42354934
## 434	RBP4	5.225653e-03	0.42337003
## 435	EHF	3.115523e-04	0.42301184
## 436	TAC1	8.353081e-03	0.42069836
## 437	MBOAT1	4.830718e-03	0.42052145
## 438	FNDC9	1.825613e-03	0.42034468
## 439	PPP1R17	5.901521e-03	0.42016807
## 440	RPRML	1.665313e-02	0.41999160
## 441	PCDH11X	8.747592e-04	0.41893590
## 442	LINC00473	2.819149e-03	0.41893590
## 443	DACH1	1.000000e-04	0.41858518
## 444	SCN1A	4.431373e-05	0.41841004
## 445	UGT8	3.207237e-04	0.41718815
## 446	AQP3	2.497162e-03	0.41701418
## 447	EGR1	2.999703e-02	0.41649313
## 448	LOC149134	6.518692e-04	0.41631973
## 449	CCNE2	1.952941e-04	0.41459370
## 450	ENPP2	3.110294e-04	0.41442188
## 451	MYO5B	7.286432e-03	0.41356493
## 452	NTNG1	5.976190e-04	0.41288192
## 453	PRKG2	1.377676e-03	0.41220115
## 454	FA2H	4.854881e-04	0.41186161
## 455	FRMD5	2.060914e-04	0.41135335
## 456	PMCH	1.126472e-02	0.41034058
## 457	CHRM2	4.600505e-03	0.40983607
## 458	PCSK6	8.747592e-04	0.40899796
## 459	ZSCAN12	1.355906e-03	0.40866367
## 460	POU4F1	1.303738e-02	0.40816327
## 461	TMEM200A	1.963351e-03	0.40783034
## 462	GNG4	9.909256e-04	0.40766408
## 463	ANLN	1.955556e-04	0.40716612
## 464	SYT6	2.089172e-03	0.40600893
## 465	SCD	7.045455e-05	0.40502228
## 466	FGF14	2.381166e-04	0.40502228
## 467	RYR2	1.059649e-03	0.40469446
## 468	NGEF	1.972678e-04	0.40322581
## 469	ALK	7.173448e-04	0.40322581
## 470	RAD54L	2.214022e-03	0.40290089
## 471	LDB3	1.271318e-04	0.40225261
## 472	GAD2	2.494279e-03	0.40160643
## 473	SGK2	2.339623e-04	0.40064103

## 474	C21orf91	1.358209e-04	0.40032026
## 475	SOX2-OT	2.421053e-03	0.40032026
## 476	ADRA1B	1.825613e-03	0.40016006
## 477	MOBP	8.291667e-05	0.39968026
## 478	C1QL1	1.160535e-03	0.39968026
## 479	TMTC2	3.147651e-04	0.39936102
## 480	KLK6	2.570833e-04	0.39904230
## 481	CNTNAP4	2.732283e-04	0.39777247
## 482	LOC100292909	1.057338e-02	0.39698293
## 483	SYN2	8.618677e-04	0.39666799
## 484	ADAMTS9	1.436170e-03	0.39635355
## 485	SAMD3	6.786427e-03	0.39478879
## 486	LOC344595	2.010526e-04	0.39447732
## 487	TMEM139	5.949458e-03	0.39447732
## 488	TMEM233	9.973046e-03	0.39339103
## 489	ZEB2	4.431373e-05	0.39292731
## 490	A3GALT2P	3.252525e-03	0.39277298
## 491	ADAMTS4	4.911003e-03	0.39215686
## 492	TF	1.366352e-03	0.39154268
## 493	ADAMTS1	8.519389e-03	0.39138943
## 494	CNDP1	4.776490e-03	0.39032006
## 495	EMILIN2	1.838509e-04	0.39016777
## 496	TMEM144	2.214022e-03	0.38910506
## 497	PRSS3P2	1.281810e-02	0.38744673
## 498	SMAD9	4.158590e-03	0.38729667
## 499	PCOLCE2	7.885010e-04	0.38714673
## 500	CAPN13	8.618677e-04	0.38699690
## 501	KLHL4	5.288221e-04	0.38684720
## 502	PLP1	1.377676e-03	0.38624952
## 503	SSTR2	4.693333e-04	0.38595137
## 504	SIM1	4.137760e-02	0.38565368
## 505	KCNA5	1.522388e-03	0.38550501
## 506	RAB3B	2.977444e-04	0.38550501
## 507	KCNJ2	2.726190e-04	0.38491147
## 508	NNAT	2.378698e-03	0.38476337
## 509	FIGNL2	4.164306e-04	0.38476337
## 510	FOS	3.722414e-02	0.38476337
## 511	SLC35F4	2.552301e-04	0.38417211
## 512	NEUROD6	1.234983e-02	0.38299502
## 513	TRPM6	1.953757e-04	0.38284839
## 514	STC1	2.214022e-03	0.38167939
## 515	OCA2	1.115681e-02	0.38124285
## 516	EVI2A	4.933333e-05	0.38095238
## 517	CREB3L3	1.318224e-02	0.38066235
## 518	RBP7	1.916329e-03	0.37965072
## 519	SYNJ2	3.147651e-04	0.37950664
## 520	GDA	6.635199e-03	0.37936267
## 521	GPR37	6.454483e-03	0.37921881
## 522	NMBR	2.497175e-03	0.37878788
## 523	IGFBP5	3.283283e-03	0.37764350
## 524	KCNG1	3.919753e-04	0.37764350
## 525	MAP6D1	3.147651e-04	0.37707391
## 526	SLC45A3	4.417344e-04	0.37707391
## 527	KLK10	1.903861e-02	0.37622272

## 528	GPR62	1.377676e-03	0.37593985
## 529	ADRA1A	4.164306e-04	0.37593985
## 530	ERMN	3.147651e-04	0.37551633
## 531	LDLRAP1	6.836283e-04	0.37523452
## 532	C21orf62	5.332817e-03	0.37453184
## 533	TOX	8.939828e-03	0.37397158
## 534	ABCA8	8.353081e-03	0.37174721
## 535	GAD1	1.906203e-02	0.37133309
## 536	YEATS2	4.342561e-03	0.37091988
## 537	MOG	7.949029e-03	0.37064492
## 538	ENPP6	2.231834e-02	0.37064492
## 539	LOC653051	2.453390e-04	0.37050760
## 540	OPRM1	1.746916e-02	0.37037037
## 541	CDH4	4.771784e-03	0.36954915
## 542	LOC100129122	2.677596e-03	0.36616624
## 543	ITGA2	7.959596e-04	0.36603221
## 544	BTBD16	1.184928e-02	0.36603221
## 545	STEAP1B	2.728249e-03	0.36483035
## 546	SLC31A2	5.821768e-03	0.36469730
## 547	HIST1H3A	3.495146e-04	0.36416606
## 548	ARC	2.689633e-02	0.36350418
## 549	DMRTC2	1.764331e-04	0.36337209
## 550	RASGRF1	2.284360e-04	0.36271309
## 551	LOC643475	1.461405e-02	0.36218761
## 552	CCDC152	5.657100e-03	0.36192544
## 553	SUSD2	6.746032e-05	0.36192544
## 554	B3GAT2	3.522618e-03	0.36101083
## 555	SHROOM4	1.597315e-04	0.35958288
## 556	SLC17A8	1.754665e-02	0.35906643
## 557	PLD1	2.384956e-04	0.35829452
## 558	B3GALTL	1.025093e-02	0.35829452
## 559	SEC14L5	1.318224e-02	0.35778175
## 560	RAMP3	2.666455e-02	0.35739814
## 561	SPOCK1	1.972826e-04	0.35727045
## 562	RTP1	4.776490e-03	0.35676061
## 563	CARNS1	1.441048e-02	0.35561878
## 564	COL5A2	5.332817e-03	0.35448423
## 565	MEIS2	1.119863e-03	0.35435861
## 566	BACE2	1.237705e-04	0.35373187
## 567	BCAS1	6.965066e-04	0.35260931
## 568	C10orf128	1.785714e-03	0.35260931
## 569	C4orf32	1.736695e-03	0.35236082
## 570	PHLDB1	6.361502e-04	0.35223670
## 571	ST18	3.089598e-03	0.35198874
## 572	HAPLN4	1.322049e-02	0.35026270
## 573	FOXA2	9.074468e-05	0.35014006
## 574	HTR1A	1.853715e-02	0.34952814
## 575	GCHFR	4.150442e-03	0.34904014
## 576	HOXD3	1.076280e-02	0.34806822
## 577	TAS2R5	3.089598e-03	0.34770515
## 578	CLCA4	2.255730e-03	0.34650035
## 579	SEMA4D	3.160336e-02	0.34638033
## 580	NELL1	1.126472e-02	0.34447124
## 581	GRIK1	2.284360e-04	0.34176350

## 582	C18orf56	9.054670e-03	0.34153005
## 583	BTG3	1.271318e-04	0.34048349
## 584	PTPRK	1.564327e-03	0.33875339
## 585	PTPRH	5.589354e-03	0.33772374
## 586	SLC5A11	3.781049e-03	0.33658701
## 587	PROX1	2.467062e-02	0.33658701
## 588	CMTM5	2.959184e-02	0.33568312
## 589	HOXD1	2.373567e-02	0.33400134
## 590	TRIM59	6.454483e-03	0.33277870
## 591	TP53TG5	6.375000e-06	0.33123551
## 592	GEM	6.836283e-04	0.33101622
## 593	STAC2	2.173760e-02	0.33057851
## 594	C12orf76	1.955381e-03	0.32959789
## 595	CDKN1C	1.128205e-03	0.32754668
## 596	RASSF2	2.779923e-04	0.32690422
## 597	PLK5	3.359684e-03	0.32658393
## 598	LOC100289058	1.266603e-02	0.32626427
## 599	CD44	1.245981e-03	0.32499188
## 600	KCNC2	2.151899e-03	0.32414911
## 601	GRM7	2.388235e-03	0.32310178
## 602	C22orf42	9.085747e-03	0.32216495
## 603	EPB41L4B	1.162791e-03	0.32113038
## 604	ATP11A	3.627760e-04	0.32071841
## 605	NDE1	7.505330e-04	0.32010243
## 606	MEI1	4.498106e-02	0.31816736
## 607	GAS7	5.332817e-03	0.31766201
## 608	TEX26	6.129944e-03	0.31705770
## 609	HAPLN2	1.163836e-02	0.31675641
## 610	CDH13	1.687943e-03	0.31645570
## 611	EGFLAM	3.362286e-02	0.31230481
## 612	TMEM125	4.386266e-03	0.31046259
## 613	KCTD8	1.952941e-04	0.31017370
## 614	SNTG2	2.794896e-02	0.30921459
## 615	TSPAN18	2.121982e-03	0.30778701
## 616	SHOX2	3.601408e-02	0.30590395
## 617	GLDN	1.160580e-02	0.30581040
## 618	DSCAML1	3.843750e-04	0.30571691
## 619	KCNIP1	5.755771e-03	0.30284676
## 620	GLTP	4.164306e-04	0.30165913
## 621	TMEM235	1.772152e-02	0.30147724
## 622	KCNC1	1.242206e-02	0.29994001
## 623	GALNT14	1.377676e-03	0.29904306
## 624	PTPRD	2.284360e-04	0.29832936
## 625	SLC26A9	9.539474e-03	0.29824038
## 626	STXBP6	1.678521e-03	0.29806259
## 627	CHRNA5	2.835545e-02	0.29717682
## 628	HMSD	1.596708e-02	0.29568303
## 629	KCNJ2-AS1	7.286432e-03	0.29498525
## 630	GRIP2	8.012048e-03	0.29359953
## 631	NKX6-1	5.332817e-03	0.28977108
## 632	LOC100270746	3.359684e-03	0.28910090
## 633	VSTM1	3.792833e-02	0.28776978
## 634	TCP11L2	1.071804e-03	0.28719127
## 635	ELOVL1	1.167665e-02	0.28661508



## 636	GJB1	1.473368e-02	0.28344671
## 637	CHST3	1.984334e-03	0.28272547
## 638	C17orf110	7.847946e-03	0.28248588
## 639	CCDC147	4.956849e-02	0.27995521
## 640	KCNE4	4.911003e-03	0.27831895
## 641	CRHBP	2.045124e-02	0.27777778
## 642	ZNF608	1.359712e-04	0.27762354
## 643	ENC1	1.377676e-03	0.27196084
## 644	POM121L1P	4.046553e-03	0.26917900
## 645	PDE1C	5.083665e-03	0.26490066
## 646	TJP2	1.267739e-02	0.26184865
## 647	SPSB2	6.831461e-04	0.26143791
## 648	SLED1	7.906504e-04	0.25839793
## 649	CHN2	2.214022e-03	0.25813113
## 650	NKX6-2	2.089172e-03	0.25693731
## 651	LOC100131170	6.129944e-03	0.25588536
## 652	APLNR	1.468804e-02	0.25529742
## 653	GPIHBP1	4.910714e-03	0.25342119
## 654	EDIL3	4.398907e-04	0.25233409
## 655	CDK18	1.237705e-04	0.24764735
## 656	ADRA2A	5.755771e-03	0.24420024
## 657	SH3TC2	1.332725e-02	0.24414062
## 658	DMRT2	1.963351e-03	0.24207214
## 659	PRR5L	6.803170e-03	0.23640662
## 660	DEPDC7	1.678521e-03	0.23457659
## 661	ZNF488	7.471980e-03	0.23424690
## 662	HIST1H4A	6.454483e-03	0.23293734
## 663	N4BP2L1	1.306836e-03	0.23201856
## 664	COL25A1	3.637218e-03	0.22904260
## 665	LLGL1	1.131498e-02	0.21877051
## 666	VAT1L	9.780622e-04	0.21105952
## 667	PPP1R14A	2.563818e-03	0.20942408
## 668	OPCML	3.670412e-03	0.20673971
## 669	ACTG2	4.430077e-02	0.20508614
## 670	ERBB3	7.022107e-03	0.19948135
## 671	BOK	3.414873e-03	0.19762846
## 672	SLC6A17	1.903861e-02	0.19447686
## 673	ATP10B	1.051797e-02	0.19230769
## 674	ACSL1	2.214022e-03	0.19029496
## 675	RBFOX1	1.677445e-02	0.18993352
## 676	SEMA3C	6.216028e-03	0.18670650
## 677	ANKRD65	2.779923e-04	0.17636684
## 678	ARHGEF2	2.213343e-02	0.17571604
## 679	GLRA2	1.816399e-02	0.16980812
## 680	WIPF1	2.453390e-04	0.16829350
## 681	PLIN3	1.838509e-04	0.16758840
## 682	GOLT1A	2.121539e-02	0.16553551
## 683	PRRG1	1.952941e-04	0.15987210
## 684	RTN4R	2.030848e-03	0.15900779
## 685	MYH11	2.978787e-02	0.15615240
## 686	NKX2-2	2.453390e-04	0.13732491
## 687	LOC728401	1.724690e-02	0.13468013
## 688	ZNF735	2.629228e-02	0.12899897
## 689	GREM1	3.584035e-02	0.11924636

## 690	BLID	2.329749e-03	0.11708231
## 691	KIAA1024L	1.140974e-02	0.11341726
## 692	KIAA1683	2.552301e-04	0.10998680
## 693	ATF3	1.471225e-02	0.09184423
## 694	P2RX7	1.175299e-02	0.08987956
## 695	MATN1-AS1	2.563818e-03	0.08946144
## 696	CLMN	2.226277e-03	0.07949758

Among the 696 genes found to be significantly upregulated or downregulated in the SNpc relative to the VTA were genes linked to dopamine, including dopamine receptor 2 (DRD2, BH-corrected  $p = 1.095541e-04$ , FC = 3.341), dopamine receptor 5 (DRD5, BH-corrected  $p = 1.169231e-06$ , FC = 0.2442002), dopamine active transporter (SLC6A3, BH-corrected  $p = 8.355197e-03$ , FC = 4.784) and vesicular monoamine transporter (SLC18A2, BH-corrected  $p = 0.01238258$ , FC = 4.656). The SLC18A2 gene codes for the VMAT2 enzyme, which transports dopamine into monoaminergic vesicles to prevent it from oxidizing [1]. VMAT2 activity has been shown to be inversely proportional to neuromelanin levels, so one may assume that higher SLC18A2 expression would contribute to less neuromelanin. However, the opposite case is evident here. With fold change = 4.656, SLC18A2 is actually upregulated in the SNpc, which contains neuromelanin, with respect to the VTA, which does not. This suggests that there may be more at play in neuromelanin formation than just the action of VMAT2.

Ideally, further analysis should be done on these 696 genes to determine which molecular pathways they are involved in. This includes gene set enrichment analysis and assigning gene ontology terms to groups of these genes. Due to time limitations, this could not be accomplished within the hackathon.

## REFERENCES

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